SURFACE SHIP CLASSIFICATION USING MULTIPOLARIZATION, MULTIFREQUENCY SKY-WAVE RESONANCE RADAR

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Experiments investigating the classification of surface ships using processed radar returns are described. The calibrated and scaled backscatter measurements of scale-model ships at several aspect and elevation angles are used to establish a catalog representing the HF sky-wave resonance region radar returns of actual ship targets. The performance of both the nearest neighbour algorithm, (using frequency domain data), and a correlation algorithm (using time domain data), is investigated. The effects of wave polarization, aspect angle, elevation angle and other key parameters are examined. The consequences of introducing forced errors into the estimates of aspect and elevation angle are studied. A novel feature set employing the ratio of vertically and horizontally polarized radar returns is described, and its classification performance is examined.

In general, classification is found to be very much dependent on the particular aspect angle and polarization of interest. The time domain algorithm, vertical polarization, and bow and stern aspects are the parameters which yield best all-round classification performance. Increasing the number of classification frequencies improves performance, but only to a limit. Errors in aspect angle and especially in elevation angle are found to significantly degrade classification performance. A number of suggestions for improving the classification process are discussed.
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CHAPTER I
INTRODUCTION

The general radar problem has, conventionally, been one of finding the spatial location, or velocity or both, of some target. This information can be extended to include a knowledge of the target's identity, if the operating frequencies of the radar are properly chosen. More specifically, if the wavelength of the radar energy is comparable to the maximum dimension of the target (i.e., in the resonance region), then certain essential information, relating to the target's dimensions and shape, will be imbedded in the radar return. Radars operating in the HF band (3 to 30 MHz) have wavelengths ranging from 10 to 100 m; hence missiles, aircraft and ships are potential candidates for resonance region target identification. Furthermore, frequencies in the HF band are capable of propagation over large distances (perhaps up to 4000 km) and thus, with an HF radar system we have, prima facie, the potential for identifying ships and aircraft, etc., at very long ranges. Figure 1.1 shows a diagram of an HF radar target classification system.
Chen [1] studied the available techniques for measuring and processing the radar returns of targets in the HF band. He concluded that sufficient signal-to-noise ratio could be attained after processing (in a reasonable amount of time) to permit target classification. He developed algorithms to implement target classification and, in conjunction with simulated radar returns of ships and aircraft, studied classification performance as a function of post-processed signal-to-noise ratio.

Chen's investigations involved the use of data for ships and aircraft at zero degrees elevation angle, using vertical polarization. Hence, for the case of ships, the data was representative of surface-wave radar returns. HF radiation can also propagate by means of sky-waves, that is, by ionspheric refraction, and target classification for this mode of propagation is the subject of investigation in this report. Classification at various polarizations and aspect angles are also considered. Hence, the problem addressed here is one of studying the classification of ship targets in a representatively noisy environment, using simulated multiple-frequency, multiple-polarization, sky-wave resonance radar returns.

Chapter II discusses the general nature and limitations of the HF sky-wave radar system and how these relate to the measurement of phase and amplitude returns. Chapter III describes the generation of a database of ship amplitude and phase returns, with particular reference to problems incurred by the use of a groundplane to simulate the surface of the sea. Chapter IV details the classification algorithms used in
experiments. Chapter V describes the experimental procedure and other
considerations, such as the specifications of the noise model, which
relate to the HF resonance radar detection problem. Chapter VI
summarizes the results of the various classification experiments and
draws conclusions from these and other observations. Representative
curves of misclassification percentage versus post-processing
signal-to-noise ratio are included in Chapter VI, but the majority of
these plots are appended, owing to their sheer bulk. Chapter VII
presents a summary of the work, emphasizing the more important findings
of Chapter VI, with conclusions and a set of recommendations for future
work.
Figure 1.1. Block diagram of target classification system. The catalog contains returns of some preselected targets ($\hat{A}, \hat{\theta}$). The output of the signal processor is a set of amplitude and phase returns of the unknown target. (from [1])
CHAPTER II

SKY-WAVE RESONANCE RADAR

2.1 INTRODUCTION

Resonance radar is a technique whereby characteristic information of an object is gained by illuminating it with electromagnetic energy of certain wavelengths. These frequencies are commonly called resonance region frequencies, and are defined by the relation \( L/\lambda = 1 \), where \( L \) is the maximum dimension of the target and \( \lambda \) is the wavelength of the energy [3,4]. Hence, resonance radar extends our knowledge of a target beyond the conventional information concerning presence, location and velocity, and tells us something of the target's identity.

The principle of resonance region radar is applicable across the entire electromagnetic spectrum. For example, ships and aircraft can be interrogated with high frequencies (HF) [5,10], ground vehicles with very high frequencies (VHF) [20], satellites with ultra-high frequencies (UHF) and insects with microwaves. The technique is applicable to acoustic environments so that underwater objects could be investigated with resonant region SONAR sound waves. This report is concerned with HF resonance region radar used for the detection of ships.
Many of the first radar systems operated in the HF band (3-30 MHz). This is because microwave devices, such as the magnetron, were not available at that time. Generally it is more advantageous, in terms of directivity for a given antenna size and unhampered propagation [8], to use microwave frequencies for radar detection. However, HF radiation has the property of propagation beyond the line of sight by either surface-waves or ionospheric refraction (sky-waves). In fact, ship detection has been accomplished using a sky-wave radar system under a variety of atmospheric (and sea state) conditions by the Naval Research Laboratory [9]. Furthermore, wavelengths in the HF region (ranging from 10 to 100 m), are comparable to the dimensions of ships, missiles and aircraft, and are therefore suited to the resonance region techniques mentioned earlier. Chen [1] showed that reliable classification of ships can be achieved by processing representative resonance region multi-frequency radar returns. Ksienki and Lin [10] demonstrated similar results for aircraft. This chapter summarizes current techniques for measuring and processing HF radar returns, and discusses some of the limitations pertinent to sky-wave propagation.
2.2 HF RADAR SPECIFICATIONS

Table 2.1 lists and compares the specifications of sky-wave and surface-wave HF radar systems. The second is characterized by a range of 200-400 km. The limitation in range (compared with sky-wave radar) is due to the exponential attenuation of surface waves [8]. The work on the classification of ships and aircraft at an elevation angle of 0 degrees [1], is applicable to a surface-wave radar system. The data used in the experimental work of this report represents radar returns of ships at elevation angles above 0°, hence, sky-wave propagation and, in particular, ionospheric refraction is considered here.

The maximum range of a sky-wave radar is stated as 4000 km [9]. This value applies to a single skip (skip distance is the distance between radar and target after refraction from the ionosphere) and could be obtained by using an elevation angle of about 8° and a frequency of 23 MHz [9]. In general, it must be remembered that range is a function of both the ionospheric condition and the propagation frequency.

The antenna of a sky-wave radar must have a large aperture (about 1-2 km) in order to provide high gain and narrow azimuthal beamwidth. High gain (20 to 30 dB) is necessary to ensure adequate signal level in the noisy HF environment. Narrow azimuthal beamwidth is necessary for high azimuthal resolution and, in conjunction with range resolution, determines the cell size of a particular scan. Range resolution is limited by the instantaneous bandwidth, B, of the ionospheric
<table>
<thead>
<tr>
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<th>SKY-WAVE RADAR</th>
<th>SURFACE-WAVE RADAR</th>
</tr>
</thead>
<tbody>
<tr>
<td>TRANSMITTED POWER</td>
<td>SEVERAL HUNDRED kW</td>
<td>HIGHER</td>
</tr>
<tr>
<td>RANGE</td>
<td>500 km to 4000 km</td>
<td>200-400 km</td>
</tr>
<tr>
<td>Antenna gain</td>
<td>20 to 30 dB</td>
<td>same</td>
</tr>
<tr>
<td>Antenna horizontal length</td>
<td>about 1 km</td>
<td>about 1 km</td>
</tr>
<tr>
<td>Waveform</td>
<td>can be pulsed sinusoidal waves</td>
<td>can be pulsed sinusoidal waves</td>
</tr>
<tr>
<td>Bandwidth</td>
<td>5 to 100 kHz</td>
<td>can be wider</td>
</tr>
<tr>
<td>pulse width</td>
<td>200 us to 10 us</td>
<td>can be narrower</td>
</tr>
<tr>
<td>Range resolution</td>
<td>2 to 40 km</td>
<td>can be smaller</td>
</tr>
<tr>
<td>Doppler resolution</td>
<td>0.01 to 0.1 Hz</td>
<td>can be less than 0.1 Hz</td>
</tr>
<tr>
<td>PRF</td>
<td>about 30 Hz</td>
<td>about 300 Hz</td>
</tr>
<tr>
<td>Targets</td>
<td>aircraft, ships, missiles</td>
<td>same</td>
</tr>
<tr>
<td>Antenna Azimuth</td>
<td>about 1°</td>
<td>about 1°</td>
</tr>
<tr>
<td>beamwidth</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>
propagation path, and is given by [8]

$$\Delta R = \frac{C}{Z B \cos \phi} \quad (2.1)$$

where $C$ is the speed of light, and $\phi$ is the elevation angle. For example, if $B = 100$ khz and $\phi = 30^\circ$, then $\Delta R = 1.7$ km. Figure 2.1 shows the effective cell size for a particular range $R$. Azimuth resolution is given by [8]

$$\Delta A_z = R \Theta B \quad (2.2)$$

where $\Theta B$ is the azimuth beamwidth. Using a $1^\circ$ beamwidth antenna at a range of 2000 km yields an azimuth resolution of 35 km. Hence, a typical minimum cell size has an area on the order of 60 km$^2$.

The work of this report is concerned with the identification of single targets. It is quite conceivable that several ships could amass in a typical resolution cell and the resulting radar returns would then be a complex representation of a number of ships. This problem is considered as too complicated for a primary analysis, and thus is not addressed here.

Cell size also affects the signal-to-clutter ratio since a signal in a larger cell will inherently acquire more clutter. Clutter can be reduced by filtering in the doppler domain, and the resolution attainable in doppler frequency is inversely proportional to the coherent observation time.

The pulse repetition frequency (PRF) must be kept low to avoid range ambiguities. Maximum unambiguous range is given by [8]

$$R_{un} = \frac{C}{2 \cdot PRF} \quad (2.3)$$
Figure 2.1. The resolution cell of a sky-wave radar, shown by the shaded area (from [1]). Note here that

\[ \phi = \text{elevation angle} \]
\[ \tau = \text{pulse width} \]
\[ C = \text{speed of light} \]
\[ R = \text{wavepath length} \]
\[ \Theta_B = \text{antenna azimuth beam width} \]
The PRF for an unambiguous maximum range of 4000 km is 37 Hz. A low PRF can lead to doppler ambiguities [8] and generally a compromise must be made between range and doppler ambiguity. The narrow bandwidth criterion imposed by the dispersive nature of the ionosphere requires that pulse widths should be about 10 μs. For a square pulse of width $\tau$, the null-to-null bandwidth is given by

$$B = 1/\tau$$  \hspace{1cm} (2.4)

A narrow pulse width is an aid to improving signal to clutter ratio, however a long pulse is important for achieving the energy required for long-range detection.

2.3 MEASUREMENT OF HF AMPLITUDE AND PHASE RETURNS

The amplitude return, $A$, is defined as the square root of the radar cross section $\sigma$, where $\sigma$ is given by

$$\sigma = \frac{\text{Scattered Power}}{\text{Incident Power Density at the Target}}$$  \hspace{1cm} (2.5)

$$A = \sqrt{\sigma} = \sqrt{\frac{p^2}{4\pi^3 R^4 L_p^2 L_s}}$$  \hspace{1cm} (2.6)

where:

- $P_T$ = Transmitter power
- $G_T$ = Transmitting antenna gain
- $G_R$ = Receiving antenna gain
- $\lambda$ = Wavelength of propagating energy
\[ G_A = \text{Receiver power gain} \]
\[ p^r = \text{Power received} \]
\[ R = \text{Range of target} \]
\[ L_p = \text{One-way propagation loss} \]
\[ L_s = \text{System loss} \]

Equation 2.6 shows that the amplitude return depends on many parameters, some of which are difficult to estimate. The propagation loss, \( L_p \), for example, is dependent on the continually changing state of the ionosphere for sky-wave radar. To obviate the need for such estimates, a reference signal return of another object close to the target is often used. The radar cross section of the reference target is given by

\[ \sigma_{\text{ref}} = \frac{p^\text{ref}(4\pi)^3R^4L_p^2L_s}{P_T G_T G_R \lambda^2 G_A} \]  

(2.7)

The range and propagation loss in equations 2.6 and 2.7 are approximately equal, and other parameters \( P_T, G_T, G_R, G_A \) and \( L_s \) are similar by virtue of the fact that the reference and the target are illuminated at the same time and are thus contained in the same resolution cell. Radar cross section can then be defined as

\[ \sigma = \frac{p^r}{p^\text{ref}} \sigma_{\text{ref}}. \]  

(2.8)
$p_r$ and $p_{ref}$ are the received powers of the target and reference signals respectively, and $\sigma_{ref}$ is estimated from a priori information or theoretical calculations.

Trizna [11] proposed a method for calibrating a target cross section by using the sea scatter as a reference. Essentially, the sea surface behaves like a diffraction grating and the moving water waves impart a doppler shift to the incident electromagnetic wave. Figure 2.2 shows a representative doppler spectrum for a ship and its reference sea echo. The doppler shifts depend upon the radial components (i.e., with respect to the direction of radar wave propagation) of the velocities of the ship and of the sea. The theoretical doppler spectrum of wind waves for a given sea state can then be computed using 1st and 2nd order radar cross section terms of the sea, in the manner described by Maresca and Barnum [12]. The calculated sea scatter cross section, $\sigma_c$, at a doppler shift of $f_r$ Hz is then used in the calibration equation (Equation (2.8)) to find the calibrated target cross section

$$\sigma_t = \sigma_s - (\sigma_r - \sigma_c) \quad \text{dBm}^2 \quad (2.9)$$

Usually the reference return is on the order of 20 to 40 dB greater than the ship return. Typical figures taken from [11] give $\sigma_s = 40 \ \text{dBm}^2$, $\sigma_r = 62 \ \text{dBm}^2$ and $\sigma_c = 51 \ \text{dBm}^2$ for a frequency of 21.8 MHz, vertically polarized, at 1600 km range, using a 17 second integration time. The resulting ship cross section, about 30 dBm$^2$ is typical of one of the smaller ships in the measured data set, for an aspect angle of 0° using vertical polarization. It is also
Figure 2.2. A doppler spectrum is shown in which a ship target return is separated from the clutter. The reference here is the larger first order sea scatter. The doppler frequency is normalized to the 1st order sea scatter doppler frequency; that is $f_r = -1$ (from [11]).

evident from Figure 2.2 that certain combinations of wave and ship velocity will render the ship and reference returns indistinguishable. Maresca and Barnum [12] defined a blind doppler frequency as one for which $\sigma_c > \sigma_t -10 \text{ dB}$. They concluded that for ships with cross sections in the order of 50 $\text{dBm}^2$, there were very few combinations of ship and wave velocity that resulted in blind doppler frequencies. For ships with cross sections of about 30 $\text{dBm}^2$, detection is contingent on the sea state. In general, the result depends on the frequency used, and the above problems can be averted by the use of higher operating frequencies when available. Of course, it is impossible to completely characterize a ship with a single radar cross section. A ship's cross section
depends on its size and orientation, as well as the frequency and polarization used. Typical cross sections might vary from 20 $\text{dBm}^2$ to over 70 $\text{dBm}^2$ and consequently the magnitude of a ship's radar returns, for a given orientation, is a characteristic of that target.

The above discussion pertains to the use of sea scatter as a reference in the calibration process. Sea scatter may not be the only reference available; for example, another ship in the immediate vicinity with an on-board repeater might serve as a reference. Furthermore, it is possible that the use of V/H polarization, discussed in Chapter VI, Section 6.5, might obviate the need to use a reference altogether.

2.4 MEASUREMENT OF PHASE RETURNS

The total phase shift, comprising the path, target and equipment phase shifts, can be derived from the demodulated received carrier. Taking the ratio of in-phase ($E_I$) and quadrature ($E_Q$) components yields

$$
\theta = \tan^{-1} \frac{E_Q}{E_I} = \frac{4\pi R}{\lambda} + \phi_i + \phi_e
$$

(2.10)

where

$4\pi R =$ Wave path length for a range $R$

$\phi_i =$ Intrinsic target phase

$\phi_e =$ Equipment phase.

In practice, it is difficult to accurately estimate the range. However, a differential method [13] can be employed which exploits the fact that
the total phase shift in the Rayleigh region is independent of the
co-polarized scattering matrix terms.

If $R = R_0$ is the nominal range in the Rayleigh region, then the
intrinsic phase is given by

$$\phi_i = \frac{\lambda \theta - \lambda_0 \theta_0}{\lambda}$$  \hspace{1cm} (2.11)

where $\theta_0$ is the total phase shift for a Rayleigh wavelength $\lambda_0$.
Unfortunately, for sky-wave propagation, $R$ is not equal to $R_0$ owing to
the frequency dependent path lengths through the ionosphere.
Furthermore, the Rayleigh region return is generally lower in amplitude
than a resonance region return, and may not be measurable. This would
suggest the choice of subtrahend wavelength much closer to the original
wavelength $\lambda$ to minimize the dependence of radar cross sections on path
lengths. This technique is described in a subsequent chapter.

In conclusion, it is difficult to recover the intrinsic phase from
a sky-wave radar return. However, this does not preclude the use of
phase data in subsequent experimental classification procedures.

2.5 CHOICE OF FREQUENCIES IN SKY-WAVE RADAR

There are a number of factors which influence the choice of
operating frequencies in sky-wave resonance radar. The most important
of these is the ionosphere [9]. The electron density of this medium
varies with height and with the solar illumination angle. Lower
electron densities require lower frequencies for refraction, and vice
versa. Consequently, elevation angle, and hence range, is dependent on
the condition of the ionosphere. The ionosphere is a dispersive medium (i.e., the velocity of a wave is a function of its frequency) and this limits the instantaneous bandwidth (or minimum pulse width) and the frequency sampling interval. A bandwidth of 100 KHz which corresponds to a pulse width of 10 µs is considered to be the maximum signal bandwidth [11].

Certain areas of the HF spectrum are subject to interference from other HF band signals, to deviative absorption in the ionosphere, and to natural forms of interference such as auroral ionization. This may necessitate a choice of frequencies that avoids the bands where these phenomena are prevalent.

Maresca and Barnum [12] showed that the sea state can be a critical factor in the detection of small ships (those with radar cross section approximately 30 dBm²). Higher frequencies may be used to avoid this problem by increasing the doppler separation between the sea echo (or Bragg return), and the ship return.

In addition to the above limitations, the operating frequencies should be in the resonance region of the target. This is, approximately, the region between \( 1 < \frac{L}{\lambda} < 10 \), where \( L \) is the maximum dimension of the target. The 6 ships used in the classification experiments discussed below average about 150 m in length. This defines an operating frequency band of 2 to 20 MHz. Furthermore, lower frequencies are preferable because there is a more distinct and reliable variation of radar cross section per unit bandwidth, and this is an aid to classification.
The above effects constrain frequency selection and thus, the choice of frequencies should be matched to these continually changing conditions. It may be necessary to sense the environment in real time to permit this matching.
CHAPTER III
GENERATING A DATA BASE

3.1 INTRODUCTION

In order to build a catalog of reference ship responses, a large amount of experimental data collection and processing using scaled model ships is necessary.

Firstly, the phase and amplitude returns of each model ship are measured at all frequencies, polarizations, aspect angles and elevation angles of interest. The raw data are calibrated to remove unwanted background and system response effects, and are converted into absolute radar cross section magnitude and phase. The calibrated data are then scaled in magnitude and frequency so that the responses are representative of real ship returns measured in the HF band.

The following discussion on calibration techniques is detailed for two reasons. First the calibration process is the most critical and intricate step in producing meaningful scaled cross section data. Second, this is the first time that a groundplane has been used in the measurement of ship radar cross section data.
The entire calibration process, including the development of software, experimentation and processing of data, represents approximately 40% of the total work done on this report.

3.2 MEASUREMENT OF DATA

The backscatter data of ships, calibration targets and backgrounds were measured using the compact radar range, described by Walton and Young, [18], of the ElectroScience Laboratory at the Ohio State University. The term 'backscatter' is used to describe radiation which is scattered by the target back in the direction of the transmitting antenna. A 12' x 12' parabolic reflector generates a plane wave and thus simulates an antenna at a much larger distance (hence the compact feature). The radar transmissions are of the continuous wave (CW) type, possessing no modulation. A more detailed account of the system, including system specifications, is given by Kimball [2]. Figure 3.1 summarizes the key features of the compact radar range by means of a schematic diagram.

Measurements of the amplitude and phase of the returns were made monostatically (i.e., transmitting and receiving horns were at the same location and orientation). The frequency band used was 2-18 GHz (continuous), with 10 MHz frequency increments. The continuous band avoids the need for taking measurements in sub-bands, a common practice in the past, and simplifies the subsequent calibration procedure.
Figure 3.1. Schematic diagram of the compact range system.
All of the data were taken prior to the modifications made on the edge of the reflector dish during August 1984, which were designed to reduce edge diffraction terms.

Amplitude data were recorded in units of dB cm\(^2\), and phase data were registered in degrees. Both were recorded sequentially onto a PDP-11/23 floppy disk (i.e., amplitude at \(f_1\), phase at \(f_1\), amplitude at \(f_2\), phase at \(f_2\), etc). The PDP-11 computer served mainly to control the system and observe the response as the measurements progressed; it was used for only limited processing at the time the data were taken.

Table 3.1 shows a summary of the target measurements. These sum to 366 data files; however the total files including backgrounds and calibration targets number 649. The average time for the collection of a single file, including setting up the target and taking the measurements is estimated to be 15 minutes. Thus, at least 162 hours, not including re-runs, were spent in acquiring the raw (measured) amplitude and phase returns. These data then needed to be calibrated and scaled before they were used in classification experiments.

Aspect angle was varied by rotating the ship about an axis perpendicular to the centre of the groundplane. Elevation angle was set by tilting the entire groundplane to the desired angle. Three polarization schemes were employed:

1. Transmit vertical, receive vertical - vertical (V) polarization
2. Transmit horizontal, receive horizontal - horizontal (H) polarization
3. Transmit vertical, receive horizontal - cross (X) polarization
TABLE 3.1
SUMMARY OF TARGET SPECIFICATIONS

<table>
<thead>
<tr>
<th>Aspect (degrees)</th>
<th>Elevation = 15°</th>
<th>Elevation = 27°</th>
</tr>
</thead>
<tbody>
<tr>
<td>V</td>
<td>H</td>
<td>X</td>
</tr>
<tr>
<td>0</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>10</td>
<td>10</td>
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<td>170</td>
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<td>180</td>
<td>180</td>
<td>180</td>
</tr>
</tbody>
</table>
A polarization was selected by rotating the horns of the transmitting and receiving antennas appropriately.

Figure 3.2 shows silhouettes of the six ships used in classification. These models were made from die-cast aluminum. A ship was attached to the groundplane by means of a highly conductive silver paste.

3.3 CALIBRATING THE DATA

The purpose of calibrating frequency data is to remove the imbedded system characteristic. Kimball [2] describes the process in detail, and it is summarized as follows.

1. Remove the range delay from the measured data.
2. Subtract backgrounds from target and calibration target.
3. Remove invalid points from resulting subtractions.
4. Filter (in range) the subtracted files to further reduce background terms.
5. Calibrate the target of interest according to equation 3.1.
6. Filter (in range) the calibrated data again if necessary.

The calibration equation is given by

\[
\hat{T}_C = \frac{\hat{E} (\hat{T} - \hat{B}_T)}{(\hat{S} - \hat{B}_S)} \tag{3.1}
\]

where \(\hat{T}_C, \hat{E}, \hat{T}, \hat{B}_T, \hat{S}, \) and \(\hat{B}_S\) are complex phasors for each frequency defined as;
Figure 3.2. Silhouettes of the six ships used in measurements and classifications, shown for a scale of 1:1700 (from [23]).
\( \hat{T}_C \), the signal voltage measured with the calibration target installed.

\( \hat{E} \), the computed (exact) backscatter \( \sigma \): in units of meters and absolute phase (deg.), from the calibration target.

\( \hat{T} \), the signal voltage measured with the target installed.

\( \hat{B}_T \), the signal voltage measured for the background (no target installed) associated with the target.

\( \hat{S} \), the signal voltage measured with the calibration target installed.

\( \hat{B}_S \), the signal voltage measured for the background (no target installed) associated with the calibration target.

3.3-1 THE GROUNDPLANE

For this particular set of ship measurements, a large flat, circular groundplane (Figure 3.3) was used to simulate the surface of the sea. This approach is valid because the \( \sigma/\omega_e \) ratio for salt water at frequencies in the HF band is about 1000, indicating that it is a good conductor [21]. Doppler processing is normally used to separate the ship return from the ocean wave spectrum [11]. This allows, as a first-order approximation, the sea to be represented as a flat surface in measurements. (A discussion on scattering from rough surfaces is given by [19]).
Figure 3.3. The groundplane and its low profile supporting structure.
Chen [1], in his analysis of ship classification at zero degrees elevation angle was able to simulate the effect of the surface of the sea by attaching a mirror image of the ship directly beneath the actual target. This technique is only applicable at zero degrees elevation and for vertical polarization, thus for the elevation angles of 15° and 27°, a groundplane must be used.

The groundplane has a diameter of about 3.4'. Walton and Young [5] found that the transient response of a scatterer typically dies out after $T = 6\tau$ where $\tau = L/C$, $C$ is the speed of light and $L$ is the maximum target dimension. With a maximum ship length of six inches, the impulse response would be zero after about 3 ns (one-way). The edge of the groundplane is about 1.7n ns from $T = 0$ meaning that, using this rule of thumb, the transient and groundplane edge responses might interfere. However, in the case of the 'low Q' ship targets considered here, the transient response can be shown to be indistinguishable from clutter after about 1 ns (one way) for the six inch targets (see Figure 3.4). Consequently, the ship response does not interfere with the groundplane edge.

The term $\tilde{T}$ in Equation 3.1 is a combination of backscatter from the ship and groundplane, and the term $\tilde{B}$ is the backscatter from the groundplane alone. Hence, a background (groundplane) subtraction should yield the ship backscatter. Unfortunately, owing to target-groundplane edge interactions and the unavoidable positional disturbance of the groundplane when the targets were installed or moved, some residual response remains at the location of the groundplane edges after the
subtraction and calibration. Sometimes this residual can have a peak magnitude similar to that of the ship (see Figure 3.4). Time domain windowing is necessary to remove these residuals as they represent a distortion of the desired ship data, and would be particularly disruptive in the time domain algorithm discussed later. Although smoothing, that is, convolution in the frequency domain with a Hanning window, (see Kimball [2]), which is equivalent to time domain windowing, is performed in the scaling routine, it is desirable to remove residuals before this stage is reached. The main reason for this is that existing software formatted the scaled data in such a way as to be incompatible with available fast Fourier transform programs. Hence, if the data was smoothed in the scaling routine to remove residuals, there would be no way (without extensive alterations to software) of performing an inverse Fourier transform to check the result.

Initially, it was thought that the residuals could be removed by using a sufficiently narrow window (or large number of smoothing points) in step 4 of the calibration process described earlier. To see how this would work, consider the impulse response of a typical target-on-groundplane shown in Figure 3.5. The purpose of the Hanning window outlined over the target area is to window out the effect of the antenna coupling and reflector terms. If made sufficiently narrow, the window might also remove the groundplane residuals. This was tried by setting the first null points of the window to ±3 ns; the points in time corresponding to the groundplane residuals. The residuals were removed successfully, but another problem was spawned in doing so, owing to receiver hardware problems.
Figure 3.4. An example of how the impulse response of a ship retains groundplane edge residuals (at ± 3 ns.) after calibration. The main response of the ship is constrained within ± 0.5 ns, and decays into clutter after 1.5 ns. Vertical polarization was used.
Figure 3.5. The impulse response, 2-18 GHz, of an uncalibrated target on groundplane after background subtraction, showing the location in time of all major scatterers and the pre-calibration Hanning window. The response from -50 ns to 50 ns is the continuation of the response from 0 ns to 50 ns.
The receiver used in the measurement process had band edges located at 2, 4, 8, 12.4 and 18 GHz. These band edges caused a step type pattern in the phase response (shown in Figure 3.6 for a raw 6" sphere data file with range delay removed). Narrow windowing, or convolution over a large number of frequencies (as the smoothing is actually performed), emphasized these steps differently for both the calibration target and the target of interest. The net result after calibration was a series of spikes of up to 10 dBm$^2$ in magnitude, located at or near the receiver band edge frequencies. Clearly then, narrow pre-calibration smoothing was not practical.

The edge residuals must then be removed in step 6 of the calibration process. The existing calibration software, (CAL 53 due to Kimball) achieved this by convolution in the frequency domain. This program was upgraded to permit the windowing to be done in the time domain, for two reasons.

1. The process is computationally more efficient.

2. The process is conceptually more relevant to the problem, i.e., the groundplane edges are recognizable as time domain phenomena, so it is logical to deal with them in the time domain.

Computational efficiency is a function of the number of frequencies, or data points. A fast Fourier transform requires $N \log_2 N$ operations, while a convolution requires $NM$ operations, where $N$ is the total number of data points and $M$ is the size of the convolution window (in data points). However, for time domain windowing, both forward and
Figure 3.6. An uncalibrated six-inch sphere with background and path delay removed showing the occurrence of phase steps at the 4, 8, and 12.4 GHz receive band edges.
inverse transforms as well as a multiplication over the number of data points in the time domain array (4096 in this case), which represents the actual windowing, are necessary. The total number of operations for the time and frequency domain smoothing routines are given by

\[ N_t = 2N \log_2 N + 4096 \]  
\[ N_f = NM \]  

(3.2)  
(3.3)

A total of 1601 data points (sample frequencies) were used, with a ± 3 ns window in the time domain corresponding to a 65 point convolution window. Substituting these values in 3.2 and 3.3 yields 38180 operations required for time domain windowing and 104065 operations for frequency convolution. Setting the first null points of the Hanning window to ± 3 ns removed the residuals completely (see Figure 3.8). For the largest target, most of the response was confined to ± 0.5 ns with the transient decaying into clutter at 1.5 ns (1.5 ns is the 6 dB down point of the Hanning window). However, close examination of a six inch ship response with the narrow ± 3 ns window revealed no discernable difference from the same response processed with a much wider ± 20 ns window.

There is a consequence of using the time domain technique which should be noted. Figure 3.9 shows the resultant frequency response after a perfectly flat 2-18 GHz frequency response (phase=0°) is inverse Fourier transformed, windowed with a ± 3 ns window, and then forward transformed. The first and last 0.3 GHz of the band are unuseable owing to the drop off caused by a narrow time window. This problem is somewhat precluded with the use of frequency domain convolution since
Figure 3.7. Calibrated ship data using a narrow ± 3ns pre-calibration window to remove groundplane edge residuals, resulting in the occurrence of spikes at the receiver band edges.
Figure 3.8. Use of a Hanning window after calibration to remove the ground plane edge residuals at ± 3 ns.
the frequencies near the band edges are reflected to permit the convolution, showing convolution and time domain gating, as implemented in our software, not to be exact transform pairs. However, the validity of data taken immediately adjacent to band edges is not certain, so it is good policy to ignore these regions irrespective of the technique used to perform the smoothing or windowing.

3.3-2 CHECKING CALIBRATED RESPONSES

Once a particular ship had been calibrated, its frequency and impulse response were then generated as plots. The frequency responses were examined for 'glitches', i.e., large spikes of about 10-100 MHz bandwidth and 10 to 30 dB in extent, caused by receiver hardware problems. Generally a glitch is hard to deal with because the phase and amplitude responses affect up to 10 points. If the glitch exists in a background or calibration target file then an alternative data file might be used. A bad glitch in a ship data file might require a new set of measurements. Small glitches, both in amplitude and bandwidth, at frequencies lower than 4 GHz can be tolerated.

The time response was the main tool for checking the validity of calibrations because the transient response gives an intuitive geometric guide to the mechanisms which cause scattering. Figure 3.9 shows a typical response, scaled 1 to 1 with the ship overlaid. Using templates in this way shows whether the main response confines itself to the length of the ship, and if structures likely to cause large amounts
Figure 3.9. Resultant frequency response when a flat 2 to 18 GHz frequency response is windowed in the time domain with a ± 3 ns Hanning window.
of scattering are indeed doing so. The bandwidth of the responses is sufficiently large to provide the necessary resolution to make these judgements. Resolution in time is given by

\[ \tau = \frac{1}{B} \]  

(3.4)

For \( B = 16 \) GHz, then \( \tau = 62.5 \) p seconds, which at the speed of light corresponds to 1.875 cm. Most of the major features on the ship models, such as masts, gun turrets, etc., are separated by about 1 to 2 cm. There was not always a 1:1 relationship between the physical features of a ship and its time response; partly owing to the resolution limit and partly because of the complex scattering interactions and interferences. Certain aspect angles, particularly those close to 90° were harder to judge than others. No standard procedure could be developed under which to judge a calibrated result, although the presence of significant precursors or other unusual phenomena would tend to indicate a bad calibration.

An alternative method for judging the validity of the calibration, based on comparing a measured hemisphere-on-a-groundplane with its theoretical counterpart, was found to be unsatisfactory. The measured and theoretical responses were too disimilar to be meaningful. Much work was done in the pursuit of finding the mechanism responsible for the differences. However, no conclusive evidence was found and this particular technique was dropped in favour of the time domain method.
Figure 3.10. A ship scaled 1:1 with its impulse response in order to check a calibration.
3.4 DATA SCALING

Data collected and calibrated in the microwave region must be scaled before being used in classification algorithms. A data point is scaled in two ways. Firstly, its amplitude is multiplied by the scale factor, and secondly, the frequency it represents is divided by the scale factor. Data measured in the 2 to 18 GHz microwave band was collected for models having scale factors of 1:1200 and 1:2400. The resulting HF bands are shown in Table 3.2.

<table>
<thead>
<tr>
<th>Scale Factor</th>
<th>1:1200</th>
<th>1:2400</th>
</tr>
</thead>
<tbody>
<tr>
<td>$f_{\text{min}}$</td>
<td>1.667 MHz</td>
<td>0.833 MHz</td>
</tr>
<tr>
<td>$f_{\text{max}}$</td>
<td>15 MHz</td>
<td>7.5 MHz</td>
</tr>
</tbody>
</table>
The figures presented above were rounded to one decimal place, and a frequency increment was selected. Since frequencies may have been selected which were not actually represented by data points, it was necessary to interpolate between data points by means of a Hamming window. Chen [1] also used a Hamming window to remove noise and clutter from the calibrated data. As explained earlier, it was necessary to perform smoothing during the calibration process.

A frequency increment of 1/32 MHz was selected for each band. Window widths of 0.5 MHz and 0.25 MHz were selected for the 1.7 to 15 MHz band and 0.8 to 7.5 MHz band respectively. The shape of the scaled data curves differed little from the calibrated data curves. This is to be expected as the window to bandwidth ratios for each are 1:27 and 1:25 respectively. Representative scaled frequency and phase returns are shown in Appendix B.

The usable bandwidth for classification must overlap both scaled bands; this is 1.7 to 7.5 MHz. For reasons mentioned in the discussion of the calibration procedure, frequencies near the band edges are avoided, so the net usable bandwidth is about 2 to 7 MHz.
A SUMMARY OF SCALED DATA GENERATION

1. Measured Data (m)

Amplitudes

\[ A_m \]

Phases

\[ \theta_m \]

(at) frequencies

\[ f_m : 2 - 18 \text{ GHz}, 10 \text{ MHz steps} \equiv 1601 \text{ data points} \]

2. Calibrated Data (c)

Amplitudes

\[ A_c \]

Phases

\[ \theta_c \]

(at) frequencies

\[ f_c : 2 - 18 \text{ GHz}, 10 \text{ MHz steps} \]

Smoothing

\[ \pm 3 \text{ ns (first nulls)} \] Hanning window equivalent to 65 point smoothing, gives a 1:25 window to bandwidth ratio.

3. Scaled Data (s)

Amplitudes

\[ A_s = A_c \text{ SF} \]

Phases

\[ \theta_s = \theta_c \]

(at) frequencies

\[ f_s = f_s / \text{SF} \]

SF = 1:1200

1.7 - 15 MHz, 1/32 MHz steps
426 data points
Interpolation using 0.5 MHz Hamming window

SF = 1:2400

0.8 - 7.5 MHz, 1/32 MHz steps
214 data points
Interpolation using 0.25 MHz Hamming window

Net useable common bandwidth : 2 - 7 MHz.

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CHAPTER IV

TARGET CLASSIFICATION

4.1 INTRODUCTION

The basic process of target classification is illustrated in Figure 4.1. The measurement system in this case is a compact radar range. It produces a measurement vector m, which is a set of amplitudes and phases at a number of frequencies, aspect angles, elevation angles, and polarizations. Each measurement vector is a point in M-dimensional space (also called the observation space). The feature extractor reduces the dimensionality of the measurement vector to produce a feature vector n, which is a point in N-dimensional space (M>N). For example, if the amplitudes and phase returns of a ship were measured at 2 polarizations, 2 elevation angles, 3 aspect angles and 4 frequencies, and it is assumed polarization and elevation are known and only amplitudes are used to classify the target, then the feature extractor reduces a 96-dimensional space to 12-dimensional space (4 frequencies x 3 aspect angles). Essentially, the feature extractor is
Figure 4.1. Basic process of a pattern classification system.

...the mechanism by which a particular data file is addressed, since a measurement vector is usually dispersed amongst several data files.

The feature vector is then passed to the object classifier which uses a particular algorithm to make a decision as to the identity of the object. The decision can be either, 1) the object is classified as a known member of a catalog or, 2) the object belongs to some other 'uncatalogued' class. The classification of catalogued and uncatalogued classes is discussed by Lin [6]. For this particular application, only catalogued classification is considered.

The ultimate goal of a classification system is to identify targets with a minimum probability of error. Since the noise in these tests was simulated, the classification could have been a parametric procedure. However, in practice, the exact probability distributions of features in
the feature space after contamination by noise are not known, thus we must resort to non-parametric methods of classification. Two such methods which do not require knowledge of probabilistic information are the nearest neighbour algorithm [1] and the time domain correlation algorithm [1], and are discussed below.

4.2 CLASSIFICATION IN THE FREQUENCY DOMAIN

The nearest neighbour (NN) algorithm uses amplitude and phase returns measured at a series of frequencies \( f = f_1, f_2, \ldots, f_N \). (Note that, in general, the NN algorithm is applicable to both the frequency and the time domain.) The amplitude is defined as the square root of the measured target cross section. For sky-wave radars, it is difficult to recover the intrinsic phase of a target, because of the dispersive and variable nature of the ionosphere and because the wave path lengths are difficult to estimate accurately. Hence a measurement of both the actual (intrinsic) phase, or relative phase may contain unacceptable errors. To circumvent these difficulties, the differential quantity \( W \) is used and is defined as:

\[
W_i = \theta_i \lambda_i - \theta_{i+1} \lambda_{i+1}
\]  

(4.1)

where

- \( \theta_i \) is the measured phase at \( \lambda = \lambda_i \),
- \( R_i \) is the wave path length at \( \lambda = \lambda_i \), and
- \( i = 1, \ldots, N - 1 \) (where \( N \) = Number of frequencies).
Equation (4.1) assumes that $R_i$ and $R_{i+1}$ differ by a small amount. This is true for sky-wave radars only when $\lambda_i$ and $\lambda_{i+1}$ differ by a small amount and the radar operating frequencies are chosen carefully.

A feature space for a given orientation and polarization can be defined as,

$$(A_1, A_2, \ldots, A_N, KW_1, KW_2, \ldots, KW_{N-1})$$

where

$N$ is the number of frequencies.

$K$ is a normalization constant relating the variances of $A$ and $W$.

$$K = \sqrt{\frac{\text{VAR}(A)}{\beta \text{VAR}(W)}}$$ (4.2)

$$\beta = \frac{\text{VAR}(A)}{\text{VAR}(KW)}$$ (4.3)

The parameter $\beta$ is a variable ranging from 0 to $\infty$ and its value depends on the reliability of amplitude and phase information.

$\beta = 0$ corresponds to using amplitudes only
$\beta = \infty$ corresponds to using differential phase only
$\beta = 1$ corresponds to the use of amplitudes and phases with equal weighting in the respective variances.

In practical use of the algorithm, the stated limits of $\beta$ (i.e., 0 and $\infty$) are not substituted into Equation 4.2 in order to select either the amplitudes-only feature or differential-phases-only feature from the
feature space; these features are selected directly and are independent of K. It is worth mentioning that a value of $\beta = 0.01$ gives classification results virtually indistinguishable from the differential-phases-only feature, and $\beta = 100$ does the same for the amplitudes-only feature. K is calculated only when both amplitudes and phases are used, and for all of the experiments done here, $\beta$ was set to 1.

Let $A_t(f_i)$, $\theta_t(f_i)$ be the target amplitude and phase returns, where $i = 1, 2, \ldots, N$. Let $A_j(f_i)$, $\theta_j(f_i)$ be the data base amplitude and phase returns, where $j$ is the target index, $j = 1, 2, \ldots, M$. The NN algorithm is as follows:

1. Compute the differential phases

$$W_{it} = \lambda_t \theta_{t}(f_i) - \lambda_{t+1} \theta_{t}(f_{i+1})$$

$$W_{ij} = \lambda_i \theta_{j}(f_i) - \lambda_{i+1} \theta_{j}(f_{i+1})$$

$i = 1, 2, \ldots, N-1$

$j = 1, 2, \ldots, M$

2. Calculate the sample averages of $A$ and $W$ in the data base.

$$\text{Avg}(A) = \frac{1}{MN} \sum_{j=1}^{M} \sum_{i=1}^{N} A_j(f_i) = \bar{A}$$

$$\text{Avg}(W) = \frac{1}{M(N-1)} \sum_{j=1}^{M} \sum_{i=1}^{N-1} W_{ij} = \bar{W}$$
3. Calculate the sample variances of A and W in the database.

\[ \text{Var}(A) = \frac{1}{MN} \sum_{j=1}^{M} \sum_{i=1}^{N} (A_j(f_i) - \bar{A})^2 \]

\[ \text{Var}(W) = \frac{1}{M(N-1)} \sum_{j=1}^{M} \sum_{i=1}^{N-1} (W_i(j) - \bar{W})^2 \]

4. Select \( \beta \) and calculate \( K \).

\[ K = \frac{\text{VAR}(A)}{\sqrt{\beta \text{VAR}(W)}} \]

5. Compute the distance between the target and each class in the database.

\[ d_{t,j} = \sqrt{\sum_{i=1}^{N} (A_t(f_i) - A_j(f_i))^2 + \sum_{i=1}^{N-1} (KW_i(t) - KW_i(j))^2} \]

\( j = 1, 2, ..., M \).

6. Apply the nearest neighbour rule:

Choose smallest \( d_{t,j} \) \( j = 1, 2, ..., M \).

If \( d_{t,m} = \min(d_{t,j}) \) classify the target (t) as m.

4.3 CLASSIFICATION IN THE TIME DOMAIN

The inverse Fourier transform of the frequency and phase responses mentioned in the previous section yields the impulse response of the target. This is the time-dependent field intensity produced when a plane electromagnetic wave washes over the object.
In the frequency domain, differential phase was used to reduce the phase error introduced when the path length varies. This error manifests itself as a time shift in the canonical time domain, and is removable by means of a correlation process. (Note that the correlation process is applicable to both the time and the frequency domain). Chen [1] discussed the correlation process; a summary of its implementation is given below.

1. Compute the correlation function.

\[ \rho_{t,r}(k) = \frac{\text{DIFT} \left[ X(m) Y^*(m) \right]}{2 \sqrt{\sum_{m=1}^{M} |X(m)|^2} \sqrt{\sum_{m=1}^{M} |Y(m)|^2}} \]

where

\[ X(m) = A_t(f_k)e^{i\theta_t(f_k)} \quad t = \text{target index} \]
\[ Y(m) = A_r(f_k)e^{i\theta_r(f_k)} \quad r = \text{catalog index} \]
\[ k = 1, 2, \ldots, M \]
\[ M = \text{Number of frequencies} \]
\[ m = f_1/\Delta f, f_2/\Delta f, \ldots, f_M/\Delta f \]
\[ \Delta f = f_i - f_{i-1} \quad i = 1, 2, \ldots, M \]

and DIFT is the Discrete Inverse Fourier Transform of the frequency domain data \(X(m) Y^*(m))\). \(Y^*\) is the complex conjugate of \(Y\).

2. Choose the time shift constant \(k\) such that \(\rho(k)\) is maximized. \(\rho_{t,r}(k)\)

For a particular set of two targets this yields \(\rho_{t,r}(k)\), \(r = 1, \ldots, N\), where \(N = \text{number of catalog members}\).
3. Choose the largest $\max (\rho_t, r)$. If $\max (\rho_t, q) = \max (\rho_t, r)$, classify the target $(t)$ as $q$.

The DIFT will work with as few as 2 frequencies, however, this low number of frequency samples is insufficient to provide any useable resolution. Recall that the cross section data, as originally measured, were sampled at 1601 frequencies from 2 to 18 GHz with a 10 MHz interval. This amount of sampling provided the necessary resolution to distinguish between individual scattering centers on the target (which allowed the calibrations to be checked). A much lower number of samples still provides scattering information of the target, but relates more to the overall dimensions rather than individual structures.

4.4 RELATIVE AMPLITUDE FEATURE

The relative amplitude feature was developed to reduce the effects of a possible multiplicative bias in the amplitude data. It is defined as follows:

$$\bar{A}_i = A_i/A_{i+1}; \quad i = 1, 2, ..., N-1.$$  

Such a bias might be due to an error in estimating the sea state, which would cause a local shift of amplitudes in the frequency response. The relative amplitude feature was not designed to remove the complex multiplicative terms impressed by the ionosphere.
CHAPTER V

EXPERIMENTAL CONSIDERATIONS

5.1 INTRODUCTION

Figure 5.1 summarizes the experimental classification procedure in the form of a flow chart. Data for M ships at a total of N frequencies are contained in the data base. The data base is a directory of 366 scaled data files, each corresponding to a ship at a particular aspect angle, elevation angle and polarization. Amplitudes and phases at the desired frequencies, aspect and elevation angles, and polarizations are selected for each ship in the data base to form a catalog (this is equivalent to feature extraction, see Chapter IV, Section 4.1). This selection, in a practical situation, would be based on all of the a priori information pertaining to the unknown target. Gaussian noise is then added to the entire catalog to produce a set of test targets. Classification proceeds for each noisy test target and decision statistics are compiled. It is worth emphasizing that in the following experiments, the whole catalog of ship amplitude and phase returns is corrupted by noise and classification proceeds for each ship target in
Figure 5.1. A flow chart of the experimental process of classification.
that 'noisy catalog'. At this point there are two catalogs; one a noise-contaminated version of the other. The returns of the first noisy ship are compared (by some algorithm) with those of the M noise-free ships, and a decision is made as to the identity of the test target. Since we added the noise to the test target, we know its identity a priori, and can therefore determine if the classification was either correct or incorrect. We say a target has been identified correctly if we choose the right ship at the right aspect and elevation angle. A target is misclassified if we choose either the wrong ship, or the right ship at the wrong elevation or aspect angle. This decision process is justified on the basis that we are investigating the classification properties of parameters, such as aspect angle, polarization etc., rather than the classification properties of individual ships.

The process is then repeated for subsequent members of the noisy catalog until all M noise-contaminated test targets have been classified. The number of misclassifications is recorded for the particular level of injected noise power. Hence the statistics of the experiments apply to the collection of ships as a whole and classification properties between individual ships, although of interest, are not investigated here.

The experiment is repeated a number of times in order to compile meaningful statistics, or representative curves (see below). The entire process is then repeated for a different injected noise power so that curves of misclassification percentage versus post-processing SNR can be drawn. The software needed to perform this process is listed with comments in Appendix C.
5.2 CHOICE OF EXPERIMENTAL FREQUENCIES

The discussion in Chapter III concluded with a summary of the data-base generation procedure, which showed that our data has a usable band of 2-7 MHz. An important concern here is that the classification frequencies should be in the lower part of the resonance region, i.e., close to \( L/\lambda = 1 \), where \( L \) is the maximum dimension of the dominant scattering structures of the target. The increment between sample frequencies is also an important parameter.

The experimental time domain ramp response in our ship data base, as previously mentioned, has a duration of less than 6 transit times across a target of length \( L \), i.e.,

\[
R(t, t_0) = 0; \quad t_0 > t > t_0 + 6L/C
\]

where \( C \) is the speed of light and \( t_0 \) is a time reference. (This is a worst-case rule of thumb; there are many exceptions, including the case at hand.) For the preceding time limited expression, Shannon's sampling theorem [22] requires that the frequency sampling interval should satisfy

\[
\Delta f < C/6L . \tag{5.1}
\]

The ships dimensions are in the order of 100 m, implying

\[
\Delta f < 0.5 \text{ MHz}.
\]

Thus, \( \Delta f = 0.4 \text{ MHz} \) was selected as the frequency increment. The corresponding selection of frequencies is given in Table 5.1.
TABLE 5.1

RANGE OF FREQUENCIES USED IN EXPERIMENTS

<table>
<thead>
<tr>
<th>N</th>
<th>Frequency Range</th>
</tr>
</thead>
<tbody>
<tr>
<td>2</td>
<td>2.0 MHz &lt; f &lt; 2.4 MHz</td>
</tr>
<tr>
<td>4</td>
<td>2.0 MHz &lt; f &lt; 3.2 MHz</td>
</tr>
<tr>
<td>8</td>
<td>2.0 MHz &lt; f &lt; 4.8 MHz</td>
</tr>
<tr>
<td>12</td>
<td>2.0 MHz &lt; f &lt; 6.4 MHz</td>
</tr>
</tbody>
</table>

Note that 2.0 MHz corresponds to a wavelength of 150 m, and 4.8 MHz (N = 8 frequencies was the most commonly used number) corresponds to a wavelength of 62.5 m. Hence, with \(0.7 < L/\lambda < 1.6\), the lower resonance region criterion is satisfied.

5.3 SELECTION OF \(\beta\) IN NEAREST NEIGHBOUR ALGORITHM

As discussed earlier, the selection of \(\beta\) is relevant only for the case where both amplitudes and (differential) phases are used in the nearest neighbour algorithm. \(\beta = 1\) corresponds to using amplitudes and phases with equal weighting in variance. Classification with \(\beta\) at some intermediate value represents a weighting towards either feature depending upon the reliability of that feature's measurement. Chen [1] acknowledged that his estimate for \(K\) was not optimal.

\[
K = \sqrt{\frac{\text{VAR}(A)}{\beta \text{VAR}(W)}} \quad .
\]  

(5.2)
An optimal expression for $K$, in addition to a carefully selected value of $B$ might provide an optimal use of the AW feature. This possibility is not explored in this work, but is discussed in the conclusion.

5.4 NOISE MODEL

Before discussing the specific characteristics of the noise model, it is appropriate to define exactly what is meant by noise and signal-to-noise ratio.

Figure 5.2 summarizes the various channel and processing stages in a radar system. Assume that, immediately after the incident radar wave impinges on a target, a radar cross section $\sigma_t$ is generated, where $\sigma_t$ comprises the ship return $\sigma_s$ and sea echo $\sigma_r$. If it were possible to be exactly adjacent to the target at this point in time, we could measure, with a hypothetically perfect system, the exact values of $\sigma_s$ and $\sigma_r$. Knowing the exact state of the sea would allow us to compute $\sigma_{\text{exact}}$ precisely from a perfect theoretical model of the sea scattering process. The (exact) radar cross section of the ship would then be given by

$$\sigma_{\text{ship}} = \frac{\sigma_s}{\sigma_r} \sigma_{\text{exact}} \quad (5.3)$$

This, of course, is not the case and the initial radar cross section $\sigma_t$ is corrupted by the ionosphere and various stages of signal processing. After doppler filtering, the resultant radar cross section $\hat{\sigma}_t$ is separated into its components $\hat{\sigma}_s$ and $\hat{\sigma}_r$. Using an imperfect estimate
\[ \sigma_t = \text{Original target radar cross section comprising } \sigma_f \text{ and } \sigma_s \]
\[ \sigma_f = \text{Sea echo radar cross section} \]
\[ \sigma_s = \text{Ship radar cross section} \]
\[ \hat{\sigma}_{\text{exact}} = \text{Estimate of theoretical exact sea echo} \]
\[ \hat{\sigma}_{\text{ship}} = \text{Post-processed radar cross section of ship.} \]

Figure 5.2. Schematic diagram summarizing the various stages of noise contamination after the initial scattering of radar energy by a ship.
of the exact theoretical sea-echo cross section \( \sigma_{\text{exact}} \), the resultant calibration yields

\[
\hat{\sigma}_{\text{ship}} = \frac{\hat{\sigma}_{s}}{\hat{\sigma}_{r}} \sigma_{\text{exact}} \quad .
\] (5.4)

Assuming that multiplicative noise terms can be ignored (for the sake of conceptualizing), Equations (5.3) and (5.4) are related by

\[
\hat{\sigma}_{\text{ship}} = \sigma_{\text{ship}} + \varepsilon \quad .
\] (5.5)

Here \( \varepsilon \) represents the remaining errors after the processing system has done its best to minimize the effects of various 'noises' accumulated after the initial scattering of energy from the target. The term \( \varepsilon \) is defined as post-processing noise and consequently, in the context of processed radar returns, the term post-processing signal-to-noise ratio is used.

Pre-processing noise is noise present in the returned signal before any processing, such as filtering, is done to reduce the effect of this contamination. For example, a lightning strike might add 10 dBm\(^2\) of pre-processed noise to a scattered radar signal, and this would eventually become a component, to some degree, of post-processed noise \( \varepsilon \).

Chen [1] concluded that the total noise power (i.e., the error variance of the final estimation of radar cross section \( \sigma \)) rather than the specific noise characteristics, has the most impact on the performance of classification algorithms. He went on to say that, because there are
numerous independent noise sources, none of which dominate, the Central Limit Theorem can be applied and the sum of these noises can be approximately described as Gaussian.

Headrick and Skolnik [9] commented that the effect of external (pre-processed) noise sources such as lightning, man-made noise and other HF transmissions, can be significantly greater than that of the internal (pre-processed) Gaussian receiver noise. The ionosphere contributes to signal distortion by introducing complex multiplicative terms by dispersion and by polarization rotation. These errors will appear, to a certain degree, in the final post-processed estimate of the radar cross section. A complete, representative model for post-processed sky-wave path distortion is clearly needed, however, a sufficiently detailed model has not yet been developed. In view of this, and based on Chen's initial conclusion, the Gaussian model was used.

Figure 5.3 shows a noise-free vector which is contaminated by adding in-phase and quadrature Gaussian noise components. The resultant vector has real and imaginary terms thus

\[ \tilde{I} = A \cos \theta + \tilde{n}_I \]  
\[ \tilde{Q} = A \sin \theta + \tilde{n}_Q \]  

(5.6)  
(5.7)

where \( \tilde{n}_I \) and \( \tilde{n}_Q \) are independent Gaussian-distributed random variables with zero mean and variance \( \sigma^2 \). The power in component \( \tilde{n}_I \) is equal to that of \( \tilde{n}_Q \). The noise amplitude is given as
Figure 5.3. The distribution of the noise on an I-Q plane (from [1]).
\[ A_n = \sqrt{n_i^2 + n_q^2} \]  

and is Rayleigh distributed [15]. The noise phase is given as

\[ \theta_n = \tan^{-1}\left( \frac{n_Q}{\bar{n}_Q} \right) \]  

and is uniformly distributed [15]. An expression for signal to noise ratio is given as

\[ \frac{S}{N} = \frac{I^2 + Q^2}{\text{VAR}(\bar{n}_I) + \text{VAR}(\bar{n}_Q)} = \frac{A^2}{\alpha^2} \]  

where

\[ \alpha^2 \text{ is the noise power, and} \]

\[ A^2 \text{ is the average signal power estimated as} \]

\[ A^2 = \sum_{i=1}^{\text{NF}} \sum_{j=1}^{\text{NS}} A_{ij}^2 \]  

where

\[ A_{ij} \text{ is the amplitude for the } i\text{th frequency and the } j\text{th ship.} \]

\[ i \text{ is a frequency index, } i = 1, 2, \ldots, \text{NF} \text{ = NUMBER OF FREQUENCIES,} \]

and,

\[ j \text{ is a target index, } j = 1, 2, \ldots, \text{NS} \text{ = NUMBER OF SHIPS.} \]
5.5 POST-PROCESSING SIGNAL-TO-NOISE RATIO

From the previous discussion, we define post-processing signal-to-noise ratio (post-processing SNR) as the ratio of signal power to error variance after the received waveform has been processed to produce a final best estimate of the amplitude or phase, or both, of the target radar cross section.

From [1] post-processing SNR can be estimated as

\[
\frac{S}{N} = \sqrt{\frac{M}{M+2}}
\]  

(5.12)

where \( M \) is the number of samples for a particular frequency. Using a typical sample time of 10 seconds and taking \( M=100 \) samples, (a total observation time of 16 minutes) yields a post-processing SNR of

\[
\frac{S}{N} = \sqrt{\frac{100}{102}} = 10 = 10 \text{ dB}.
\]

This value is used as a reference point in the experiments discussed below.

Post-processing SNR is limited by the total time available to take the \( M \) samples and by the duration of a sample. Assuming that the sea state is used as a reference, then the total time available is limited to between 1 and 12 hours; the period over which the sea state statistics are stationary [16]. The sample duration is the inverse of doppler resolution.

\[
t_s = \frac{1}{T_d}
\]

(5.13)
where \( t_s \) is the coherent sampling interval (seconds) and \( T_d \) is the doppler resolution (Hz). The upper bound on \( t_s \) is set by the maximum coherent observation time through the ionosphere. This depends on the particular ionization layer; 25 seconds is a typical maximum for F-layer \([11]\) and 100 seconds is possible for E-layer \([11]\). The lower bound, set by the required doppler resolution depends on the radial component of ships velocity, and of the sea state \([12]\). Generally 10 seconds (\( T_d=0.1 \) Hz) is a safe minimum. Assuming a 12 hour interval, the maximum attainable SNR is then

\[
\frac{S}{N} = \frac{(12\times3600/10)}{(12\times3600/10) + 2} \approx 18 \text{ dB}
\]

In view of the fact that the radar system will usually scan thousands of square miles of ocean and possibly need to identify several targets, it might not be possible to devote 12 hours to observing one target at one frequency. Consequently, the figure of 18 dB would then be an unrealistic upper limit. For a maximum observation time of 1 hour, post-processing SNR has a maximum of 13 dB.

Another important consideration concerning (post-processing) SNR is the variability of signal powers between individual frequencies, between ships and between classes of ships.

From Figure B.1 in Appendix B it is evident that signal powers of returns at individual frequencies can vary from 0 to 40 dBm\(^2\). The average signal power for this ship might be 30 dBm\(^2\), and so for a 10 dB post processing SNR, it is necessary to add 20 dBm\(^2\) of noise power to
each amplitude. This yields SNRs for individual amplitude returns ranging from -20 dB to 20 dB. Clearly the lower amplitudes are much more distorted than the larger ones, for a particular level of injected noise power. This is reasonable since for a given noise environment, weaker returns are more susceptible to degradation by noise. The same argument can be developed for ships; the individual amplitude returns of larger ships over several frequencies are, on average, larger than amplitude returns of smaller ships. And finally, we consider the case of classes. If a classification is performed using a catalog comprising ships at aspect angles 0°, 90°, and 180°, then the average signal power, used to determine the amount of noise required for a given post-processing SNR, would be based on the average power for the 3 classes. The 90° aspect angle amplitude returns are generally higher than those of 0° and 180° aspect angles, consequently we would maintain a constant level of noise power amongst the 3 classes, but SNR would vary from class to class and the resulting classification curve would not be representative of any one particular aspect angle.

In general, the signal level of a received amplitude return (not including the noise power) can be estimated with a useful degree of accuracy. This information can be used, a priori, with the estimates of aspect and elevation angles to reduce the size of the catalog. As a consequence, we are interested in the classification of ships at specific aspect and elevation angles, rather than over a group of markedly different aspect or elevation angles. Hence if we compare classification results for 90° aspect with a classification results
for 180° aspect at a given post-processing SNR, we must bear in mind that the amount of injected noise differs for each case.

The consequences of this discussion are somewhat dependent on the effect of the difference in average signal powers, (between any two given ships), on classification errors. One way to access the importance of this difference would be to compare classifications where all catalog members are normalized with respect to their own average signal powers, with classifications where the variability of average signal power is retained.

5.6 ESTIMATION OF THE PROBABILITY OF MISCLASSIFICATION

Chen [1] used the Maximum Likelihood Estimate (MLE) [14] as the estimate of probability of error for a given target;

$$PE = \hat{P}_E$$

where $PE$ is the probability of error and $\hat{P}_E$ is the MLE of the probability of error. The proximity of the terms expressed in the equation above is usually stated in terms of confidence interval. A (1-\(\alpha\)) confidence interval for $PE$ is given as

$$\hat{P}_E \pm \varepsilon_\alpha \sqrt{\frac{\hat{P}_0(1-\hat{P}_0)}{Mn}} \quad (5.14)$$
where

\[ M = \text{number of targets} \]
\[ n = \text{number of experiments} \]
\[ \hat{P}_o = \text{probability for which the confidence interval is desired} \]
\[ \xi_a \text{ is such that} \]

\[
\phi(\xi_a) = 1 - \alpha/2 = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\xi_a} e^{-x^2/2} \, dx \tag{5.15}
\]

where \( \phi \) is the standard (zero mean, unit variance) normal distribution.

For example, the calculated 90% confidence interval at 30% error (a typical error value) for 18 targets and 50 experiments is found as

\[ \alpha = 1 - 0.9 = 0.1 \]
\[ \phi(\xi_a) = 1 - \alpha/2 = 0.95 \]
\[ \xi_a = 1.65 \]
\[ P_0 = 30\% = 0.3 \]
\[ PE = \hat{P}_E \pm 1.65 \sqrt{\frac{0.3(1-0.3)}{18 \times 50}} \]
\[ PE = \hat{P}_E \pm 2.5\% \]

Clearly, the result is made more accurate by increasing the number of experiments. It must be emphasized that the confidence interval is not an expression of probability, i.e., the above calculation does not show that 9 times out of 10 the actual probability of error at 30% misclassification will be within ±2.5% (or 27.5% < PE < 32.5%). DeGroot [17] pointed out that "confidence" is a more subtle and less defined
term than "probability" and that under certain experimental conditions, the term "confidence" interval can be misleading. In regard to this, we must view confidence interval as a measure of accuracy for the measured result, not as an absolute probability.

In his experiments, Chen [1] used \( M=15 \) experiments. The significant increase to \( M=50 \) for this work is for two reasons. Firstly, the curves are more reliable, especially when close together, if the confidence interval is smaller. A \( \pm 5.6\% \) confidence interval (using \( M=15 \)) in some cases meant that two curves, representing the same experiment, using the same parameters but with different random number seeds, differed by 11\%. Using \( M=50 \) experiments cuts this worst case error by half, and makes conclusions about particular sets of curves more reliable. More importantly, the large number of classification runs required that curves be plotted automatically after each run. Hence it is desirable to have a high degree of similarity between curves from different runs representing the same classification parameters.
6.1 INTRODUCTION

This chapter presents most of the original work done in this report. The purpose of the experiments discussed here is to study classification behaviour under a wide variety of the available classification parameters. These are listed below (a list of commonly used terms is given in Table 6.1).

1. Frequency Band (in the allotted 2-7 MHz)
2. Comparisons with previous work on ships
3. Elevation Angle (known, unknown and with error)
4. Polarization
5. Aspect Zone (known, unknown)
6. Aspect Error (magnitude of error and location in aspect zone).
TABLE 6.1
DEFINITIONS

1. ASPECT ZONE: A small range of aspect angles centred on or adjacent to a particular aspect angle.

<table>
<thead>
<tr>
<th>ASPECT ZONE</th>
<th>NAME</th>
<th>ASPECT ANGLES</th>
</tr>
</thead>
<tbody>
<tr>
<td>0°</td>
<td>BOW</td>
<td>0°, 10°</td>
</tr>
<tr>
<td>90°</td>
<td>BROADSIDE</td>
<td>80°, 90°, 180°</td>
</tr>
<tr>
<td>180°</td>
<td>STERN</td>
<td>170°, 180°</td>
</tr>
</tbody>
</table>

2. KNOWN ASPECT: If ship data having more than one aspect angle is used in a classification, then the aspect angle is assumed known when the algorithm only allows comparisons between an 'unknown' noisy target at aspect $\theta_K$, with 'noise-free' catalog targets at the same aspect angle $\theta_K$. The same follows for known elevation.

3. UNKNOWN ASPECT: If ship data having more than one aspect angle is used in a classification, then the aspect angle is assumed unknown when the algorithm permits comparisons between an 'unknown' noisy target at $\theta_K$ with all 'noise-free' catalog targets at all aspect angles used in the classification.

4. ALGORITHM: One of the following:
   1. Nearest neighbour (NN), using any of the features listed in 5.
   2. Time domain (T)

5. NN ALGORITHM FEATURES:
   - A: Amplitude only
   - W: Differential phase only
   - AW: Amplitude and differential phase
   - $\overline{A}$: Relative amplitude
   - $\overline{AW}$: Relative amplitude and differential phase.
Note that, in general, the term 'feature', in the context of resonance region radar returns, applies to any property or quality associated with the returns. This includes, for example, the magnitudes of the time domain impulse response.

6. PARAMETER: A variable in the measurement or classification process such as frequency, aspect angle, elevation angle, polarization, feature or algorithm.

7. DATA BASE: A collection of data files containing a measurement vector for each ship.

8. CATALOG: A single 2-dimensional complex array containing the amplitudes and phases of all ships at the selected frequencies, and other parameters.

9. FEATURE VECTOR: A single vector for a particular target, dimensional in frequency or time, derived from the catalog by selecting a particular feature.

10. POLARIZATION: One of the following:

1. Vertical, V
2. Horizontal, H
3. Cross, X
4. Vertical divided by horizontal, V/H.

The term described in 4 is not really a polarization scheme, in the sense of the preceding three terms, but is called a polarization for convenience. Strictly speaking, the polarization of a wave describes the instantaneous orientation of the electric field vector; the terms listed above refer to a particular measurement scheme or use of radar cross section returns, with respect to polarization.
If each of these parameters were to be assessed in terms of the others, over 10,000 curves would be needed. Clearly this is not practical; however, an intelligent approach toward choosing the experiments allows a thorough investigation without incurring too much processing of data.

First, experimenting with a sub-band in the 2-7 MHz band indicates the best region of frequencies (if any) for a particular type of classification. Second, an evaluation of classification performance as a function of the number of frequencies (NF) provides a comparison with work presented in [1] and also establishes the ranking of a particular number of frequencies. Hence we can choose NF=8 for subsequent experiments and refer to this section for results pertaining to other values of NF. The problem of choosing the desired classification frequencies is thus solved. Next, a study of elevation angle and its effect on classification allows classification using only one elevation in subsequent experiments. The remaining parameters are then classified using various features and algorithms, polarizations and aspect zones. Sometimes it is useful to reduce the number of features in the NN algorithm by eliminating $W$ and $\tilde{AW}$, or $W$ and $\tilde{A}$. $W$ is generally the least useful of NN features and it is unlikely that only phase would be measured practical situation.

It must be remembered that certain parameters, such as relative amplitude ($\tilde{A}$), and V/H polarization, are not really tested in these experiments since the difficulties which they were designed to overcome were not simulated. This should be taken into consideration when making comparisons between features such as $A$ and $\tilde{A}$, or polarizations such as $V$
and V/H. For example, V/H polarization purportedly has the property that multiplicative errors cancel out by virtue of the division of a vertically polarized phasor by a horizontally polarized phasor. To test the validity of this assumption, the ship data must be contaminated with multiplicative noise before classification proceeds. For reasons mentioned earlier, multiplicative noise models were excluded from this analysis, and what we are really measuring is the 'insertion loss' of the algorithm which uses this parameter.

In the following sections, each experiment is introduced, detailing the aim of the experiments with a brief discussion of any relevent points. The results of all experiments in the form of misclassification percentage versus post-processing SNR curves, are contained in Appendix A. These results are summarized by means of histograms representing averaged misclassification percentage at 10 dB post-processing SNR, and are presented, along with typical misclassification curves.

Conclusions drawn from the histograms often compare the classification performance of one parameter with the performance of another by saying that misclassification percentage (at 10 dB post-processing SNR) is higher or lower by x%; here x is always the difference between two misclassification percentages, not the percentage increase of one misclassification percentage compared with another.
6.1.2 INTERPRETATION OF HEADERS IN CLASSIFICATION RESULTS

The data presented in subsequent sections were plotted with an automatic header system to aid the batch processing of classification experiments. The header comprises of up to 11 lines and these are described as follows:

1. LINE 1. TARGET TYPE. i.e., ships, aircraft, ground vehicles, etc.

2. LINE 2. POLARIZATION. This can be either vertical ('V'), horizontal ('H'), cross ('X') or vertical divided by horizontal ('V/H'). The polarization for each curve is printed if this varies from curve to curve.

3. LINE 3. A PRIORI KNOWLEDGE OF ELEVATION ANGLE. This can be either 'known', 'unknown', or 'known/unknown' if some of the curves use known elevation angle and others use unknown elevation angle. The system is specific about which is known and which is unknown only if 2 curves are present.

4. LINE 4. ELEVATION ANGLE(S). This can be '15' for 15° elevation data, '27' for 27° elevation data or '15,27' if both elevation angles are used. Elevation angle is printed if it varies between curves, so if there were 5 curves and only two entries in LINE 4, say '15,27' and '27', the last 4 curves would be representative of 27° elevation data.

5. LINE 5. A PRIORI KNOWLEDGE OF ASPECT ANGLE. This is compiled in the same way as LINE 3.
6. **LINE 6. MINIMUM, MAXIMUM AND INCREMENT OF ASPECT.** The minimum and maximum aspect angles can be any of those listed in Table 3.1. If these parameters vary from curve to curve, they are printed out for each curve in the same way as LINE 4 (or LINE 2).

7. **LINE 7. NUMBER OF FREQUENCIES.** This is printed for each curve.

8. **LINE 8. NUMBER OF TARGETS.** This is always a multiple of 6, the number of ships. (No. of targets = 6 x No. aspect angles x No. elevation angles). This is printed for each curve.

9. **LINE 9. 90% CONFIDENCE INTERVAL AT 30% MISCLASSIFICATION.** This is calculated according to the discussion in 5.6, and is printed for each curve.

10. **LINE 10. CLASSIFICATION FEATURES.** These are listed as:

    'A' Amplitudes only
    'W' Differential phases only
    'AW' Amplitudes and phases
    'R' Relative Amplitude
    'RW' Relative Amplitude and phase
    'T' Time domain algorithm.

11. **LINE 11. IDENTITY OF CURVES WITH ASPECT OR ELEVATION ANGLE ERRORS.** This is printed only if there is an error in elevation angle or aspect angle. Note that the classification software does not allow both types of errors at once. For example, if 3 curves were plotted and the last two had aspect errors, LINE 11 would read

    'ASPECT ERR IN CURVE 2 3'

Curves are identified as follows:

- - - Curve 1
- - - - Curve 2
- - - - - Curve 3
- - - - - - Curve 4
- - - - - - - Curve 5
- - - - - - - - Curve 6
6.2 FREQUENCY BAND

6.2.1 INTRODUCTION

This set of experiments was designed to examine the effect of choosing a particular frequency band for various aspect angles, polarizations and algorithms. (The term algorithm implies the use of various features in the NN algorithm; see Table 6.1).

The available band of 2 to 7 MHz was split into 3 non-overlapping sub-bands, each containing 4 discrete frequencies separated by 0.4 MHz. These were

Band 1: 2 - 3.2 MHz
Band 2: 4 - 5.2 MHz
Band 3: 5.8 - 7.0 MHz.

Generally, it is expected that the lower of frequencies will provide the best classification performance. This is because the radar cross section at lower frequencies tends to vary less per unit bandwidth compared with higher frequencies. Consequently, for higher frequencies, (in the upper resonance region) a small change in a parameter such as aspect angle, or the addition of noise to the cross section amplitudes results in a greater change in the selected features compared to the change at lower frequencies. In this sense, the radar cross section frequency response is more reliable (i.e., impervious to small changes in orientation, frequency, etc.) at lower frequencies than at higher frequencies.

Note that for this experiment only a single elevation angle (27°) was used. This is justified by a study of the effect of elevation angle on classification presented below.
6.2.2 RESULTS AND CONCLUSIONS

Figure 6.1 shows a set of typical curves representing misclassification percentage versus post-processing SNR. Similar plots using other parameters are contained in Appendix A; the results for all curves have been summarized by compiling averages of misclassification percentage at 10 dB post-processing SNR and are presented in Figures 6.2 to 6.4.

From Figure 6.2 it is evident that for 0° and 180° aspect zones, there is no distinct preference for a particular frequency band. 90° aspect zone favours the lower bands, but not significantly. Taking an average across all 3 aspect zones results in a marginal preference for band 1, showing that in general, the performance for a given sub-band in the 2-7 MHz band is not significantly affected by aspect zone. This is encouraging in one respect; namely that if the ionosphere or range conditions necessitated the use of higher frequencies (in the 2-7 MHz band), then performance is not degraded for a particular azimuthal orientation of the target ship.

Figure 6.4 shows a clear preference for band 1 (2-3.2 MHz) for vertical and cross polarizations. For the other two polarizations, frequency band seems unimportant. In Figure 6.3, a clear precedence of band 1, band 2, followed by band 3 is set, for each algorithm. This seems to best conform to the initial expectation that classification is more reliable at lower frequencies (i.e., closest to L/λ=1). However, certain results show an insensitivity to frequency band, at least for the range of frequency bands available here. Choosing a band of much
Figure 6.1. Misclassification percentage versus post-processing SNR for 3 sub-bands in the 2-7 MHz band.
Figure 6.2. Average misclassification percentages at 10 dB post-processing SNR for 3 bands, as a function of aspect zone.
Figure 6.3. Average misclassification percentages at 10 dB post-processing SNR for 3 bands, as a function of algorithm.

Figure 6.4. Average misclassification percentages at 10 dB post-processing SNR for 3 bands, as a function of polarization.
higher frequencies, for example 15 to 16.2 MHz, might show a dependence on frequency band not measurable here. In conclusion, certain parameters, such as vertical and cross polarizations, and the classification algorithm (i.e., A, AW, W, A, or T) are markedly dependent on the selection of frequencies in the 2 to 7 MHz band.

6.3 COMPARISON WITH PREVIOUS WORK

6.3.1. INTRODUCTION

The purpose of this experiment was to provide a direct comparison between the work done by Chen [1] and that done here. The essential difference between Chen's data and the data used here is that the former exclusively uses an elevation angle of 0°, while the latter has elevation angles of 15° and 27°. The comparison was done using 27° elevation only, since classification statistics are similar for 27° and 15° elevations.

Five features (A, AW, W, A, and AW) in the NN algorithm, and the time domain algorithm were evaluated at 2, 4, 6, 8 and 12 frequencies, using vertical polarization, 27° elevation and aspect angles of 0°, 90° and 180° (aspect angles assumed known). The frequencies selected by using the above numbers were discussed in Chapter V, Section 5.2. The data at aspect angles 0°, 90° and 180° are nearly independent so that the results do not pertain to a particular aspect zone. (Chen used this approach in most of his work.) The evaluation using 12 frequencies was included to see if performance could be improved.
Classification results using known and unknown aspect angles at 8 frequencies, and other parameters as above, were compared with those in [1]. When the aspect angle is known, a noise-corrupted target is compared only with catalog members of the same aspect angle. Unknown aspect implies that the noisy target is compared with members of the catalog at all aspects (see Table 6.1).

6.3.2 RESULTS AND CONCLUSIONS

Figures 6.5 and 6.6 show typical classification results. Figures 6.7 and 6.8 indicate a general increase in misclassification percentage when 27° elevation is used in comparison with 0°. This increase is about 6% for each selected number of frequencies. Chen used 5 ships compared with the 6 ships used here, which should account for 3% of the difference (4/5 - 5/6). This still leaves a 3% improvement for 0° elevation over 27°. Although this difference is apparent for all numbers of frequencies, it is not sufficiently large to be conclusive about the effect of elevation angle on classification. However, the tilting of a target when elevation angles above 0° are used is equivalent to a compression in range of the various scattering structures, from the point of view of an incident electromagnetic wave. Furthermore, structures that are vertical at 0° elevation (hence good scattering centres for vertically polarized waves) become weaker scatterers when tilted. Consequently, the process by which nulls and peaks are produced in the amplitude returns, which essentially characterize a particular target, is weakened. Intuitively
Figure 6.5. Misclassification percentage versus post-processing SNR for 4 different numbers of frequencies, using amplitudes only.
Figure 6.6. Misclassification percentage versus post-processing SNR for 4 different numbers of frequencies, using the time domain algorithm.
Figure 6.7. Average misclassification percentages at 10 dB post-processing SNR for 3 different numbers of frequencies as a function of algorithm (from Chen [1]).

Figure 6.8. Average misclassification percentages at 10 dB post-processing SNR for 3 different numbers of frequencies as a function of algorithm.
classification is expected to be less reliable at higher elevation angles. Data sets at say 50° elevation might be taken to confirm this hypothesis.

Chen [1] established a ranking of features and algorithms in terms of lowest misclassification percentage for each number of frequencies. For NF=2, AW<A<W<T, for NF=4, T<AW=A<W, and for NF=8, T<A<AW<W=AW<A. Examination of Figures 6.7 and 6.8 shows this same ranking to hold at 27° as well. The time domain feature is still critically affected by the use of only 2 frequencies for 27° elevation (as one would expect, since this is the absolute minimum allowable to permit the operation of the DFT).

For both 27° and 0° elevation it is evident that the addition of phase information (comparing A with AW) plays a more significant role at 2 frequencies than at a higher number of frequencies. For example, with N=2 and 27° elevation, AW is produces 10% fewer misclassification errors than A alone. The contribution that phase information makes to classification performance seems to become smaller as the number of frequencies is increased. This phenomenon also manifests itself in the relative amplitude results. Using AW gives an average improvement of 10% over A, for N=4, 8, and 12, but 15% for N=2. Phase information, therefore, could become particularly valuable when certain conditions constrain the measurements, such as being restricted to two frequency measurements or having a multiplicative bias in the amplitude data. Generally if 4 or more frequencies are available and phase information is at hand, it is best to use the time domain algorithm.
The addition of 4 extra frequencies above the previously used number of 8 yields a small improvement of 1%. In some cases, the use of 12 frequencies results in more classification errors than the use of 8 frequencies. Figure 6.12 shows the variation of average misclassification percentage with number of frequencies. The minimal increase in performance in going from 8 to 12 frequencies and, in general, from a lower to a higher number of frequencies, is consistent with the findings of the previous chapter. There it was shown that using lower frequencies in the allotted 2-7 MHz band resulted in better performance, depending on which parameters were used. Increasing the number of frequencies means that higher frequencies would be included, thus yielding a minimal improvement.

The number of frequencies is usually the most costly parameter (in terms of processing time) in a classification procedure. Bearing this in mind, and using results like those in Figure 6.12, it should be possible to obtain an optimal number of frequencies using counterposing weightings for the number of frequencies and the acceptable level of misclassification percentage. In subsequent classification experiments 8 frequencies are used. This tends to de-emphasize the affect of isolated errors or biases.

The difference in average misclassification percentage between classifications using known aspect and unknown aspect angle is about 6% for 27° elevation angle data, and 5% for 0° elevation angle data. Figure 6.9 shows a typical classification result and Figures 6.10 and 6.11 summarize the results of all curves produced in this part of the experiment, the remainder of which are contained in Appendix A.
Figure 6.9. Misclassification percentage versus post-processing SNR for known and unknown aspect angles.
Figure 6.10. Average misclassification percentages at 10 dB post-processing SNR for known and unknown aspect angles as a function of algorithm (from [1]).

Figure 6.11. Average misclassification percentages at 10 dB post-processing SNR for known and unknown aspect angles as a function of algorithm.
Figure 6.12. Average misclassification percentage versus the number of frequencies used.

6.4 ELEVATION ANGLE

6.4.1 INTRODUCTION

This experiment was divided into 3 parts. The first was designed to examine the differences in classification performance between ship data at elevation angles of 15° and 27°. The second part investigated classification performance when the elevation angle is known and unknown, and the final part measured the effect of having an error in elevation angle.

Each part of the experiment was conducted at the various aspect zones (0°, 90°, 180°), polarizations (V, H, X, V/H) and algorithms. In order to keep the data processing task to a manageable size,
only the A, AW and $\ddot{A}$ features of the nearest neighbour algorithm were considered. The time domain algorithm was also used. This combination alone leads to a total of 48 graphs, with each part of the experiment for a given set of parameters on the same graph.

Data for only 2 elevation angles, 15° and 27° were available. For the 'unknown elevation' case, data for a noise-contaminated target ship is compared to data for ships at 15° elevation, and 27°. The only error syndrome available was ±12°. Hence, a noise-contaminated target ship known to be at 27° elevation would be compared with 15° elevation catalog members and vice versa. Chen's data [1] at 0° elevation were considered for use in this experiment but were ruled out on the grounds of the additional complexity necessary in the classification software for the minimal comparisons which would be obtained. (The data included 5 of the 6 ships used here, but only at vertical polarization and aspect angles of 0°, 90° and 180°.)

6.4.2 RESULTS AND CONCLUSIONS

Figures 6.13 to 6.16 show typical classification results (see Appendix A for others) and Figures 6.17 to 6.21 summarize all results in terms of the average misclassification percentage at 10 dB post-processing SNR.

Figures 6.17 to 6.19 show a high degree of similarity between classifications at 15° elevation and those at 27°. For aspect zone, polarization and algorithm, 15° elevation has, on average, an improvement of 1.7% fewer classification errors than 27° elevation. Although this improvement is small, it is consistent with all parameters.
Figure 6.13. Misclassification percentage versus post-processing SNR for various elevation angles.
Figure 6.14. Misclassification percentage versus post-processing SNR for various elevation angles.
Figure 6.15. Misclassification percentage versus post-processing SNR for various elevation angles.
CLASSIFICATION OF SHIPS
POLARIZATION V/H
ELEV ASSUMED KNOWN / UNKNOWN
ELEVATION (DEG.) 15 27 15.27
ASPECT ASSUMED KNOWN
MIN, MAX, INC ASPECT 0 10 10
NO OF FREQUENCIES 8
NO OF TARGETS 12 12 24 24 24
90% CI (±30%) +/- 3.1% 3.1% 2.2% 2.2% 2.2%
CLASS. FEATURES A A A A A
ELEV ERR IN CURVE 4

15 DEG. ELEVATION
27 DEG. ELEVATION
15 AND 27 DEG. ELEVATION (KNOWN)
15 AND 27 DEG. ELEVATION (WITH ERROR)
15 AND 27 DEG. ELEVATION (UNKNOWN)

Figure 6.16. Misclassification percentage versus post-processing SNR for various elevation angles.
Figure 6.17. Average misclassification percentages at 10 dB post-processing SNR for various elevation angles as a function of aspect zone.

Figure 6.18. Average misclassification percentages at 10 dB post-processing SNR for various elevation angles as a function of polarization.

96
Figure 6.19. Average misclassification percentages at 10 dB post-processing SNR for various elevation angles as a function of algorithm.

Figure 6.20. Average misclassification percentages at 10 dB post-processing SNR for an elevation error of ±12° as a function of aspect zone and polarization.
except X polarization (where 27° is preferred), and it agrees with the results of the previous section where 0° elevation was found to have, on average, 3% less misclassification percentage as compared to 27° elevation. The difference in angle between 15° and 27° is clearly not large enough to provide a distinct change in classification performance.

Figures 6.17 to 6.19 also show the difference in misclassification percentage between known and unknown elevation angles when a data set comprising 2 elevation angles is used. Examining the known elevation results reveals an average classification error of 27%, which is 4% higher than the average of 15° and 27° elevations (23%) taken alone. Intuitively these would be expected to be the same since

Figure 6.21. Average misclassification percentages at 10 dB post-processing SNR for an elevation error of ±12° as a function of aspect zone and algorithm.
algorithmically, known elevation (or aspect) simply compiles errors averaged over the known elevations (or aspects). The difference in performance is particularly distinct for the case of AW using V/H polarization at broadside (see Figure A.40), and this is a clue as to the cause of this discrepancy. The average of separate classifications at 15° and 27° elevation gives 15% classification error, while 15° and 27° known elevation gives 44% misclassification. Examining the average signal powers for each 'nose-free' data base yields 38 dB for 15°, 21 dB for 27° and 35 dB for 15° and 27° elevation (known). The average of 38 dB and 21 dB is

$$10 \log_{10} \left( \frac{10^{38/10} + 10^{21/10}}{2} \right) = 35 \text{ dB}.$$  

Consider the case of 15° and 27° elevation (known). If a SNR of 10 dB is required, then 25 dBm² of noise power must be added to the data base amplitude (and phase) returns. However, this leads to 13 dB SNR for the 15° part of the data base, and -4 dB SNR for the 27° part of the data base. Consequently the 15° and 27° known elevation case is an average of a slightly better classification using 15° elevation data, and a very much poorer classification using 27° elevation data (at 10 dB SNR). In other words, the 15° and 27° known elevation result is only the average of the individual results when other parameters allow the average signal powers of the 'noise-free' data bases to be the same (X polarization is such a case).
This result follows directly from the discussion in Chapter 5.5 where two points were established. First, the classifications for classes having widely varying average signal powers have a constant injected noise power, but variable (post-processed) SNR for each class. Second, the separate classifications for the classes mentioned above will have constant (post-processed) SNR with corresponding different injected noise levels. Results due to Chen [1] and results established here can be used for meaningful comparisons if the above points are taken into consideration.

It is evident from Figures 6.17 to 6.19 that a priori knowledge of elevation angle results in 5% fewer in classification errors. This result is independent of aspect zone but does vary with polarization and algorithm. Specifically, H polarization, and relative amplitude $\bar{A}$, (not necessarily together) produce significant increases in misclassification percentage for an unknown elevation angle, where there is a 10% difference between the results for known and unknown elevation angle. These parameters are, in general, less favourable toward low misclassification and thus, classification without a priori knowledge of the elevation angle is particularly sensitive to 'adverse' parameters.

The high values of misclassification percentage when an error is made in elevation angle (see Figures 6.20 and 6.21) are due in part to the varying SNR problem mentioned earlier. However, note that for X polarization, where signal powers for 27° elevation and 15° elevation are equal (see Table B.1) and the special consideration does not apply, that classification error is still about 66%. Hence it must be
concluded that the majority of the classification errors introduced with an elevation ambiguity of ±12° stem from the dissimilarity of targets at 15° and 27° elevations. This might seem contrary to previous findings where classification performance at 15° was found to be very similar to classification performance at 27°. It must be noted that similarity in classification performance between two given parameters is not directly related to the similarity of the targets associated with those parameters.

Classification error, as a result of an elevation angle error of ±12°, is about 74% on average. This result is independent of aspect zone and algorithm. Hence erroneous a priori elevation information has a catastrophic impact on classification; in such an event it would be better to discard the elevation angle information and assume that elevation were unknown, assuming there was no other recourse: (see the suggestions for future work in Chapter VII).

6.5 POLARIZATION

6.5.1 INTRODUCTION

The purpose of this experiment was to determine the relative classification performance of each polarization; vertical, horizontal, cross and vertical divided by horizontal, as a function of aspect zone and algorithm. An elevation of 27° was used in this experiment.

Corrupting features particular to the ionosphere, such as Faraday rotation, were not simulated here. Also, the fact that cross
polarization may not be practical in conjunction with ionospheric propagation did not preclude its investigation.

It is worth amplifying the purpose of using V/H polarization. Returns for this 'polarization' are produced by simply dividing the vertically polarized complex returns \( \text{AV}, \theta_V \) by the corresponding horizontally polarized complex returns \( \text{AH}, \theta_H \), i.e.,

\[
\text{AV/H} = \frac{\text{AV}}{\text{AH}} \quad \text{dBm}^2 \tag{6.1}
\]

\[
\theta_{V/H} = \frac{\theta_V}{\theta_H} \tag{6.2}
\]

where \( i \) is the frequency index.

In general, a target scatters components of vertically and horizontally polarized electromagnetic energy when illuminated by radiation of a given polarization (vertical or horizontal). Since these components have the same frequency (and doppler shift), and are scattered simultaneously, they take the same ionospheric path and are hence subject to the same ionospheric distortion. This distortion can be approximately modeled as a complex multiplicative factor. If \( \tilde{A}_V^i \) and \( \tilde{A}_H^i \) are the complex target returns at a given frequency, and \( C e^{j\phi} \) is the appropriate ionospheric distortion factor, then the received returns are given by

\[
\tilde{A}_{V,r} = \tilde{A}_V \cdot C e^{j\phi} \tag{6.3}
\]

\[
\tilde{A}_{H,r} = \tilde{A}_H \cdot C e^{j\phi} \tag{6.4}
\]
Hence,

$$\tilde{A}_{V/H} = \frac{\tilde{A}_{V,r}}{\tilde{A}_{H,r}}$$

and the multiplicative factor is removed. However, since multiplicative corruption was not simulated in this experiment, this hypothesis was not tested. Furthermore, this simple analysis might not adequately describe the corrupting nature of the ionospheric channel, although intuitively at least, one would expect some reduction in multiplicative factors by this type of operation.

6.5.2 RESULTS AND CONCLUSIONS

Figure 6.22 shows a typical result of classification performance using various polarizations the remainder of which are contained in Appendix A. Figures 6.23 and 6.24 summarize the classification performances at 10 dB post-processing SNR in terms of averages for a given parameter. On average, V polarization and X polarization rank first (with about 20% misclassification error), followed by V/H polarization (29%) then H polarization (44%). At 0° and 180°, V polarization and X polarization give similar performance, however at 90°, X polarization has 10% fewer classification errors compared with V polarization. V polarization and X polarization also give similar performances for all algorithms except $\tilde{A}$ (relative amplitude). It is significant that performance differs only when the parameters in question are particularly adverse toward classification (i.e.,
CLASSIFICATION OF SHIPS
POLARIZATION V H X V/H
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN
MIN, MAX, INC ASPECT 0 10 10
NO OF FREQUENCIES 8
NO OF TARGETS 12 12 12 12
90% CI (+30%) +/- 3.1% 3.1% 3.1% 3.1%
CLASS. FEATURES A A A A

Figure 6.22. Misclassification percentage versus post-processing SNR for various polarizations.
Figure 6.23. Average misclassification percentage at 10 dB post-processing SNR for various polarizations as a function of aspect zone.

Figure 6.24. Average misclassification percentage at 10 dB post-processing SNR for various polarizations as a function of algorithm.
90° aspect zone or A). This similarity in performance between V and X polarizations might stem from a combination of weaker scattering from the tilted vertical structures in the first case, and enhanced scattering from the tilted structures in the second case. Following this line of reasoning, V polarization might be expected to have better classification performance than X polarization, at zero degrees elevation angle.

H polarization consistently exhibits the poorest classification performance and this is particularly true at 0° and 180° aspect zones where H polarization is 20% and 30% worse than the V polarization results. This tends to reinforce the idea that, for bow and stern aspects, the major scatters are vertical and that these play a significant role in the classification process. H polarization exhibits its best performance at 90° aspect, where it has a similar misclassification percentage to V polarization (32%). At broadside, the ship's hull is a long horizontal structure, which is a good scatterer of horizontally polarized energy; at stern and bow aspects, there are few horizontal structures which may account for H polarization's preference for the 90° aspect zone. Data in Table B.1 in Appendix B support this conclusion. Radar cross sections for H polarization average to about 27 dBm², whilst those for V polarization average 20 dB higher. At 90° for H polarization, radar cross section is about 29 dBm², and for bow and stern, radar cross section is about 26 dBm².

X polarization has the best all-round performance. Although on average X and V polarization has the same 20% classification error, V
polarization is particularly affected at 90° aspect zone, having an error of 30%, whereas X polarization stays at the average.

V/H polarization on average has a misclassification error of about 30% which is roughly half way between that of V polarization and H polarization (which intuitively would seem reasonable). However, when dealing with data from a real system, we may be able to estimate V/H amplitudes more accurately than V amplitudes, assuming that it is possible to measure both the horizontally and vertically polarized components of the scattered energy. The discussion in 6.5.1 showed that multiplicative errors, which are characteristic of the ionosphere, cancel (see Equations (6.1) to (6.5)) when using the V/H parameter. This same division may also eliminate the need to use a reference, (since it would also cancel out). Hence the use of V/H polarization, as opposed to say V polarization, according to this analysis, removes two of the major sources of error in our process, i.e., the ionosphere and the referencing system. This is, potentially, a very valuable feature and the extent to which these errors are removed or reduced is a potential subject of further research.

The discussion thus far has concentrated much on the relative performance of polarizations with respect to each other. It is not the intention here to find an optimal polarization and then build a system using it; the same is true of any of the other parameters of classification. Indeed, limitations in a real sky-wave radar system might require the use of a particular polarization, so it is desirable to know the associated classification performance resulting from such a restriction.
6.6 ASPECT ZONE

6.6.1 INTRODUCTION

This experiment was designed to investigate how classification is affected by a particular aspect zone. From the above experiments, it is evident that classification performance is very dependent on the azimuthal orientation of the ship (i.e., aspect angle). Furthermore, from the discussion in 6.4.2, it might not be meaningful to do classifications where the catalog contains data pertaining to ships at widely varying aspect angles (for example 0° and 90°). Hence, a thorough investigation of classification performance at given aspect angles is of interest.

An elevation angle of 27° was used, and classifications at the 3 aspect zones, bow, stern and broadside, were done for each polarization and algorithm. A 45° aspect zone, comprising angles 30°, 45° and 60° was also included for the case of vertical polarization.

Based on the work of previous experiments, it was expected that 0° and 180° aspect zones would have similar classification performance, and 90° aspect zone would be worse.

6.6.2 RESULTS AND CONCLUSIONS

Figures 6.25 and 6.26 show typical curves of misclassification percentage versus post-processing SNR, the remainder of which are contained in Appendix A. Figures 6.27 and 6.28 summarize the results of
this experiment in terms of the average error at 10 dB post-processing
SNR, for various polarizations and algorithms.

The initial expectation that classification performance at 0\degree aspect zone would be similar to that at 180\degree is true in general; the average classification error at 0\degree is 31\%, and 29\% for 180\degree (averaged across all polarizations, or algorithms). However, in the particular case, the above is only true for vertical polarization, the
amplitude-only (A) feature, or the time domain algorithm. These 3 classification variables are generally the most favourable towards achieving low misclassification percentage, and 0\degree aspect behaves like 180\degree only under these conditions.

90\degree aspect zone exhibits best classification performance, for H polarization and V/H polarization. This is complimentary to the finding of the previous section where the above polarizations had best performance at 90\degree aspect zone. This does not mean, however, that it is more advantageous to use H polarization if the target is known to be at broadside; from Figure 6.27, we see that either X or V/H polarization are be the most advantageous.

For V/H and H polarizations, 90\degree aspect zone has 14\% fewer classification errors on average, compared with bow and stem aspects. For V and X polarizations, 180\degree and 0\degree aspect zones are better than broadside by an average of about 9\%. Hence when averages are compiled across polarizations to produce the figures used in Figure 6.28, the difference in performance between the two pairs of polarizations
CLASSIFICATION OF SHIPS
POLARIZATION    V
ELEV ASSUMED    KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED  KNOWN / KNOWN
MIN, MAX, INC ASPECT 0 10 10, 30 60 15, 80 100 10, 170 180 10
NO OF FREQUENCIES 8
NO OF TARGETS 12 18 18 12
90% CI (#30%) +/- 3.1% 2.5% 2.5% 3.1%
CLASS. FEATURES  A A A A

Figure 6.25. Misclassification percentage versus post-processing SNR for various aspect zones using vertical polarization.
Figure 6.26. Misclassification percentage versus post-processing SNR for various aspect zones using V/H polarization.
Figure 6.27. Average misclassification percentage at 10 dB post-processing SNR for various aspect zones, as a function of polarization.

Figure 6.28. Average misclassification percentage at 10 dB post-processing SNR for various aspect zones, as a function of algorithm.
(i.e., V and X, compared to H and V/H) tends to enhance the performance of the 90° aspect zone. (See, for example, the time domain algorithm averages in Figure 6.28.) Consequently, when examining the performance of specific algorithms at various aspect zones, one should bear in mind the particular polarization used.

6.7 ERROR IN ASPECT

6.7.1 INTRODUCTION

The heading of a ship, and hence its aspect can be found quite accurately by simply plotting its position at two instances of time. Knowing the time interval and range between the reference positions allows the average velocity to be calculated, and this in conjunction with the doppler shifts at the two positions allows the aspect measurement to be refined.

An error in the aspect can be introduced when the heading of a ship is altered by a sea current. Figure 6.29 shows how this may happen. A ship traveling from \( P_1 \) to \( P_2 \) with velocity \( V_s \) is estimated to have an apparent aspect of \( \theta_a \). However, the component of sea current \( V_c \) forces the ship to take an aspect of \( \theta_a + \theta_e \) in order to travel in the direction \( P_1 \) to \( P_2 \).

\[
\theta_e = \tan^{-1} \left( \frac{V_c}{V_s} \right).
\]

Typical values for \( V_c \) and \( V_s \) might be 5 and 20 knots, respectively, which leads to an error of \( \theta_e = 14^\circ \).
Figure 6.29. A ship travelling with velocity $V_s$ from point $P_1$ to point $P_2$ has an apparent aspect angle of $\theta_a$ degrees owing to the sea current having a velocity $V_c$.

The purpose of this experiment was to examine classification performance as a function of various parameters when an aspect error is introduced. In general, for a given aspect, a $\pm \theta_e$ aspect error must be considered. For $0^\circ$ aspect zone, using $0^\circ$ and $10^\circ$ aspects, a $\pm 10^\circ$ error was simulated by comparing $0^\circ$ noise-contaminated ship returns with $10^\circ$ catalog ship returns and vice versa. At broadside, where 3 aspects were used ($80^\circ$, $90^\circ$ and $100^\circ$), a $\pm 10^\circ$ aspect error was simulated by comparing $80^\circ$ and $100^\circ$ aspect noise-contaminated targets with a $90^\circ$ catalog, and $90^\circ$ aspect noise-contaminated targets with $80^\circ$ and $100^\circ$ catalogs. This represents performing an additional experiment for each ship and this
must be taken into consideration when calculating the average misclassification percentage for a given SNR.

This experiment was divided into 2 parts: the first to examine the effect of different aspect errors near to 0° aspect angle, and the second to examine the effect of 10° aspect errors at different aspect zones. Curves representing classification performance at 0° and 15° using unknown aspect for each of the parameters used in the above experiments, were derived in order to establish a threshold at which it would be wiser to stop estimating an aspect angle. An elevation angle of 27° was used in all experiments.

6.7.2 RESULTS AND CONCLUSIONS

Figures 6.30 to 6.32 show typical classification results for these experiments, the remainder of which are contained in Appendix A. Figures 6.33 and 6.34 show average misclassification percentages at 10 dB post-processing SNR for various errors in aspect near to 0° aspect. It is clear that performance is dependent on the particular polarization used. For V and H polarizations, a ±10° error is roughly equivalent to saying that the aspect is unknown. Hence for aspect angles near to 0° which can only be estimated with an accuracy of less than ±10°, it would be better to re-evaluate the use of a priori aspect angle information. For X and V/H polarizations an error of ±10° aspect leads to a large increase in misclassification percentage (to about 60%) compared with 15% for V polarization. Based on the increase in misclassification
Figure 6.30. Misclassification percentage versus post-processing SNR for various aspect errors, near to $0^\circ$ aspect, using vertical polarization.
CLASSIFICATION OF SHIPS
POLARIZATION X
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN / KNOWN
MIN, MAX, INC ASPECT 0 10 10, 0 15 15, 0 30 30
NO OF FREQUENCIES 8
NO OF TARGETS 12 12 12
90% CI (#30%) +/- 3.1% 3.1% 3.1%
CLASS. FEATURES A A A
ASP ERR IN CURVE 2 3

Figure 6.31. Misclassification percentage versus post-processing SNR for various aspect errors, near to 0° aspect, using cross polarization.
Figure 6.32. Misclassification percentage versus post-processing SNR for an aspect error of ±10° at various aspect zones.
Figure 6.33. Average misclassification percentage at 10 dB post-processing SNR for various aspect errors, near to 0° aspect, as a function of polarization.

Figure 6.34. Average misclassification percentage at 10 dB post-processing SNR for various aspect errors, near to 0° aspect, as a function of algorithm.
percentage (shown in brackets) when an aspect error of ±10° is introduced, the polarizations acquire the following ranking

\[ V(4\%) < H(13\%) < V/H(28\%) < X(48\%) \]

It is evident that X polarization, which was previously established as having good classification properties, is particularly susceptible to an error in aspect angle. Furthermore, V polarization is the only polarization with 30% or less misclassification for errors of ±10° or less.

Establishing a ranking of algorithms based on the increase in misclassification percentage when an error of ±10° is introduced yields

\[ \tilde{A} W(16\%) < A W(18\%) < A(24\%) < T(39\%) \]

This is the exact reverse order that these algorithms appear if there is no error in aspect. The time domain algorithm is more affected by an aspect error than the other algorithms. At 0° error, T has about 10% classification error on average, which is the same as A. However, at ±10° error T has about 50% error compared with 35% for A. In conclusion, for the aspect zone of 0° (bow), the parameters X polarization and the time domain algorithm, which previously had good records of classification performance, are particularly sensitive to errors in aspect angle.

Figures 6.35 and 6.36 show how a ±10° aspect error affects classification percentage at different aspect zones. Note that the histogram bars represent the difference between the results of classification at 0° error and classification at ±10° error. For X
polarization it has already been mentioned that classification performance is significantly degraded (by 45% on average) by the introduction of a ±10° error at 0° aspect zone. This is also true of X polarization for 90° and 180° aspect zones.

For the remaining polarizations it is also evident that 180° is more affected by an error of ±10° aspect error than is 0° aspect zone, and performance is poor (always above 50% classification error) irrespective of the polarization. V/H and H polarization have poor performance at all aspect zones when a ±10° aspect error is introduced; typically misclassification percentage is always above 50%.

Generally, the time domain algorithm is most affected by an aspect error of ±10°, especially at 90° aspect zone. Based on the results, the choice of algorithms to provide lowest misclassification percentage at 0°, 90° and 180° aspect zones would be

<table>
<thead>
<tr>
<th>ASPECT ZONE</th>
<th>0°</th>
<th>90°</th>
<th>180°</th>
</tr>
</thead>
<tbody>
<tr>
<td>ERROR</td>
<td>T</td>
<td>T</td>
<td>T</td>
</tr>
<tr>
<td>0°</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>±10°</td>
<td>A</td>
<td>AW</td>
<td>T</td>
</tr>
</tbody>
</table>

Chen [1] also did an analysis of the affect of aspect errors. Using sets of ship data at 0°, 15° and 30° aspects (vertically polarized) he compared the no-noise distances and correlation coefficients for various ships and found that about 4 times out of 5, the closest neighbour to a particular ship was the same ship at a
Figure 6.35. Average difference in misclassification percentage at 10 dB post-processing SNR between classifications having a $0^\circ$ aspect error and classifications having a $\pm 10^\circ$ aspect error, at various aspect zones, as a function of polarization.

Figure 6.36. Average difference in misclassification percentage at 10 dB post-processing SNR between classifications having a $0^\circ$ aspect error and classifications having a $\pm 10^\circ$ aspect error, at various aspect zones, as a function of algorithm.
different aspect angle, indicating a certain toleration towards errors in aspect angle. Results here would tend to confirm Chen's findings for the particular case of vertical polarization and 0° aspect zone (see Figure 6.33). However, for other polarizations and aspect zones, classification was found to be significantly degraded by the inclusion of an aspect error of ±10°. Also, the inclusion of added noise in this analysis revealed a hitherto indistinguishable sensitivity in the particular algorithm (especially in the time domain algorithm) to errors in aspect angle.
CHAPTER VII
CONCLUSIONS

A series of experiments investigating the classification properties of a group of ships at various azimuthal and elevation angles, using a variety of polarizations, have been performed. These experiments, to a certain extent, were intended to represent classification based on HF sky-wave resonance radar returns.

The characteristics and specifications of an HF sky-wave radar system were discussed. These include the considerably longer range (compared with ground-wave propagation) afforded by sky-wave propagation, which is attained at the cost of contamination and dispersion by the ionosphere. These and other factors tend to severely restrict the range of operating frequencies available for classification purposes. The methods of measuring amplitude and phase returns and the use of a reference, such as sea-scatter, for calibration of the radar cross section amplitudes have been discussed. A technique using the ratio of the returns at vertical and horizontal polarizations has been suggested as a means of obviating the need to calibrate with a separate reference.
The construction of a data base, using radar backscatter measurements from scaled-model ships, has been described. The use of a groundplane to simulate the surface of the sea resulted in residual time responses corresponding to the location of the groundplane edges in the target's impulse response. Impulse responses were used to check the validity of calibrations. The calibration process was found to be both time consuming and intricate, but necessarily so in order to provide representative data. The calibration and scaling procedures converted measured ship amplitude and phase returns from 2 to 18 GHz into scaled amplitude and phase returns from 2 to 7 MHz.

Various sources of noise and errors have been discussed. Noise in the context of our classification experiments was defined as post-processing noise; the errors remaining after the processing system had done its best to minimize the various contaminations. A need was expressed to incorporate into the noise analysis a sufficiently detailed representation of the ionospheric channel and its associated distortions. No such model was available and in view of this and under other considerations, an independent additive Gaussian noise model was used.

In view of the limitations placed on available operating frequencies, the 2 to 7 MHz band was divided into 3 equal sub-bands of 4 frequencies to investigate the classification properties of each band. In general, the lower band (as expected) gave the lowest probabilities of misclassification, but in particular, the result depended on the parameter of interest. Vertical and cross polarizations each showed a
marked difference of about 16% in misclassification percentage between the lower and middle bands, whereas horizontal and V/H polarization had an average difference of about 3%. The difference in performance between band 1 and band 3 was not great (always less than 20%), hence if conditions dictated the use of higher frequencies in the 2-7 MHz band, then classification performance does not suffer badly.

The next set of experiments served as a comparison to previous work done by Chen [1]. In general the results using 27° elevation data, were similar to those of Chen's 0° elevation data. Increasing the number of frequencies was found to reduce misclassification percentage, but only to a limit. For example the use of 12 frequencies, compared to 8 frequencies, yielded a 1% improvement on average. The same ranking of algorithms established in [1] was also established here. For 2 frequencies this was, in terms of lowest misclassification percentage, AW < A < W < T and for 8 frequencies, T < A < AW < W. Generally the addition of phase information (for example AW, or AW) was found to be more significant in terms of improved classification performance, for a lower number of frequencies.

A priori knowledge of aspect angles provided a 6% decrease in misclassification percentage (compared to assuming the aspect angle as unknown), which was similar to Chen's finding. In general, there was a small but accountable improvement of about 3% less misclassification for the 0° data compared to the 27° data.

The investigation performed on classifications at elevation angles of 27° and 15°, revealed that there was a high degree of similarity
between the results for the two angles. A 1.7% reduction in classification error was established for the 15° elevation data over the 27° data; and this low value was representative of all polarizations, aspect zones and algorithms. When using a catalog containing data relating to two elevation angles, a 5% reduction in classification error resulted when the elevation angle was known a priori. This was comparable to the result for a priori knowledge of aspect angle. A ±12° error in elevation lead to a drastic degradation in classification performance; typically misclassification percentage was always worse than 65%, with some parameters, such as horizontal polarization being little better than random guessing (i.e., 5/6 or 88%), at 80% misclassification. No one polarization, aspect zone or algorithm showed any particular immunity to the effects of a ±12° error in elevation.

Vertical and cross polarizations were found to have similar classification properties, with about 20% classification error on average. V/H followed with 29%, and H polarization was consistently worst with 44% misclassification. These results were very much dependent on the particular aspect zone used, with vertical and cross polarizations doing best at 0° and 180° aspect zones, and horizontal and V/H polarizations doing best at 90° aspect zone.

V/H polarization was found to have a performance lying between that of vertical polarization and horizontal polarization. However, the V/H polarized amplitudes may, in practice, be measurable with a higher degree of accuracy than the vertically polarized amplitudes. This is because theoretically at least, multiplicative factors impressed by the
ionosphere should cancel out. Also the use of V/H polarization might preclude the need to use a reference. Unfortunately these sources of distortion could not be simulated here and so a real test of V/H polarization was not performed. However, it was established that the diversion of amplitudes in this manner does not drastically reduce their information content, otherwise the misclassification levels would have been much higher.

The analysis of classification performance at various aspect zones revealed a high degree of dependence on the particular aspect zone (0°, 90° or 180°). The initial expectation that 0° aspect zone would possess similar classification properties to 180° aspect zone was true only for vertical polarization, the amplitude only feature (A), or the time domain algorithm. These 3 classification variables generally were found to be the most favourable towards achieving low misclassification. The behaviour of 0° and 180° aspect zones for other parameters differed. 90° aspect zone favoured horizontal and V/H polarization. Its apparently good performance for certain features, such as AW in the nearest neighbour algorithm, was due to the relative performances of polarizations at certain aspect angles. As a result of this, the performance at a particular aspect zone needed to be judged with respect to a particular polarization, and averaging across polarizations tends to somewhat obscure the analysis.

Finally, the investigation of the effects of aspect errors revealed a particular sensitivity of some parameters to such errors. For example X polarization and the time domain algorithm both had a
misclassification percentage increase of more than 40% when a ±10° error was introduced. Vertical and horizontal polarizations had error levels of 13% and 49% respectively, which was equivalent to their performances when the aspect angle was assumed known. Generally, the levels of classification error, with the exception of vertical polarization, were quite high for a ±10° error, at all aspect zones; with misclassification levels of 50% and 60% being common.

The above results have illustrated how the classification of ships is dependent on the frequencies, polarizations aspect and elevation angles, and algorithms used in the classification process. Assuming that no errors have been made in estimating aspect and elevation angles, ship targets were correctly identified with a probability ranging approximately from 50% to 100%, depending on the prevailing conditions. Furthermore, the results presented here for the given catalog of ships are an 'upper-bound', for a variety of reasons. Firstly, in practice there would be a vast amount of a priori information, such as the knowledge of ship movements (perhaps from satellite photography), which could further reduce the catalog size. Estimation of the average amplitude returns for a ship, in conjunction with aspect and elevation angle information, could also allow further reduction of the catalog. In these experiments, a ship was considered misclassified even if the right ship was chosen at the wrong aspect or elevation angle; allowing such a decision to be considered as correct would also reduce the probability of making an incorrect decision. Finally, there is considerable potential for refining and optimizing the various algorithms for classification (some suggestions are discussed below).
Based on the experience gained whilst performing the various investigations of this report, it is suggested that future work be directed in the following areas.

The nearest neighbour technique for classifying targets involves calculating the distances at specific sample frequencies, between the test target and a particular catalog reference. The square root of the sum of distances over these frequencies is called 'the distance' between the unknown test target and the noise-free catalog member. This distance, compared with other such distances is used to make a classification decision. The following example is designed to demonstrate how the individual distances, for each frequency, contain useful information.

Consider 2 sets of arbitrary distances between an unknown target, \((S_u)\), and 2 catalog targets, \((S_1, S_2)\).

<table>
<thead>
<tr>
<th>Distance</th>
<th>(D_{u,1})</th>
<th>(D_{u,2})</th>
</tr>
</thead>
<tbody>
<tr>
<td>(f_1)</td>
<td>3</td>
<td>8</td>
</tr>
<tr>
<td>(f_2)</td>
<td>2</td>
<td>8</td>
</tr>
<tr>
<td>(f_3)</td>
<td>4</td>
<td>10</td>
</tr>
<tr>
<td>(f_4)</td>
<td>72</td>
<td>10</td>
</tr>
<tr>
<td>Distance</td>
<td>9</td>
<td>6</td>
</tr>
</tbody>
</table>

On this basis, \(S_u\) would be classified as \(S_2\). Observe though that 3 out of the 4 frequencies suggest that \(S_u\) is \(S_1\) and the fourth frequency
has biased the overall result sufficiently to change the classification. Such a syndrome is representative of impulsive noise, for example, lightning discharges which affect HF radar returns.

From the above demonstration, it is clear that the examination of distances for individual frequencies is necessary in order to deal with impulsive noise. A simple 'majority vote' system would be one way of implementing such a procedure. However, the majority vote system would involve extensive alterations to existing algorithms and software, and presents a problem in resolving ties.

One way around this problem would be to identify the 'rogue' distances on a statistical basis. For example, if a distance was more than, say, 1 standard deviation from the mean, then it could be set to some other value, perhaps the mean of the other distances. In the above example, the means are 20 and 9, and the standard deviations are 30 and 16 respectively. Hence all the frequencies of $S_2$ pass the test, but $f_4$ of $S_1$ has a distance which is 1.7 standard deviations above the mean. Consequently, it is set to 3, the average of the unaffected frequencies, resulting in a new total distance of 3.5. $S_u$ is now closer to $S_1$ than $S_2$.

Research is needed concerning the statistical distribution of individual target distances. The above test works with the arbitrary numbers presented, but may not do so with more practical figures. It would be undesirable to alter the distance of a particular target that is not subject to impulsive contamination.
From the previous discussion it is evident that information is contained in the actual values of NN distances (and correspondingly in time domain algorithm correlation coefficients). This information might further be exploited to reduce the sort of classification errors which are encountered when errors are made in the estimation of aspect and elevation angles. For example, if the heading of an unknown ship, \( S_u \), is measurable to within ±15°, then classification can proceed for each of 7 catalogs, \( C_1 \) to \( C_7 \), one for the estimated aspect angle and the others at 5 degree increments above and below the estimate. From each catalog there would be a candidate classification, \( S_n \), with a corresponding minimum distance (or maximum correlation coefficient), \( D_n \), for \( n = 1, 2, \ldots, 7 \). The choice of a particular candidate, for the case of the NN algorithm, would be based on these distances. Average distance in the NN algorithm is a function of post-processed SNR, hence if the latter could be estimated accurately, the candidate classification having a distance corresponding to the measured post-processing SNR (for a given angle) would be chosen.

Let \( D_{n,\theta}(p) \) be the distance (generated by the NN algorithm) between the noise-free returns of a ship, \( n \), at aspect angle \( \theta \), and the same returns contaminated by noise such that post-processing SNR was \( p \) dB. Assume for this example that \( p \) is 10 dB, and the aspect angle was estimated at 15°.
Where

\[ D_n = \text{Distance between ships } S_u \text{ and } S_n \]
\[ D_{n,\theta}(10) = \text{Distance between ships } S_n \text{ and } S_n^* \]
\[ S_n^* \text{ is a noise-contaminated version of } S_n. \]

The classification would proceed by choosing the smallest magnitude

\[ |D_{n,\theta} - D_n| \text{ for } n = 1, 2, \ldots, 7 \]
\[ \theta = 0.5, \ldots, 30. \]

Research is needed to examine the relationship between post-processing SNR and distance (or correlation coefficient), and how this varies with aspect and elevation angles and their increments.

Classification experiments thus far have produced misclassification percentages of between 0 to 50% for a nominal 6 ship catalog. In practice, even with a large amount of a priori information a working catalog size might be in the order of 100 ships or more. As a result, classification errors will be higher because of the greater chance that
the noise-contaminated returns of an unknown ship would look like the noise-free returns of a catalog ship.

Perhaps one way of addressing such a problem might be to investigate the classification properties of 'generic' ship targets. The term generic is used here to describe a set of amplitude and phase returns representative of a group of ships. Hence, a catalog would not contain the returns of, say 30 destroyers, but rather the 'generic' returns of, say 3 representative destroyers.

The generic forms could be constructed from basic shapes representing what are thought to be the major scattering structures common to a group of targets. In a previous chapter, it was postulated that vertical structures, such as masts, play a significant part in classification using vertically polarized returns. The generic form could simulate, to varying degrees, such structures and this would help establish the extent to which the various features of a ship contribute to the characterisation of its amplitude and phase returns. In a sense, such experiments would investigate the 'resolution' of the classification system, i.e., the degree to which the system can distinguish between similar targets.

The use of the AW feature in the NN algorithm produces a hitherto unresolved effect. Referring to Figures A.14 and A.16, it is evident that for two frequencies the AW feature is always better than the A feature, whereas for 8 frequencies the A feature is better than the AW feature above 5 dB post-processing SNR. Intuitively, one would expect the AW feature to be better than the A feature because it appears to
contain more information about a target (having both amplitudes and phases).

This discrepancy is probably a consequence of Chen's assertion that the expression for normalizing constant, $K$ (see Chapter IV Section 6.2) is not optimal. Since the relative performance of the A and AW features seems to be dependent on post-processing SNR, it is likely that a given amount of noise contaminates the A features and AW features to different extents.

Figure 7.1 shows the variation of noise-dependent normalizing constant $K_n$ with post-processing SNR. Here the constant is determined from the variances of the noisy amplitudes and the noisy phases, rather than the noise-free ones. This figure shows that the amplitudes are much more affected by the addition of large amounts of noise than the differential phases, but only slightly more affected for small amounts of noise.

Compared with previous results, the use of this noise-dependent normalization constant results in a slight improvement in classification error of about 3 to 4% for the AW feature at post-processing SNRs of 10 to 20 dB. Further research might provide a substantially better implementation of the AW feature.
Figure 7.1. Variation of normalization constant, derived from the variances of the noise contaminated amplitudes ($A_n$) and differential phases ($W_n$), with post-processing SNR. The curves are representative of 2 and 8 frequencies using the AW feature in the NN algorithm, for ships at 0°, 90° and 180° aspect angles and 27° elevation angle.
REFERENCES


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APPENDIX A

CLASSIFICATION RESULTS

This appendix contains plots of misclassification percentage versus post-processing signal-to-noise-ratio for the experiments of Chapter VI. A guide to interpreting the headers of these curves is given in Chapter VI, Section 6.1.2.
CLASSIFICATION OF SHIPS
POLARIZATION V
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN
MIN, MAX, INC ASPECT 0 10 10
NO OF FREQUENCIES 4 4 4
NO OF TARGETS 12 12 12
90% CI (±30%) +/- 3.1% 3.1% 3.1%
CLASS. FEATURES A A A

Figure A.1. Misclassification percentage versus post-processing SNR, comparing the performance of 3 sub-bands in the 2-7 MHz band.
Figure A.2. Misclassification percentage versus post-processing SNR, comparing the performance of 3 sub-bands in the 2-7 MHz band.
Figure A.3. Misclassification percentage versus post-processing SNR, comparing the performance of 3 sub-bands in the 2-7 MHz band.
CLASSIFICATION OF SHIPS

POLARIZATION: M
ELEVATION ASSUMED: KNOWN
ELEVATION (DEG.): 27
ASPECT ASSUMED: KNOWN
MIN, MAX, INC ASPECT: 0-180, 90
NO OF FREQUENCIES: 4
NO OF TARGETS: 18
90% CI (+30%): +/- 2.5% 2.5% 2.5%
CLASS. FEATURES: A A A

---

Figure A.4. Misclassification percentage versus post-processing SNR, comparing the performance of 3 sub-bands in the 2-7 MHz band.
CLASSIFICATION OF SHIPS
POLARIZATION X
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN
MIN, MAX, INC ASPECT 0 180 90
NO OF FREQUENCIES 4 4 4
NO OF TARGETS 18 18 18
90% CI (+30) +/- 2.5% 2.5% 2.5%
CLASS. FEATURES A A A

Figure A.5. Misclassification percentage versus post-processing SNR, comparing the performance of 3 sub-bands in the 2-7 MHz band.
Figure A.6. Misclassification percentage versus post-processing SNR, comparing the performance of 3 sub-bands in the 2-7 MHz band.
CLASSIFICATION OF SHIPS
POLARIZATION V
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN
MIN, MAX, INC ASPECT 0 180 90
NO OF FREQUENCIES 4 4 4
NO OF TARGETS 18 18 18
90% CI (+30%) +/- 2.5% 2.5% 2.5%
CLASS. FEATURES T T T

Figure A.7. Misclassification percentage versus post-processing SNR, comparing the performance of 3 sub-bands in the 2-7 MHz band.
Figure A.8. Misclassification percentage versus post-processing SNR, comparing the performance of 3 sub-bands in the 2-7 MHz band.
CLASSIFICATION OF SHIPS
POLARIZATION V
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN
MIN, MAX, INC ASPECT 0 180 90
NO OF FREQUENCIES 4 4 4
NO OF TARGETS 18 18 18
90% CI (1030%) +/- 2.5% 2.5% 2.5%
CLASS. FEATURES A&W A&W A&W

Figure A.9. Misclassification percentage versus post-processing SNR, comparing the performance of 3 sub-bands in the 2-7 MHz band.

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Figure A.10. Misclassification percentage versus post-processing SNR, comparing the performance of different numbers of frequencies.
CLASSIFICATION OF SHIPS
POLARIZATION V
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN
MIN, MAX, INC ASPECT 0 180 90
NO OF FREQUENCIES 2 4 8 12
NO OF TARGETS 18 18 18 18
90% CI (03%) +/- 2.5% 2.5% 2.5% 2.5%
CLASS. FEATURES A&W A&W A&W A&W

Figure A.11. Misclassification percentage versus post-processing SNR, comparing the performance of different numbers of frequencies.
CLASSIFICATION OF SHIPS

POLARIZATION V
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN
MIN, MAX, INC ASPECT 0 180 90
NO OF FREQUENCIES 2 4 6 12
NO OF TARGETS 18 18 18 18
90% CI (5%2) +/- 2.5% 2.5% 2.5% 2.5%
CLASS. FEATURES A A A A

<table>
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<th>NUMBER OF FREQUENCIES</th>
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<th>4</th>
<th>6</th>
<th>12</th>
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<td>POST-PROCESSING S/N IN DB</td>
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<td></td>
<td></td>
</tr>
</tbody>
</table>

Figure A.12. Misclassification percentage versus post-processing SNR, comparing the performance of different numbers of frequencies.
Figure A.13. Misclassification percentage versus post-processing SNR, comparing the performance of different numbers of frequencies.
Figure A.14. Misclassification percentage versus post-processing SNR, comparing the performance of different algorithms.
CLASSIFICATION OF SHIPS
POLARIZATION V
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN
MIN, MAX, INC ASPECT 0 180 90
NO OF FREQUENCIES 4
NO OF TARGETS 18 18 18 18 18 18 18
90% CI (±30%) ±/- 2.5% 2.5% 2.5% 2.5% 2.5% 2.5%
CLASS. FEATURES A W R4W R4W R T

Figure A.15. Misclassification percentage versus post-processing SNR, comparing the performance of different algorithms.
Figure A.16. Misclassification percentage versus post-processing SNR, comparing the performance of different algorithms.
Figure A.17. Misclassification percentage versus post-processing SNR, comparing the performance of known and unknown aspect angles.
Figure A.18. Misclassification percentage versus post-processing SNR, comparing the performance of known and unknown aspect angles.
Figure A.19. Misclassification percentage versus post-processing SNR, comparing the performance of known and unknown aspect angles.
CLASSIFICATION OF SHIPS

POLARIZATION V  
ELEV ASSUMED KNOWN  
ELEVATION (DEG.) 27  
ASPECT ASSUMED KNOWN / UNKNOWN  
MIN, MAX, INC ASPECT 0 180 90  
NO OF FREQUENCIES 8  
NO OF TARGETS 18 18  
90% CI (+30%) +/- 2.5% 2.5%  
CLASS. FEATURES R&W R&W

Figure A.20. Misclassification percentage versus post-processing SNR, comparing the performance of known and unknown aspect angles.
CLASSIFICATION OF SHIPS
POLARIZATION V
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN / UNKNOWN
MIN, MAX, INC ASPECT 0 180 90
NO OF FREQUENCIES 8
NO OF TARGETS 18 18
90% CI (±30%) +/- 2.5% 2.5%
CLASS. FEATURES T T

Figure A.21. Misclassification percentage versus post-processing SNR, comparing the performance of known and unknown aspect angles.
Figure A.22. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
### Classification of Ships

<table>
<thead>
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<th>V</th>
<th>ELEV Assumed</th>
<th>Known / Unknown</th>
</tr>
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<td>Elevation (deg.)</td>
<td>15 27 15.27</td>
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<td></td>
<td></td>
</tr>
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<td></td>
</tr>
<tr>
<td>No of Targets</td>
<td>12 12 24 24</td>
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<td></td>
<td></td>
</tr>
<tr>
<td>90% CI (+/-30%)</td>
<td>3.1% 3.1% 2.2% 2.2% 2.2%</td>
<td></td>
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<tr>
<td>Class. Features</td>
<td>A A A A A A</td>
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<td></td>
<td></td>
</tr>
<tr>
<td>ELEV ERR IN CURVE</td>
<td>4</td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

---

**Figure A.23.** Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
Figure A.24. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
CLASSIFICATION OF SHIPS
POLARIZATION H
ELEV ASSUMED KNOWN / UNKNOWN
ELEVATION (DEG.) 15 27 15.27
ASPECT ASSUMED KNOWN
MIN, MAX, INC ASPECT 170 180 10
NO OF FREQUENCIES 8
NO OF TARGETS 12 12 24 24 24
90% CI (+30%) +/- 3.1% 3.1% 2.2% 2.2% 2.2%
CLASS. FEATURES A A A A A
ELEV ERR IN CURVE 4

Figure A.25. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
Figure A.26. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
Figure A.27. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
<table>
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<tr>
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</tr>
<tr>
<td>ELEVATION (DEG.): 15 27 15.27</td>
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<tr>
<td>ASPECT ASSUMED: KNOWN</td>
</tr>
<tr>
<td>MIN, MAX, INC ASPECT: 80 100 10</td>
</tr>
<tr>
<td>NO OF FREQUENCIES: 8</td>
</tr>
<tr>
<td>NO OF TARGETS: 18, 18, 36, 36, 36, 36</td>
</tr>
<tr>
<td>90% CI (±30%): 2.5%, 2.5%, 1.8%, 1.8%, 1.8%, 1.8%</td>
</tr>
<tr>
<td>CLASS. FEATURES: A A A A A</td>
</tr>
<tr>
<td>ELEV ERR IN CURVE: 4</td>
</tr>
</tbody>
</table>

Figure A.28. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
CLASSIFICATION OF SHIPS

Polarization V/H

Elev Assumed Known / Unknown

Elevation (deg.) 15 27 15.2°

Aspect Assumed Known

Min, Max, Inc Aspect 170 180 10

No of Frequencies 8

No of Targets 12 12 24 24 24

90% CI (±30%) +/− 3.1% 3.1% 2.2% 2.2% 2.2%

Class. Features A A A A A

Elev Err in Curve 4

Figure A.29. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
### Classification of Ships

#### Polarization
- V

#### Elevation Assumed
- KNOWN / UNKNOWN

#### Elevation (Deg.)
- 15°, 27°, 15.27°

#### Aspect Assumed
- KNOWN

#### Min, Max, Inc Aspect
- 0°, 10°, 10°

#### No of Frequencies
- 8

#### No of Targets
- 12, 12, 24, 24, 24

#### 90% CI (%30%) +/-
- 3.1%, 3.1%, 2.2%, 2.2%, 2.27%

#### Class Features

#### Elevation Error in Curve
- 4

---

**Figure A.30.** Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
CLASSIFICATION OF SHIPS
POLARIZATION V
ELEV ASSUMED KNOWN / UNKNOWN
ELEVATION (DEG.) 15 27 15.27
ASPECT ASSUMED KNOWN
MIN, MAX, INC ASPECT 80 100 10
NO OF FREQUENCIES 8
NO OF TARGETS 18 18 36 36 36
90% CI [30%] +/- 2.5% 2.5% 1.8% 1.8% 1.8%
CLASS. FEATURES A&W A&W A&W A&W A&W
ELEV ERR IN CURVE 4

Figure A.31. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
Figure A.32. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
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<tr>
<td>Elevation (deg.) 15 27 15 27</td>
</tr>
<tr>
<td>Aspect Assumed Known</td>
</tr>
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<td>Min, Max, Inc Aspect 0 10 10</td>
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<td>No of Frequencies 8</td>
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<tr>
<td>No of Targets 12 12 24 24 24</td>
</tr>
<tr>
<td>90% CI (+/-) 3.1% 3.1% 2.2% 2.2% 2.2%</td>
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<tr>
<td>Class. Features A4W A4W A4W A4W A4W</td>
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<td>Elev Err in Curve 4</td>
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**Figure A.33.** Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
Figure A.34. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
Figure A.35. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
Figure A.36. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
CLASSIFICATION OF SHIPS
POLARIZATION X
ELEV ASSUMED KNOWN / UNKNOWN
ELEVATION (DEG.) 15 27 15,27
ASPECT ASSUMED KNOWN
MIN, MAX, INC ASPECT 80 100 10
NO OF FREQUENCIES 8
NO OF TARGETS 18 18 36 36 36
90% CI (±30%) +/- 2.5% 2.5% 1.8% 1.8% 1.8%
CLASS. FEATURES A4W A4W A4W A4W A4W
ELEV ERR IN CURVE 4

Figure A.37. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
Figure A.38. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
### Classification of Ships

**Polarization:** V/H  
**Elev Assumed:** Known / Unknown  
**Elevation (deg.):** 15 27 15,27  
**Aspect Assumed:** Known  
**Min, Max, Inc Aspect:** 0 10 10  
**No of Frequencies:** 6  
**No of Targets:** 12 12 24 24 24  
**90% CI (±30%)** +/− 3.1% 3.1% 2.2% 2.2% 2.2%  
**Class. Features:** A&W A&W A&W A&W A&W  
**Elev Err in Curve:**  

<table>
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</table>

#### Figure A.39

Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
CLASSIFICATION OF SHIPS
POLARIZATION V/H
ELEV ASSUMED KNOWN / UNKNOWN
ELEVATION (DEG.) 15 27 15.27
ASPECT ASSUMED KNOWN
MIN, MAX, INC ASPECT 80 100 10
NO OF FREQUENCIES 8
NO OF TARGETS 18 18 36 36 36
90% CI (+30%) +/- 2.5% 2.5% 1.8% 1.8% 1.8%
CLASS. FEATURES A&W A&W A&W A&W A&W
ELEV ERR IN CURVE

<table>
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<th>15 DEG. ELEVATION</th>
<th>27 DEG. ELEVATION</th>
<th>15 AND 27 DEG. ELEVATION (KNOWN)</th>
<th>15 AND 27 DEG. ELEVATION (WITH ERROR)</th>
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<tr>
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<tr>
<td>-30.</td>
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<td></td>
<td></td>
</tr>
<tr>
<td>-20.</td>
<td></td>
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<td></td>
<td></td>
</tr>
<tr>
<td>-10.</td>
<td></td>
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<td></td>
</tr>
<tr>
<td>0.</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>10.</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>20.</td>
<td></td>
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<td></td>
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<td></td>
</tr>
<tr>
<td>30.</td>
<td></td>
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</tr>
</tbody>
</table>

Figure A.40. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
CLASSIFICATION OF SHIPS
POLARIZATION  V/H
ELEV ASSUMED  KNOWN / UNKNOWN
ELEVATION (DEG.)  15  27  15,27
ASPECT ASSUMED  KNOWN
MIN,MAX,INC ASPECT  170  180  10
NO OF FREQUENCIES  8
NO OF TARGETS  12  12  24  24  24
90% CI (90%) +/-  3.1%  3.1%  2.2%  2.2%  2.2%
CLASS. FEATURES  A4W  A4W  A4W  A4W  A4W
ELEV ERR IN CURVE  4

![Graph showing Misclassification percentage versus post-processing SNR]

**Figure A.41.** Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
Figure A.42. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
Figure A.43. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
Figure A.44. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
Figure A.45. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
CLASSIFICATION OF SHIPS
POLARIZATION H
ELEV ASSUMED KNOWN / UNKNOWN
ELEVATION (DEG.) 15 27 15.27
ASPECT ASSUMED KNOWN
MIN, MAX, INC ASPECT 80 100 10
NO OF FREQUENCIES 8
NO OF TARGETS 18 18 36 36 36
90% CI (+30%) +/- 2.5% 2.5% 1.8% 1.8% 1.8%
CLASS. FEATURES A R R A R
ELEV ERR IN CURVE 4

Figure A.46. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
CLASSIFICATION OF SHIPS
POLARIZATION H
ELEV ASSUMED KNOWN / UNKNOWN
ELEVATION (DEG.) 15 27 15,27
ASPECT ASSUMED KNOWN
MIN,MAX,INC ASPECT 170 180 10
NO OF FREQUENCIES 8
NO OF TARGETS 12 12 24 24 24
90% CI (±30%) +/- 3.1% 3.1% 2.2% 2.2% 2.2%
CLASS. FEATURES R R R R R
ELEV ERR IN CURVE 4

Figure A.47. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
Figure A.48. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
CLASSIFICATION OF SHIPS
POLARIZATION \( x \)
ELEV ASSUMED KNOWN / UNKNOWN
ELEVATION (DEG.) 15 27 15.27
ASPECT ASSUMED KNOWN
MIN, MAX, INC ASPECT 80 100 10
NO OF FREQUENCIES 8
NO OF TARGETS 18 18 36 36 36
90% CI (+30%) +/- 2.5% 2.5% 1.8% 1.8% 1.8%
CLASS. FEATURES A A A A A
ELEV ERR IN CURVE 4

---

![Graph](image)

Figure A.49. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
Figure A.50. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
Figure A.51. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
Figure A.52. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
Figure A.53. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
Figure A.54. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
| Classification of Ships | Polarization | V | Elevation Assumed | Known / Unknown | Elevation (Deg.) | 15 27 15,27 | Aspect Assumed | Known | Min, Max, Inc Aspect | 80 100 10 | No of Frequencies | 8 | No of Targets | 18 18 36 36 36 | 90% CI (±30%) | +/- 2.5% 2.5% 1.6% 1.6% 1.6% | Class. Features | T T T T | Elev Err in Curve | 4 |

**Figure A.55.** Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
CLASSIFICATION OF SHIPS

POLARIZATION  V
ELEV ASSUMED  KNOWN / UNKNOWN
ELEVATION (DEG.)  15  27  15,27
ASPECT ASSUMED  KNOWN
MIN, MAX, INC ASPECT 170 180 10
NO OF FREQUENCIES  8
NO OF TARGETS  12  12  24  24  24
90% CI (±30%) +/− 3.1% 3.1% 2.2% 2.2% 2.2%
NO OF FEATURES  T T T T T
ELEV ERR IN CURVE  4

Figure A.56. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
CLASSIFICATION OF SHIPS
POLARIZATION H
ELEV ASSUMED KNOWN / UNKNOWN
ELEVATION (DEG.) 15 27 15,27
ASPECT ASSUMED KNOWN
MIN, MAX, INC ASPECT 0 10 10
NO OF FREQUENCIES 8
NO OF TARGETS 12 12 24 24
90% CI (30%) +/- 3.1% 3.1% 2.2% 2.2%
CLASS. FEATURES T T T T T
ELEV ERR IN CURVE 4

Figure A.57. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
### Classification of Ships

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<td>Known</td>
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<td>No of Targets</td>
<td>18, 18, 36, 36, 36</td>
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<td>90% CI (±30%)</td>
<td>±2.5%, ±2.5%, ±1.8%, ±1.8%, ±1.8%</td>
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<td>Class. Features</td>
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<td>Elev Err in Curve</td>
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#### Figure A.58.
Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
Figure A.59. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
Figure A.60. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
CLASSIFICATION OF SHIPS
POLARIZATION X
ELEV ASSUMED KNOWN / UNKNOWN
ELEVATION (deg.) 15 27 15,27
ASPECT ASSUMED KNOWN
MIN, MAX, INC ASPECT 80 100 10
NO OF FREQUENCIES 8
NO OF TARGETS 18 18 36 36 36
90% CI (90%) +/- 2.5% 2.5% 1.8% 1.8% 1.8%
CLASS. FEATURES T T T T T
ELEV ERR IN CURVE 4

Figure A.61. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
### Classification of Ships

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<td>Min, Max, Inc Aspect</td>
<td>170, 180, 10</td>
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<tr>
<td>No of Frequencies</td>
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<tr>
<td>No of Targets</td>
<td>12, 12, 24, 24, 24</td>
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<tr>
<td>90% CI (±3σ% )</td>
<td>+/− 3.1%, 3.1%, 2.2%, 2.2%, 2.2%</td>
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<td>Class. Features</td>
<td>T, T, T, T, T</td>
</tr>
<tr>
<td>ELEV ERR in Curve</td>
<td>4</td>
</tr>
</tbody>
</table>

#### Figure A.62
Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
Figure A.63. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
Figure A.64. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
Figure A.65. Misclassification percentage versus post-processing SNR, comparing the performance of various elevation angles.
Figure A.66. Misclassification percentage versus post-processing SNR, comparing the performance of various polarizations.
Figure A.67. Misclassification percentage versus post-processing SNR, comparing the performance of various polarizations.
CLASSIFICATION OF SHIPS

POLARIZATION  V  H  X  V/H
ELEV ASSUMED  KNOWN
ELEVATION (DEG.)  27
ASPECT ASSUMED  KNOWN
MIN,MAX,INC ASPECT  0 10 10
NO OF FREQUENCIES  8
NO OF TARGETS  12  12  12  12
90% CI (e30%) +/-  3.1%  3.1%  3.1%  3.1%
CLASS. FEATURES  W  W  W  W

---

Figure A.68. Misclassification percentage versus post-processing SNR, comparing the performance of various polarizations.
Figure A.69. Misclassification percentage versus post-processing SNR, comparing the performance of various polarizations.
Figure A.70. Misclassification percentage versus post-processing SNR, comparing the performance of various polarizations.
CLASSIFICATION OF SHIPS

POLARIZATION V H X V/H

ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN
MIN, MAX, INC ASPECT 0 10 10
NO OF FREQUENCIES 8
NO OF TARGETS 12 12 12 12
90% CI (±30%) +/- 3.1% 3.1% 3.1% 3.1%

CLASS. FEATURES A4W A4W A4W A4W

Figure A.71. Misclassification percentage versus post-processing SNR, comparing the performance of various polarizations.
CLASSIFICATION OF SHIPS

POLARIZATION V H X V/H
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN
MIN.MAX,INC ASPECT 80 100 10
NO OF FREQUENCIES 8
NO OF TARGETS 18 18 18 18
90% CI (95%) +/- 2.5% 2.5% 2.5% 2.5%
CLASS. FEATURES A/W A/W A/W A/W

---

Figure A.72. Misclassification percentage versus post-processing SNR, comparing the performance of various polarizations.
CLASSIFICATION OF SHIPS
POLARIZATION  V  H  X  V/H
ELEV ASSUMED  KNOWN
ELEVATION (DEG.)  27
ASPECT ASSUMED  KNOWN
MIN, MAX, INC ASPECT  170 180 10
NO OF FREQUENCIES  8
NO OF TARGETS  12  12  12  12
90% CI (±30%)  3.1%  3.1%  3.1%  3.1%
CLASS. FEATURES  A&W  A&W  A&W  A&W

Figure A.73. Misclassification percentage versus post-processing SNR, comparing the performance of various polarizations.
Figure A.74. Misclassification percentage versus post-processing SNR, comparing the performance of various polarizations.
Figure A.75. Misclassification percentage versus post-processing SNR, comparing the performance of various polarizations.
CLASSIFICATION OF SHIPS
POLARIZATION V H X V/H
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN
MIN, MAX, INC ASPECT 170 180 10
NO OF FREQUENCIES 8
NO OF TARGETS 12 12 12 12
90% CI (+30%) +/- 3.1% 3.1% 3.1% 3.1%
CLASS. FEATURES R R R R

![Graph](image)

Figure A.76. Misclassification percentage versus post-processing SNR, comparing the performance of various polarizations.
Figure A.77. Misclassification percentage versus post-processing SNR, comparing the performance of various polarizations.
Figure A.78. Misclassification percentage versus post-processing SNR, comparing the performance of various polarizations.
Figure A.79. Misclassification percentage versus post-processing SNR, comparing the performance of various polarizations.
Figure A.80. Misclassification percentage versus post-processing SNR, comparing the performance of various polarizations.
CLASSIFICATION OF SHIPS
POLARIZATION V H X V/H
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN
MIN, MAX, INC ASPECT 80 100 10
NO OF FREQUENCIES 8
NO OF TARGETS 18 18 18 18
90% CI (+30%) +/- 2.5% 2.5% 2.5% 2.5%
CLASS. FEATURES T T T T

Figure A.81. Misclassification percentage versus post-processing SNR, comparing the performance of various polarizations.
CLASSIFICATION OF SHIPS
POLARIZATION V H X V/H
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN
MIN,MAX,INC ASPECT 170 180 10
NO OF FREQUENCIES 8
NO OF TARGETS 12 12 12 12
90% CI (±30%) +/− 3.1% 3.1% 3.1% 3.1%
CLASS. FEATURES T T T T

Verticle Polarization
Horizontal Polarization
Cross Polarization
V/H Polarization

Figure A.82. Misclassification percentage versus post-processing SNR, comparing the performance of various polarizations.
Figure A.83. Misclassification percentage versus post-processing SNR, comparing the performance of various aspect zones.
Figure A.84. Misclassification percentage versus post-processing SNR, comparing the performance of various aspect zones.
CLASSIFICATION OF SHIPS
POLARIZATION V
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN / KNOWN
MIN, MAX, INC ASPECT 0 10 10, 30 60 15, 80 100 10, 170 180 10
NO OF FREQUENCIES 8
NO OF TARGETS 12 18 18 12
90% CI (±30%) +/- 3.1% 2.5% 2.5% 3.1%
CLASS. FEATURES A A A

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Figure A.85. Misclassification percentage versus post-processing SNR, comparing the performance of various aspect zones.
Figure A.86. Misclassification percentage versus post-processing SNR, comparing the performance of various aspect zones.
Figure A.87. Misclassification percentage versus post-processing SNR, comparing the performance of various aspect zones.
CLASSIFICATION OF SHIPS
POLARIZATION  H
ELEV ASSUMED  KNOWN
ELEVATION (DEG.)  27
ASPECT ASSUMED  KNOWN / KNOWN
MIN, MAX, INC ASPECT  0 10 10, 80 100 10, 170 180 10
NO OF FREQUENCIES  8
NO OF TARGETS  12 18 12
90% CI (±30%)  +/- 3.1% 2.5% 3.1%
CLASS. FEATURES  A A A

Figure A.88. Misclassification percentage versus post-processing SNR, comparing the performance of various aspect zones.
CLASSIFICATION OF SHIPS

POLARIZATION H

ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27

ASPECT ASSUMED KNOWN / KNOWN

MIN, MAX, INC ASPECT 0 10 10, 80 100 10, 170 180 10

NO OF FREQUENCIES 6

NO OF TARGETS 12 18 12

90% CI (±30%) ± 3.1% 2.5% 3.1%

CLASS. FEATURES W W W

Figure A.89. Misclassification percentage versus post-processing SNR, comparing the performance of various aspect zones.
Classification of Ships

Polarization: H

Elevation Assumed: Known

Elevation (deg.): 27

Aspect Assumed: Known / Known

Min., Max., Inc. Aspect: 0 10 10, 80 100 10, 170 180 10

No. of Frequencies: 8

No. of Targets: 12 18 12

90% CI (+30%) ± 3.1% 2.5% 3.1%

Class. Features: A4W A4W A4W

Figure A.90. Misclassification percentage versus post-processing SNR, comparing the performance of various aspect zones.

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Figure A.91. Misclassification percentage versus post-processing SNR, comparing the performance of various aspect zones.
Figure A.92. Misclassification percentage versus post-processing SNR, comparing the performance of various aspect zones.
CLASSIFICATION OF SHIPS
POLARIZATION  H
ELEV ASSUMED  KNOWN
ELEVATION (DEG.)  27
ASPECT ASSUMED  KNOWN / KNOWN
MIN,MAX,INC ASPECT  0 10 10, 80 100 10, 170 180 10
NO OF FREQUENCIES  8
NO OF TARGETS  12 18 12
90% CI (+30%) +/- 3.1% 2.5% 3.1%
CLASS. FEATURES  T T T

Figure A.93. Misclassification percentage versus post-processing SNR, comparing the performance of various aspect zones.
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![Graph](image)

Figure A.94. Misclassification percentage versus post-processing SNR, comparing the performance of various aspect zones.
### Classification of Ships

**Polarization:** X  
**Elevation assumed known**  
**Elevation (deg.)** 27  
**Aspect assumed known/known**  
**Min, Max, Inc Aspect 0-10-10-80-100-10-170-180-10**  
**No of Frequencies:** 8  
**No of Targets:** 12 18 12  
**90% CI (+90%) +/- 3.1% 2.5% 3.1%**  
**Class. Features W W W**

---

![Graph showing misclassification percentage versus post-processing SNR, comparing the performance of various aspect zones.](image)

**Figure A.95.** Misclassification percentage versus post-processing SNR, comparing the performance of various aspect zones.
CLASSIFICATION OF SHIPS
POLARIZATION X
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN / KNOWN
MIN, MAX, INC ASPECT 0 10 10, 60 100 10, 170 180 10
NO OF FREQUENCIES 8
NO OF TARGETS 12 18 12
90% CI (+30%) +/- 3.1% 2.5% 3.1%
CLASS. FEATURES A4W A4W A4W

Figure A.96. Misclassification percentage versus post-processing SNR, comparing the performance of various aspect zones.
CLASSIFICATION OF SHIPS
POLARIZATION X
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN / KNOWN
MIN, MAX, INC ASPECT 0 10 10, 80 100 10, 170 180 10
NO OF FREQUENCIES 8
NO OF TARGETS 12 18 12
90% CI (±30%) +/- 3.1% 2.5% 3.1%
CLASS. FEATURES R R R

Figure A.97. Misclassification percentage versus post-processing SNR, comparing the performance of various aspect zones.
CLASSIFICATION OF SHIPS
POLARIZATION X
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN / KNOWN
MIN, MAX, INC ASPECT 0 10 10, 80 100 10, 170 180 10
NO OF FREQUENCIES 8
NO OF TARGETS 12 18 12
90% CI (90%) +/- 3.1% 2.5% 3.1%
CLASS. FEATURES R&W R&W R&W

Figure A.98. Misclassification percentage versus post-processing SNR, comparing the performance of various aspect zones.
Classification of Ships

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<td>NO OF TARGETS</td>
<td>12, 18, 12</td>
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<td>90% CI (+/- 30%)</td>
<td>3.1%, 2.5%, 3.1%</td>
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Figure A.99. Misclassification percentage versus post-processing SNR, comparing the performance of various aspect zones.
Figure A.100. Misclassification percentage versus post-processing SNR, comparing the performance of various aspect zones.
CLASSIFICATION OF SHIPS
POLARIZATION V/H
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN / KNOWN
MIN,MAX,INC ASPECT 0 10 10, 80 100 10, 170 180 10
NO OF FREQUENCIES 8
NO OF TARGETS 12 18 12
90% CI (±30%) +/- 3.1% 2.5% 3.1%
CLASS. FEATURES A&W A&W A&W

Figure A.101. Misclassification percentage versus post-processing SNR, comparing the performance of various aspect zones.
Figure A.102. Misclassification percentage versus post-processing SNR, comparing the performance of various aspect zones.
CLASSIFICATION OF SHIPS
POLARIZATION V/H
ELEV ASSUMED KNOWN
ELEV (DEG.) 27
ASPECT ASSUMED KNOWN / KNOWN
MIN, MAX, INC ASPECT 0 10 10, 60 100 10, 170 180 10
NO OF FREQUENCIES 8
NO OF TARGETS 12 18 12
90% CI (±30%) ±3.1% 2.5% 3.1%
CLASS. FEATURES RAW RAW RAW

Figure A.103. Misclassification percentage versus post-processing SNR, comparing the performance of various aspect zones.
CLASSIFICATION OF SHIPS
POLARIZATION V/H
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN / KNOWN
MIN, MAX, INC ASPECT 0 10 10, 60 100 10, 170 180 10
NO OF FREQUENCIES 8
NO OF TARGETS 12 18 12
90% CI (+30%) +/− 3.1% 2.5% 3.1%
CLASS. FEATURES T T T

Figure A.104. Misclassification percentage versus post-processing SNR, comparing the performance of various aspect zones.
Figure A.105. Misclassification percentage versus post-processing SNR, comparing performance with various aspect angle errors, near to 0° aspect.
CLASSIFICATION OF SHIPS
POLARIZATION  V/H
ELEV ASSUMED  KNOWN
ELEVATION (DEG.)  27
ASPECT ASSUMED  KNOWN / KNOWN
MIN,MAX,INC ASPECT  0 10 10, 0 15 15, 0 30 30
NO OF FREQUENCIES  8
NO OF TARGETS  12 12 12
90% CI (±30%)  +/- 3.1%  3.1%  3.1%
CLASS. FEATURES  A A A
ASP ERR IN CURVE  2 3

Figure A.106. Misclassification percentage versus post-processing SNR, comparing performance with various aspect angle errors, near to 0° aspect.
CLASSIFICATION OF SHIPS
POLARIZATION V
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN / KNOWN
MIN, MAX, INC ASPECT 0 10 10. 0 15 15. 0 30 30
NO OF FREQUENCIES 8
NO. TARGETS 12 12 12
90% CI (+30%) +/- 3.1% 3.1% 3.1%
CLASS. FEATURES A&W A&W A&W
ASP ERR IN CURVE 2 3

Figure A.107. Misclassification percentage versus post-processing SNR, comparing performance with various aspect angle errors, near to 0° aspect.
CLASSIFICATION OF SHIPS
POLARIZATION H
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN / KNOWN
MIN, MAX, INC ASPECT 0 10 10, 0 15 15, 0 30 30
NO OF FREQUENCIES 8
NO OF TARGETS 12 12 12
90% CI (±30%) +/- 3.1% 3.1% 3.1%
CLASS. FEATURES A&W A&W A&W
ASP ERR IN CURVE 2 3

Figure A.108. Misclassification percentage versus post-processing SNR, comparing performance with various aspect angle errors, near to 0° aspect.
CLASSIFICATION OF SHIPS
POLARIZATION: X
ELEV ASSUMED KNOWN
ELEVATION (DEG.): 27
ASPECT ASSUMED KNOWN / KNOWN
MIN, MAX, INC ASPECT: 0 10 10, 0 15 15, 0 30 30
NO OF FREQUENCIES: 8
NO OF TARGETS: 12 12 12
90% CI (+30%) -/ 3.1% 3.1% 3.1%
CLASS. FEATURES: A&H A&H A&H
ASP ERR IN CURVE: 2 3

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Figure A.109. Misclassification percentage versus post-processing SNR, comparing performance with various aspect angle errors, near to 0° aspect.
CLASSIFICATION OF SHIPS
POLARIZATION V/H
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN / KNOWN
MIN, MAX, INC ASPECT 0 10, 0 15, 0 30 30
NO OF FREQUENCIES 8
NO OF TARGETS 12 12 12
90% CI (+30%) +/- 3.1% 3.1% 3.1%
CLASS. FEATURES A&W A&W A&W
ASP ERR IN CURVE 2 3

Figure A.110. Misclassification percentage versus post-processing SNR, comparing performance with various aspect angle errors, near to 0° aspect.
CLASSIFICATION OF SHIPS
POLARIZATION V
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN / KNOWN
MIN, MAX, INC ASPECT 0 10 10, 0 15 15, 0 30 30
NO OF FREQUENCIES 8
NO OF TARGETS 12 12 12
90% CI (+30%) +/- 3.1% 3.1% 3.1%
CLASS. FEATURES R&W R&W R&W
ASP ERR IN CURVE 2 3

Figure A.111. Misclassification percentage versus post-processing SNR, comparing performance with various aspect angle errors, near to 0° aspect.
CLASSIFICATION OF SHIPS
POLARIZATION \( H \)
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN / KNOWN
MIN,MAX,INC ASPECT 0 10 10, 0 15 15, 0 30 30
NO OF FREQUENCIES 8
NO OF TARGETS 12 12 12
90% CI (±30%) +/- 3.1% 3.1% 3.1%
CLASS. FEATURES R&W R&W R&W
ASP ERR IN CURVE 2 3

Figure A.112. Misclassification percentage versus post-processing SNR, comparing performance with various aspect angle errors, near to 0° aspect.
CLASSIFICATION OF SHIPS
POLARIZATION X
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN / KNOWN
MIN, MAX, INC ASPECT 0 10 10, 0 15 15, 0 30 30
NO OF FREQUENCIES 8
NO OF TARGETS 12 12 12
90% CI (3.1%) +/- 3.1% 3.1% 3.1%
CLASS. FEATURES R&W R&W R&W
ASP ERR IN CURVE 2 3

Figure A.113. Misclassification percentage versus post-processing SNR, comparing performance with various aspect angle errors, near to 0° aspect.
CLASSIFICATION OF SHIPS
POLARIZATION V/H
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN / KNOWN
MIN, MAX, INC ASPECT 0 10 10, 0 15 15, 0 30 30
NO OF FREQUENCIES 8
NO OF TARGETS 12 12 12
90% CI (+30%) +/- 3.1% 3.1% 3.1%
CLASS. FEATURES R&W R&W R&W
ASP ERR IN CURVE 2 3

Figure A.114. Misclassification percentage versus post-processing SNR, comparing performance with various aspect angle errors, near to 0° aspect.
CLASSIFICATION OF SHIPS
Polarization V
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN / KNOWN
MIN, MAX, INC ASPECT 0 10 10, 0 15 15, 0 30 30
NO OF FREQUENCIES 8
NO OF TARGETS 12 12 12
90% CI (+30%) +/- 3.1% 3.1% 3.1%
CLASS. FEATURES 1 1 1
ASP ERR IN CURVE 2 3

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Figure A.115. Misclassification percentage versus post-processing SNR, comparing performance with various aspect angle errors, near to 0° aspect.
CLASSIFICATION OF SHIPS
POLARIZATION H
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN / KNOWN
MIN, MAX, INC ASPECT 0 10 10, 0 15 15, 0 30 30
NO OF FREQUENCIES 8
NO OF TARGETS 12 12 12
90% CI (+30%) +/- 3.1% 3.1% 3.1%
CLASS. FEATURES T T T
ASP ERR IN CURVE 2 3

Figure A.116. Misclassification percentage versus post-processing SNR, comparing performance with various aspect angle errors, near to 0° aspect.

255
CLASSIFICATION OF SHIPS
POLARIZATION X
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN / KNOWN
MIN, MAX, INC ASPECT 0 10 10, 0 15 15, 0 30 30
NO OF FREQUENCIES 8
NO OF TARGETS 12 12 12
90% CI (+30%) +/- 3.1% 3.1% 3.1%
CLASS. FEATURES T T T
ASP ERR IN CURVE 2 3

Figure A.117. Misclassification percentage versus post-processing SNR, comparing performance with various aspect angle errors, near to 0° aspect.
Figure A.118. Misclassification percentage versus post-processing SNR, comparing performance with various aspect angle errors, near to 0° aspect.
Figure A.119. Misclassification percentage versus post-processing SNR, comparing performance with a ± 10° aspect error at various aspect zones.
CLASSIFICATION OF SHIPS
POLARIZATION X
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN / KNOWN
MIN, MAX, INC ASPECT 0 10 10, 0 10 10, 80 100 10, 170 160 10
NO OF FREQUENCIES 8
NO OF TARGETS 12 12 18 12
90% CI (+30%) +/- 3.1% 3.1% 2.5% 3.1%
CLASS. FEATURES A A A A
ASP ERR IN CURVE 2 3 4

Figure A.120. Misclassification percentage versus post-processing SNR, comparing performance with a ±10° aspect error at various aspect zones.

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CLASSIFICATION OF SHIPS
POLARIZATION V/H
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN / KNOWN
MIN, MAX, INC ASPECT 0 10 10, 0 10 10, 80 100 10, 170 180 10
NO OF FREQUENCIES 8
NO OF TARGETS 12 12 18 12
90% CI (±30%) +/- 3.1% 3.1% 2.5% 3.1%
CLASS. FEATURES A A A A
ASP ERR IN CURVE 2 3 4

Figure A.121. Misclassification percentage versus post-processing SNR, comparing performance with a ± 10° aspect error at various aspect zones.
CLASSIFICATION OF SHIPS

Polarization: V

ELEV Assumed: KNOWN

Elevation (Deg): 27

Aspect Assumed: KNOWN / KNOWN

Min, Max, Inc Aspect: 0, 10, 10, 0, 10, 10, 80, 100, 10, 170, 180, 180

No of Frequencies: 8

No of Targets: 12, 12, 18, 18

90% CI (3σ%) +/− 2.5%, 3.1%, 3.1%, 3.1%


Asp Err in Curve: 2, 3, 4

---

Figure A.122. Misclassification percentage versus post-processing SNR, comparing performance with a ±10° aspect error at various aspect zones.

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Figure A.123. Misclassification percentage versus post-processing SNR, comparing performance with a ± 10° aspect error at various aspect zones.
Figure A.124. Misclassification percentage versus post-processing SNR, comparing performance with a ±10° aspect error at various aspect zones.
CLASSIFICATION OF SHIPS
POLARIZATION V/H
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN / KNOWN
MIN, MAX, INC ASPECT 0 10 10, 0 10 10, 80 100 10, 170 180 10
NO OF FREQUENCIES 8
NO OF TARGETS 12 12 18 12
90% CI (+/−) 3.1% 3.1% 2.5% 3.1%
CLASS. FEATURES A4W A4W A4W A4W
ASP ERR IN CURVE 2 3 4

Figure A.125. Misclassification percentage versus post-processing SNR, comparing performance with a ± 10° aspect error at various aspect zones.
Classification of Ships

Polarization: V
Elev Assumed: Known
Elevation (deg.): 27
Aspect Assumed: Known
Min, Max, Inc. Aspect: 0 10 10, 0 10 10, 80 100 10, 170 180 10
No. of Frequencies: 8
No. of Targets: 12 12 18 12
90% CI (+30%): 3.1% 3.1% 2.5% 3.1%
Class. Features: R&W R&W R&W R&W

Aspect Error in Curve: 2 3 4

---

Figure A.126. Misclassification percentage versus post-processing SNR, comparing performance with a ± 10° aspect error at various aspect zones.

265
CLASSIFICATION OF SHIPS
POLARIZATION  H
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN / KNOWN
MIN, MAX, INC ASPECT 0 10 10, 0 10 10, 80 100 10, 170 180 10
NO OF FREQUENCIES 8
NO OF TARGETS 12 12 18 12
95% CI (+/- 3.1%) 3.1% 3.1% 2.5% 3.1%
CLASS. FEATURES R&W R&W R&W R&W
ASP ERR IN CURVE 2 3 4

Figure A.127. Misclassification percentage versus post-processing SNR, comparing performance with a ± 10° aspect error at various aspect zones.
Figure A.128. Misclassification percentage versus post-processing SNR, comparing performance with a ± 10° aspect error at various aspect zones.
CLASSIFICATION OF SHIPS
Polarization: V/H
Elevation Assumed: Known
Elevation (deg.): 27
Aspect Assumed: Known
Min, Max, Inc Aspect: 0 10 10, 0 10 10, 60 100 10, 170 180 10
No of Frequencies: 8
No of Targets: 12 12 18 18
90% CI (90%) +/- 3.1% 3.1% 2.5% 3.1%
Class. Features: R&W R&W R&W R&W
Aspect Error in Curve: 2 3 4

Figure A.129. Misclassification percentage versus post-processing SNR, comparing performance with a ±10° aspect error at various aspect zones.
CLASSIFICATION OF SHIPS
POLARIZATION V
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN / KNOWN
MIN, MAX, INC ASPECT 0 10 10, 0 10 10, 80 100 10, 170 180 10
NO OF FREQUENCIES 8
NO OF TARGETS 12 12 18 12
90% CI (±30%) +/- 3.1% 3.1% 2.5% 3.1%
CLASS. FEATURES T T T T
ASP ERR IN CURVE 2 3 4

Figure A.130. Misclassification percentage versus post-processing SNR, comparing performance with a ± 10° aspect error at various aspect zones.
Figure A.131. Misclassification percentage versus post-processing SNR, comparing performance with a ±10° aspect error at various aspect zones.
CLASSIFICATION OF SHIPS
POLARIZATION X
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN / KNOWN
MIN, MAX, INC ASPECT 0 10 10, 0 10 10, 60 100 10, 170 180 10
NO OF FREQUENCIES 8
NO OF TARGETS 12 12 10 10
90% CI (±30%) +/- 3.1% 3.1% 2.5% 3.1%
CLASS. FEATURES 1 T T T
ASP ERR IN CURVE 2 3 4

Figure A.132. Misclassification percentage versus post-processing SNR, comparing performance with a ±10° aspect error at various aspect zones.
CLASSIFICATION OF SHIPS
POLARIZATION V/H
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED KNOWN / KNOWN
MIN, MAX, INC ASPECT 0 10 10, 0 10 10, 80 100 10, 170 180 10
NO OF FREQUENCIES 8
NO OF TARGETS 12 12 18 12
90% CI (+30%) +/- 3.1% 3.1% 2.5% 3.1%
CLASS. FEATURES T T T T
ASP ERR IN CURVE 2 3 4

Figure A.133. Misclassification percentage versus post-processing SNR, comparing performance with a ± 10° aspect error at various aspect zones.
Figure A.134. Misclassification percentage versus post-processing SNR, comparing the performance of various algorithms using unknown aspect.
Figure A.135. Misclassification percentage versus post-processing SNR, comparing the performance of various algorithms using unknown aspect.
CLASSIFICATION OF SHIPS
POLARIZATION x
ELEV ASSUMED KNOWN
ELEVATION (DEG.) 27
ASPECT ASSUMED UNKNOWN
MIN, MAX, INC ASPECT 0 15 15
NO OF FREQUENCIES 8
NO OF TARGETS 12 12 12 12
90% CI (30%) +/- 3.1% 3.1% 3.1% 3.1%
CLASS. FEATURES A R&W R&W T

Figure A.136 Misclassification percentage versus post-processing SNR, comparing the performance of various algorithms using unknown aspect.

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Figure A.137. Misclassification percentage versus post-processing SNR, comparing the performance of various algorithms using unknown aspect.

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APPENDIX B

AMPLITUDE AND PHASE RETURNS

This appendix contains plots of the amplitudes and phases of processed radar returns for one ship, at 3 aspect angles, (0°, 90° and 180°), at 27° elevation angle, using vertical, horizontal and cross polarizations.
Figure B.1. Ship at 0 degree aspect, vertical polarization.
Figure B.2. Ship at 90 degree aspect, vertical polarization.
Figure B.3. Ship at 180 degree aspect, vertical polarization.
Figure B.4. Ship at 0 degree aspect, horizontal polarization.
Figure B.5. Ship at 90 degree aspect, horizontal polarization.
Figure B.6. Ship at 180 degree aspect, horizontal polarization.
Figure B.7. Ship at 0 degree aspect, horizontal polarization.
Figure B.8. Ship at 90 degree aspect, horizontal polarization.
Figure B.9. Ship at 180 degree aspect, horizontal polarization.
TABLE 8.1

AVERAGE AMPLITUDES FOR VARIOUS CATALOGS, CONTAINING SIX SHIPS

<table>
<thead>
<tr>
<th></th>
<th>0°</th>
<th>90°</th>
<th>180°</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>15°</td>
<td>27°</td>
<td>15°</td>
</tr>
<tr>
<td>V</td>
<td>46</td>
<td>40</td>
<td>50</td>
</tr>
<tr>
<td>H</td>
<td>29</td>
<td>25</td>
<td>23</td>
</tr>
<tr>
<td>X</td>
<td>29</td>
<td>28</td>
<td>30</td>
</tr>
<tr>
<td>V/H</td>
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<td>36</td>
<td>38</td>
</tr>
<tr>
<td>AVG</td>
<td>43</td>
<td>36</td>
<td>44</td>
</tr>
</tbody>
</table>

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APPENDIX C

PROGRAMS

The source codes, written in FORTRAN 7, developed for this research are listed in this appendix. Program names are CAL61.FOR, CATLOG.FOR and TAFCAL.FOR (Time and Frequency Classification Algorithm). All subroutines have been included, along with LINKING command procedures.

The command procedure FILESORT.COM is used to provide a directory file of file names to be used in conjunction with CATLOG.
PROGRAM NAME: CALS1
THIS PROGRAM CALIBRATES THE RADAR CROSS-SECTION DATA FROM 
MEASUREMENTS MADE ON THE COMPACT RADAR RANGE.
The input units are in db and degrees.

ORIGINALLY CALS0,
MODIFIED 4 APRIL 1984 BY N.F.CHAMBERLAIN TO PERFORM 
GATING IN THE TIME DOMAIN.

INCLUDE 'CALCOM.FOR'

Note CALCOM.FOR simply provides values for MD and other 
dimensioning variables.

COMMON /BLK1/BUFF /BLK2/MP,FMIN,FINC /BLK3/IB 
COMMON /BLK4/NDIM,ANST,AINC 
COMMON /BLK5/ARRA, NRPT 
COMMON /BLK6/PLUTOPT1,IPS1,LPL1,NDIG1,FXE,FXS,FYB,FYE,FYS 
COMMON /BLK7/PLUTOPT2,IPS2,LPL2,NDIG2,TXE,TXS,TYS,TYE,TYS 
INTEGER*2 TFILE(7),BFILE(7),CFILE(7),EFILE(7),RFILE(16) 
INTEGER*2 BCFILE(7),LINE2(30),INFILE(7) 
BYTE LINE1(60),LINES(60),BUFF(40),TLINE1(60),TLINES(60),STRING(22) 
INTEGER NSPRT(MD),NSPAG(MD) 
EQUIVALENCE(LINE1(1),BUFF(1)),(LINE1(1),BUFF(1)) 
(LINES(1),BUFF(121)) 
COMPLEX*8 TA(MD),BT(MD),CAS(MD),BC(MD),BEX(MD),R5(6000) 
COMPLEX*8 CMN(MD),TM8(MD) 
DIMENSION ARRA(30)

TPI=6.2831853
CHE=2.988E+08
CFT=9.830E+06
NP=MD

Initialize control variables.
ISUP=1 when SUP (set up command) is invoked.
KRPT is a counter which stops the repetition process
when NRPT repetitions are done.
NCUP indexes ARRA(I) and increments after each command.

KSTP=0
2020 NCUP=P
2020 KBP=1
2020 ISUP=0
2020 TYPE "," 
2020 TYPE "," "HELP" TO PRINT LIST OF COMMANDS' 
2020 IF(ISUP,ED.1) THEN 
2020 NCUP=NCUP+1 
2020 CALL DEUP(ARR,NCUP,KRPT,IEND,SPAG,KSTP) 1 SUP control subroutine 
2020 IF(IEND,ED.1) GOTO 267 
2020 GOTO 3803
2020 END IF

267 TYPE 288 
268 FORMAT( ',/', ' COMMAND: ' ,0)
ACCEPT 3862, ARR

3862 FORMAT(3)

1233 IF(ARR.EQ.'SUP') THEN
   CALL SUP(IUBP,ACFN,YDES,WTP,ALPHA,GWID1,GWID2,DEMOD,PAS)
   GOTO 2020
END IF

3863 IF(ARR.EQ.'FPL') TYPE *, 'TYPE "CHK" BEFORE ATTEMPTING FPL'
IF(ARR.EQ.'CAL') TYPE *, 'TYPE "CHK" BEFORE ATTEMPTING CAL'
IF(ARR.EQ.'WNT') TYPE *, 'DO CAL BEFORE WRITING FILE '
IF(ARR.EQ.'D WD') TYPE *, 'DO CAL BEFORE TIME DOMAIN GATE '
IF(ARR.EQ.'TPL') TYPE *, 'DO CAL BEFORE TIME DOMAIN PLOT '

1234 IF(ARR.EQ.'TAR') CALL TAR(TA,TFILE,TLINE1,TLINE3,TPT,&2020)
IF(ARR.EQ.'TXT') CALL TXT(TX,TFILE,TLINE1,TLINE3,TPT,&2020)
IF(ARR.EQ.'PHD') CALL PHD(PH,TFILE,TLINE1,TLINE3,TPT,&2020)
IF(ARR.EQ.'STR') CALL STR(ST,TFILE,TLINE1,TLINE3,TPT,&2020)
IF(ARR.EQ.'TAR') CALL TAR(TA,TFILE,TLINE1,TLINE3,TPT,&2020)
IF(ARR.EQ.'TXT') CALL TXT(TX,TFILE,TLINE1,TLINE3,TPT,&2020)
IF(ARR.EQ.'PHD') CALL PHD(PH,TFILE,TLINE1,TLINE3,TPT,&2020)
IF(ARR.EQ.'STR') CALL STR(ST,TFILE,TLINE1,TLINE3,TPT,&2020)
IF(ARR.EQ.'TAR') CALL TAR(TA,TFILE,TLINE1,TLINE3,TPT,&2020)
IF(ARR.EQ.'TXT') CALL TXT(TX,TFILE,TLINE1,TLINE3,TPT,&2020)
IF(ARR.EQ.'PHD') CALL PHD(PH,TFILE,TLINE1,TLINE3,TPT,&2020)
IF(ARR.EQ.'STR') CALL STR(ST,TFILE,TLINE1,TLINE3,TPT,&2020)
IF(ARR.EQ.'TAR') CALL TAR(TA,TFILE,TLINE1,TLINE3,TPT,&2020)
IF(ARR.EQ.'TXT') CALL TXT(TX,TFILE,TLINE1,TLINE3,TPT,&2020)
IF(ARR.EQ.'PHD') CALL PHD(PH,TFILE,TLINE1,TLINE3,TPT,&2020)
IF(ARR.EQ.'STR') CALL STR(ST,TFILE,TLINE1,TLINE3,TPT,&2020)
IF(ARR.EQ.'TAR') CALL TAR(TA,TFILE,TLINE1,TLINE3,TPT,&2020)
IF(ARR.EQ.'TXT') CALL TXT(TX,TFILE,TLINE1,TLINE3,TPT,&2020)
IF(ARR.EQ.'PHD') CALL PHD(PH,TFILE,TLINE1,TLINE3,TPT,&2020)
IF(ARR.EQ.'STR') CALL STR(ST,TFILE,TLINE1,TLINE3,TPT,&2020)

5782 IF(DWIN.NE.0.AND.FINC.NE.0) THEN
FIN=FINC*1.E09
NS=2.*CFT/(FIN*DWIN)
IF((NS/2.)*2.35, NS)=NS+1
END IF
ELSE
NS=0
END IF

ICHK enables the checking of parameters to be done in
another part of the program

ICHK=0
ICHK2=0

1019 TYPE *, ' ' TYPE *, 'LIST OF PARAMETERS'
TYPE *, 'NAME OF TARGET
TYPE *, 'NAME OF BACKGROUND

3473 FORMAT(1X, 'TARGET DISTANCE IN METERS---->', ',F5.2')
TYPE 3473, DWIN

3474 FORMAT(1X, 'WINDOW WIDTH IN FEET---------------->', ',F5.2')
TYPE 3472, NS

3472 FORMAT(1X, 'NO. OF POINTS OF SMOOTHING------------>', ',I4')
TYPE *, ' ' TYPE *, 'PRESENT INPUT DATA FILES'

3467 FORMAT(1X, 'TAR',TARGET---------------->', ',7A2')
TYPE 3466, TFILE

3468 FORMAT(1X, 'BTA',TARGET BACKGROUND------------>', ',7A2')
TYPE 3469, CFILE

3469 FORMAT(1X, 'PHD',CALIBRATION TARGET------------>', ',7A2')
TYPE 3470, BFILE
C
3470 FORMAT(1X,'"BSP",CALIBRATION TARGET BACKGROUND-> ','7A2)
TYPE 3471, EFILE
3471 FORMAT(1X,'"EXA",EXACT CALIBRATION TARGET-------> ','7A2)
IF(I0KH.EQ.1) GOTO 1029
IF(I0KH.EQ.1) GOTO 3333
C
IF(ISUP.EQ.1) THEN
NCSUP=NCSUP+1
CALL CSUP(NRSUP,NCUP,KRPT,IEND,BPAG,KSTP)
IF(IEND.EQ.1) GOTO 1022
GOTO 1029
END IF
C
1022 TYPE 288
ACCEPT 3862, ARR
1029 IF(ARR.EQ.'CAL') GOTO 1032
IF(ARR.EQ.'HLP') THEN
CALL HLP
GOTO 1022
END IF
IF(ARR.EQ.'OHK') THEN
I0KH2=1
GOTO 1019
3333 I0KH2=0
GOTO 1022
END IF
IF(ARR.NE.'WRT'.AND.ARR.NE.'TDW'.AND.ARR.NE.'FPL') GOTO 1234
IF(ARR.NE.'TPL') GOTO 1234
TYPE *, 'DO CAL BEFORE WRT,FPL,TPL OR TDW'
   ! Error checking
GOTO 1022
C
1032 IF(ISUP.NE.1) THEN
TYPE *, 'TYPE Y OR N FOR AUTOMATIC CALIBRATED FILE NAMER'
TYPE *, 'FILE WILL BE NAMED *****.CAL'
ACCEPT 1122, ACFN
1122 FORMAT(A1)
END IF
C
Note auto file namer can take directory prefixes and suffixes
such that the total file name is 28 characters long.
C
IF(ACFN.EQ.'W') THEN
   DO INDG=1,15
      RFILF(INDG)='.
   END DO
   TYPE *, '
   TYPE 52
52 FORMAT(1X,/') INPUT A FILE NAME FOR THE RESULT')
ACCEPT 120, RFILF
   DO 10001 INDG=15,1,-1
      IF (RFILF(INDG).EQ. ' ') THEN
         GOTO 10001
      ELSE
         GOTO 10001
IP3=INOS+1
GOTO 10002
END IF
10001 CONTINUE
10002 IP31=IP3
IP32=IP3
ELSE
C
Automatic calibrated file name
C
IP1=0
IP2=0
DO 171 I=1,7
IF(TFILE(I).AND. 'FFOO'X).EQ. '2ED0'X) IP2=I
IF(TFILE(I).AND. '00FF'X).EQ. 'D2EE'X) IP1=I
171 CONTINUE
IF(IP1.NE.0) THEN
DO 172 I=1,IP1-1
RFILE(I)=TFILE(I)
172 CONTINUE
RFILE(IP1)= 'C
RFILE(IP1+1)= 'AL
END IF
IF(IP2.NE.0) THEN
DO 173 I=1,IP2
RFILE(I)=TFILE(I)
173 CONTINUE
RFILE(IP2+1)= 'CA
RFILE(IP2+2)= 'L
END IF
IP3=0
IF(IP1+1.LT.15.AND.IP1.NE.0) IP3=IP1+2
IF(IP2+2.LT.15.AND.IP2.NE.0) IP3=IP2+3
IF(IP3.NE.0) THEN
DO 174 I=IP3,15
RFILE(IP3)=
174 CONTINUE
IF(IP3.EQ.0) IP3=15
IP31=IP3
IP32=IP3
END IF
C
TYPE *,
TYPE 118,RFILE
END IF
118 FORMAT(1X,'CALIBRATED FILENAME ',15A2)
120 FORMAT(15A2)
C
NPMIN=MIN(NPT, MPBT, NPC, MPBC, NPE)
IF(NPMIN.LT.1) NPMIN=NP
NP=NPMIN
C
Remove background and target distance
C
711 DO 261 I=1,NPMIN

292
IF (CAS(I).EQ.(0.,0.) .AND. BC(I).EQ.(0.,0.)) TMB(I)=TMB(I-1)
ELSE IF (CABS(CMB(I)).EQ.0.) CMB(I)=CMB(I-1)
END IF

C NREMOV=0
IF (ISUP.NE.1) THEN
TYPE *. '
TYPE *, 'REMOVE INVALID POINTS BEFORE CALIBRATION ? Y OR N'
ACCEPT 844,YDES
END IF

FORMAT(A11)
IF (YDES.EQ. 'Y') THEN
TYPE *, '
TYPE *, 'POINT REMOVAL FROM TARGET-BACKGROUND TERM'
NRM=0
IF (TMB(I).NE.(1.,0.)) CALL CREMVE(TMB, NPHIN, NRM, NSPKT, FINC)
TYPE *,'
TYPE *, 'POINT REMOVAL FROM CALIBRATION TARGET-BACKGROUND TERM'
NRM=0
IF (CMB(I).NE.(1.,0.)) CALL CREMVE(CMB, NPHIN, NRM, NSPKC, FINC)
NREMOV=NM-T NRM
C TYPE *, 'DO YOU WANT TO RE-DO THE POINT REMOVAL Y OR N'
ACCEPT 844,YDES
C IF(YDES.EQ. 'Y') GO TO 711
END IF
C
TYPE *,'
TYPE *, 'CALIBRATING DATA'
IF (WIN.EQ.0.) GO TO 1314
IF (CMB(I).NE.(1.,0.)) CALL HAMM(CMB, NPHIN, NS)
IF (TMB(I).NE.(1.,0.)) CALL HAMM(TMB, NPHIN, NS)
1314 DO 292 J=NPHN
IF (CABS(EXS(I)).EQ.0.) EXS(I)=(1.,0.)
RS(I)=TMB(I)/CMB(I)*EXS(I)
292 CONTINUE
C
C FIND THE LOCATION OF THE PERIOD IN THE INPUT FILE NAMES
C
DO 113 I=1,8
IF (((FILE(I).AND. 'FFOO'X).EQ. '2E00'X) .JTDO=I
IF (((FILE(I).AND. 'DDFF'X).EQ. '002E'X) .JTDO=I
IF (((FILE(I).AND. 'FFOO'X).EQ. '2E00'X) .JTDO=I
IF (((FILE(I).AND. 'DDFF'X).EQ. '002E'X) .JTDO=I
IF (((FILE(I).AND. 'FFOO'X).EQ. '2E00'X) .JTDO=I
IF (((FILE(I).AND. 'DDFF'X).EQ. '002E'X) .JTDO=I

293
IF((BCFILE(I).AND.'FF00'X).EQ.'2ED0'X) JBDO0=I
IF((BCFILE(I).AND.'DD0F'X).EQ.'DD0E'X) JBDDE=0
IF((EFILE(I).AND.'FF00'X).EQ.'2ED0'X) JEDDO=I
IF((EFILE(I).AND.'DD0F'X).EQ.'DD0E'X) JEDDE=I
CONTINUE
JT=JTDO0-3
JB=JBDDE-3
JC=JBDO0-3
JE=JEDDE-3
IF(JT.LT.1) JT=1
IF(JB.LT.1) JB=1
IF(JC.LT.1) JC=1
IF(JE.LT.1) JE=1
LINE2(4)=1
LINE2(8)=1
LINE2(12)=1
LINE2(16)=1
DO 2855 I=20,30
2855 LINE2(I)=1
C STORE THE 1ST 6 CHARACTERS BEFORE THE PERIOD IN
C EACH OF THE INPUT DATA SET NAMES
C DO 478 I=1,3
LINE2(I)=TFILE(JT-1+I)
LINE2(I+4)=TFILE(JB-1+I)
LINE2(I+8)=CFILE(JC-1+I)
LINE2(I+12)=CFILE(JC-1+I)
LINE2(I+16)=EFILE(JE-1+I)
CONTINUE
DO 2245 I=1,60
LINE1(I)=TLINE1(I)
2248 LINES(I)=TLINE3(I)
C PUT TARGET DISTANCE AND WINDOW WIDTH AND POINTS REMOVED
C IN 3RD LINE OF HEADER
C ENCODE(22,108,STRING)TDIST,WIN,WRENED
108 FORMAT(1,TD-1,F4.1,1 W,F4.1,' PR=',I2,X)
DO 727 NSTRG=1,15
727 BUFFER191=NSTRG=BUFFER190-NSTRG
DO 728 NSTRG=1,22
728 BUFFER144=NSTRG=STRING(NSTRG)
C 1051 IF(ISUP.EQ.1) THEN
NCUP=NCUP+1
CALL CBUP(ARR,NCUP,KRPT,IEND,SPAG,KSTP)
IF(IEND.EQ.1) GOTO 1081
GOTO 1052
END IF
1081 TYPE 288
ACCEPT 3852,ARR
Here begins the new order of post-calibration options

1052 IF (ARR.ED.'WRIT') THEN
  TYPE = 'Writ File'
  CALL WRFILE(RFILE,RS)
  IF(KSTP.EQ.0) THEN
    GOTO 2020
  ELSE
    GOTO 1091
  END IF
END IF

1092 IF (ARR.ED.'TDW') THEN
  ITPL=0
  CALL TDW(INFILE,RS,NMIN,ITPL,ISUP,SWID1,SWID2,WTP,ALPHA,DEMOD,SPAG)
END IF

1053 IF (ARR.ED.'FPL') THEN
  MIN=NP
  ANGT=MIN*1000.
  AIND=MINT(FINC*1000.,)
  CALL APP(RFILE,RS,SPAG,PAG)
END IF

1054 IF (ARR.ED.'TPL') THEN
  ITPL=1
  CALL TDW(RFILE,RS,NMIN,ITPL,ISUP,SWID1,SWID2,WTP,ALPHA,DEMOD,SPAG)
END IF

Note TPL uses a modified version of TDW to achieve a plot

ITPL=0
END IF

1055 IF (ARR.ED.'CHK') THEN
  IC=1
  GOTO 1019
1029 IC=0
  END IF

1056 IF (ARR.ED.'HLP') THEN
  CALL HLP
  END IF

1057 IF (ARR.ED.'TDW'.OR.ARR.ED.'FPL'.OR.ARR.ED.'CHK') GOTO 1051
IF (ARR.ED.'TPL') GOTO 1051
IF (ARR.ED.'WIN'.OR.ARR.ED.'EXA'.OR.ARR.ED.'EX') GOTO 1234
IF (ARR.ED.'TAR'.OR.ARR.ED.'BTA'.OR.ARR.ED.'BPA') GOTO 1234
IF (ARR.ED.'BSP') GOTO 1234
IF (ARR.ED.'BUP') THEN
  TYPE 5599
5599 FORMAT(1X,'Y OR N FOR NEW SET UP ',1,4)
  ACCEPT 5588,NEW_BUP
SUBROUTINE RDFLE(INFILE, CA)
INCLUDE 'CALGOM, FOR'
COMMON /BLK1/BUFF /BLK2/np, FMIN, FMax /BLK3/IB
INTEGER 8 INFLE(7)
REAL 4 AP(JD), A(MD), P(ND)
BYTE BUFF(ID], IDUFF
1512)
EQUIVALENCE (BUFF(351) ,AP(l))
INCLUDE 'SYSLIBRARY: FORIOSDEF'

INFILE(7:0 = 0
IB = 0
ICT = 0
OPEN (UNIT=8, NAME=INFILE, READONLY, TYPE=' Old', IOSTAT=IERR, ERR=8100)
IF (IB EQ 1) LEN=512-8*4
IF (IB GT 1) LEN=512-28*4
READ(8, 80, END=80) (BUFF(I),I=1,812)
FORMAT(512A1)
DO 86 I=1, LEN
85 BUFF(ICT+I)=BUFF(I)
ICT=ICT+LEN
GO TO 82
80 DO 86 I=1, LEN
86 BUFF(ICT+I)=BUFF(I)
DO 40 I=1,18D
40BUFF(I)=BUFF(2*I-1)
TYPE 1882, (BUFF(I),I=1,18D)
FORMAT(1X,60A1)
CALL DCDE
PI=4.*ATAN(1.)
DO 198 NN=1, NP
ATMP=10***(AP(2*NN-1)/20.)
PTMP=AP(2*NN)/18D.*PI
198 CA(NN)=CMPLX(ATMP*COS(PTMP),ATMP*SIN(PTMP))
SUBROUTINE WRTFILE(AF,CA,*)

INCLUDE 'CALCDM.FOR'
COMMON /BLK1/BLK2/BLK3/IB
COMMON /BLK1/BUFF/BLK2/BLK3/IB

READING(5D) CA
BYTE LINE1(60),LINE2(60),LINE3(60),BUFF(I),TBUFF(512)
CHARACTER*1 YN
REAL*4 API,J0
INTEGER*4 RFILE(15)

GO TO 40 1(I-1,101)

IF(YN.NE.'Y')GO TO 122

OPEN(UNIT=10,NAME=AF,TYPE='NEW',RECORDSIZE=NB)

IF(INF.0,0)
ELSE IF(INF.1,0)

FORMAT('FILE: ',8A2,' WAS NOT FOUND',/,'ENTER FILENAME AGAIN')
ELSE IF(INF.0,0)

FORMAT('FILE: ',8A2,' WAS BAD, ENTER NEW FILENAME')
ELSE

TYPE *, 'UNRECOVERABLE ERROR, CODE = ',IERR
STOP
ENDIF
ACCEPT 8818,(INFILE(IIT),IIT=1,6)

FORMAT(8A2)
GO TO 8010

CLOSE(UNIT=8,DISP='SAVE')
RETURN
END

SUBROUTINE WRTFILE(AF,CA,*)

INCLUDE 'CALCDM.FOR'
COMMON /BLK1/BLK2/BLK3/IB
COMMON /BLK1/BUFF/BLK2/BLK3/IB

READING(5D) CA
BYTE LINE1(60),LINE2(60),LINE3(60),BUFF(I),TBUFF(512)
CHARACTER*1 YN
REAL*4 API,J0
INTEGER*4 RFILE(15)

GO TO 40 1(I-1,101)

IF(YN.NE.'Y')GO TO 122

OPEN(UNIT=10,NAME=AF,TYPE='NEW',RECORDSIZE=NB)

IF(INF.0,0)
ELSE IF(INF.1,0)

FORMAT('FILE: ',8A2,' WAS NOT FOUND',/,'ENTER FILENAME AGAIN')
ELSE IF(INF.0,0)

FORMAT('FILE: ',8A2,' WAS BAD, ENTER NEW FILENAME')
ELSE

TYPE *, 'UNRECOVERABLE ERROR, CODE = ',IERR
STOP
ENDIF
ACCEPT 8818,(INFILE(IIT),IIT=1,6)

FORMAT(8A2)
GO TO 8010

CLOSE(UNIT=8,DISP='SAVE')
RETURN
END

SUBROUTINE WRTFILE(AF,CA,*)

INCLUDE 'CALCDM.FOR'
COMMON /BLK1/BLK2/BLK3/IB
COMMON /BLK1/BUFF/BLK2/BLK3/IB

READING(5D) CA
BYTE LINE1(60),LINE2(60),LINE3(60),BUFF(I),TBUFF(512)
CHARACTER*1 YN
REAL*4 API,J0
INTEGER*4 RFILE(15)

GO TO 40 1(I-1,101)

IF(YN.NE.'Y')GO TO 122

OPEN(UNIT=10,NAME=AF,TYPE='NEW',RECORDSIZE=NB)

IF(INF.0,0)
ELSE IF(INF.1,0)

FORMAT('FILE: ',8A2,' WAS NOT FOUND',/,'ENTER FILENAME AGAIN')
ELSE IF(INF.0,0)

FORMAT('FILE: ',8A2,' WAS BAD, ENTER NEW FILENAME')
ELSE

TYPE *, 'UNRECOVERABLE ERROR, CODE = ',IERR
STOP
ENDIF
ACCEPT 8818,(INFILE(IIT),IIT=1,6)

FORMAT(8A2)
GO TO 8010

CLOSE(UNIT=8,DISP='SAVE')
RETURN
END
SUBROUTINE DCDE
INCLUDE 'CALC6VM.FOR'
COMMON /BK1/BUFF /BK2/NP,FMIN,FINC /BK3/IB
BYTE BUFF(I)
INTEGER I,M,NP

C NO. OF DATA POINTS IS STORED IN THREE CHARACTERS, AND
C STARTING FREQ. AND FREQ. INC. IN 5 CHARACTERS
C
CHARACTER*4 CNL
CHARACTER*5 OFF,CFF
C
EQUIVALENCE (BUFF(123),CNL),(BUFF(131),CFF),(BUFF(140),CINC)
C
CONVERT CHARACTERS TO THEIR NUMERICAL EQUIVALENTS
C
IF(BUFF(123).EQ.'-') BUFF(123)='0'
C
DECODE(4,100,CNL)NP
NP=MNH(NP,NPP)

100 FORMAT(I5)
DECODE(5,100,CFF)MIN
DECODE(5,100,CINC)INC
FMIN=FLOAT(MIN)/1000.
FINC=FLOAT(INC)/1000.
RETURN
END

C K=TOTAL NO. OF POINT
C N=NO. OF POINTS OF SMOOTHING
C
SUBROUTINE HAMM(CS,K,N)
INCLUDE 'CALC6VM.FOR'
COMMON /BK1/BUFF /BK2/NP,FMIN,FINC /BK3/IB
COMPLEX CS(MD),CSS(MD),SUM,FACT

N2=(N-1)/2
DO 3 KM=1,K
SUM=CMPLX(0.,0.)
KOUNT=1

3 KM=KM*N2+KOUNT-1
OFF=FLOAT(KM*KM)*TPI/FLOAT(N)

298
IF(KW.LT.11  
M2-KW  
IF(KW.GT.K) KW=2  
K  
KkV  
L  
~~W3S(0FF)+1.  
4  
FACT=W*MS(  
KW)  
0WJ+fACT  
IF(KCIJNT.EU.N)  
90  
TO  
I  
KDIJNT=-KDINT+1  
90  
TO 2  
i  
CONTINUE  
9UM/J9FL0AT(NJ  
Cas  
(Lm)=ajN  
a  
CONTINUE  
DO06  
K1  
CS(KN)=-CSS(KNJ  
6  
CONTINUE  
ED  
C  
C  
SUBROUTINE TAR(TA,TFILE,TLINE1,TLINE3,NPT,*  
INCLUDE 'CALCDM.FOR'  
COMMON /BLKI/BUFF /BLK2/NP,FMIN,FINC /BLK2/IB  
COMPLEX TA(MD)  
INTEGER*2 TFILE(7)  
BYTE LINE1(60),LINES(60),BUFF(ID),TLINE1(60),TLINE3(60)  
EQUIVALENCE(LINE1(1),BUFF(1)),(LINES(1),BUFF(121))  
10  
FORMAT(I3,' INPUT THE TARGET FILE NAME')  
TYPE 10  
ACCEPT 120,(TFILE(IIN),IIN=-1,7)  
120  
FORMAT(7AJ2)  
CALL RDPLE(TFILE,TA)  
NPT=N  
DO 2346 I=1,N  
TLINE1(I)=LINE1(I)  
2346  
TLINE3(I)=LINES(I)  
RETURN1  
END  
C  
C  
SUBROUTINE BTA(BT,BFILE,NPBT,*)  
INCLUDE 'CALCDM.FOR'  
COMMON /BLKI/BUFF /BLK2/NP,FMINC /BLK2/IB  
COMPLEX BT(MD)  
INTEGER*2 BFILE(7)  
BYTE BUFF(ID)  
TYPE 20  
20  
FORMAT(I3,' INPUT THE BACKGROUND FILE NAME FOR THE TARGET')  
ACCEPT 120,(BFILE(IIN),IIN=-1,7)  
120  
FORMAT(7AJ2)  
CALL RDPLE(BFILE,BT)  
NPBT=N  
299
SUBROUTINE SRF(CAB, CFILE, TFILE, TLINE1, TLINE3, NPC, *)

INCLUDE 'CALCOM.FOR'
COMPLEX CAB(MD)
INTEGER*2 CFILE(7), TFILE(7)
BYTE LINE1(60), LINE3(60), BUFF(12)
EQUIVALENCE (LINE 1, BUFF(1)), (LINE 3, BUFF(12))
TYPE 40
FORMAT (1X, /, 'INPUT THE CALIBRATION TARGET FILE NAME,')
ACCEPT 120 (CFILE(IN), IIN=1,7)
120 FORMAT (7A)
CALL RFILE(CFILE, CAB)
NPC = NP
RETURN
END

C C C
SUBROUTINE BSR(BC, BCFILE, NPC, *)
INCLUDE 'CALCOM.FOR'
COMMON /BLK1/BUFF /BLK2/NP, FMIN, FINC /BLK3/IB
COMPLEX BC(MD)
INTEGER*2 BCFILE(7)
BYTE BUFF(12)
EQUIVALENCE (LINE 1, BUFF(1)), (LINE 3, BUFF(12))
TYPE 41
FORMAT (1X, /, 'INPUT THE BACKGROUND FILE NAME FOR THE')
$ CALIBRATION TARGET,')
ACCEPT 120 (BCFILE(IN), IIN=1,7)
120 FORMAT (7A)
CALL RFILE(BCFILE, BC)
NPC = NP
RETURN
END

C C C
SUBROUTINE EXA(EXS, EFILE, NPC, *)
INCLUDE 'CALCOM.FOR'
COMMON /BLK1/BUFF /BLK2/NP, FMIN, FINC /BLK3/IB
COMPLEX EXS(MD)
INTEGER*2 EFILE(7)
BYTE BUFF(ID)

50 FORMAT(1X,'INPUT THE EXACT CALIBRATION FILE NAME')
ACCEPT 120, (EFILE(IIN),IIN=1,7)

120 FORMAT(7A2)
CALL RDPL(EFILE,EXG)
NPE=NP
RETURN
END

SUBROUTINE WIN(DWIN,TDIST,TO,*):
CNE=2.99E+08
CFI=9.698E+08

C COMPUTE TIME SHIFT IN DATA TO CENTER IT OVER THE HAMMING WINDOW
TDIST=DISTANCE TO TARGET IN METERS

C TYPE *, 'ENTER ONE WAY DISTANCE TO TARGET IN METERS'
C TYPE *, 'DUAL AEL HOMS'
C TYPE *, 'OVER & UNDER PRIOR JULY 4, 1983 = 12.1 METERS'
C TYPE *, 'OVER & UNDER AFTER JULY 4, 1983 = 11.8 METERS'
C TYPE *, 'SIDE BY SIDE AFTER JULY 4, 1983 = 11.8 METERS'
ACCEPT *,TDIST
TO=2*TDIST/CME
C
C COMPUTE HAMMING WINDOW SIZE, (FEET, SECONDS)

C TYPE *, 'ENTER 3DB DOWN HAMMING WINDOW SIZE IN FEET'
C TYPE *, 'ENTER 0 FOR NO HAMMING WINDOW'
ACCEPT *,DWIN
RETURN
END

SUBROUTINE TNW(INFILE,RG,NPWIN,ITPL,ISUP,GWID1,GWID2,WT,WTP,ALPHA,DEMOD
$,
SPAG,PAG)
INCLUDE 'CALCON.FOR'
COMMON /BLK1/BUFF /BLK2/NP,FKIN,FINC /BLK3/IB
COMMON /BLK4/NDI,N,ANST,AINC
CONPLEX*8 RS(5000)
REAL*4 AM(5000),PH(5000)
INTEGER*2 INFILE(15)
C
IF (ITPL.EQ.1) GOTO 8014
 IF (ISUP.NE.1.OR. (GWID1.EQ.0.0.AND.GWID2.EQ.0.0)) THEN
  TYPE '*.1 *
  TYPE*, 'TYPE STARTING AND END POINTS FOR WINDOW'
  ACCEPT *,GWID1,GWID2
END IF
C
301
Calculate the number of points in time domain filter

\[
\text{NTDPTS1}=\text{NINT}((\text{GWID1}+4.096)^*\text{NINT}(\text{FINC}^*1000.))
\]
\[
\text{NTDPTS2}=\text{NINT}((\text{GWID2}+4.096)^*\text{NINT}(\text{FINC}^*1000.))
\]
\[
\text{NTDPTS} = \text{NTDPTS2} + \text{NTDPTS1}
\]

Convert from COMPLEX(x,y) to AMPLITUDE(dB) and PHASE(deg)

for compatibility with FTRAN modules.

DO 9015 I=1, NMFIN
    AM(I)=20.*ALOG10(CABS(RS(I)))
    PH(I)=ATAN2D(AIMAIG(RS(I)),REAL(RS(I)))
    CONTINUE

IF (ITPL.NE.1) THEN
    CALL PMN(AM, PH)
END IF

CALL IFT(AM, PH, ISUP, DEMOD)

IF (ITPL.NE.1) THEN
    CALL RPL(INFILE, AM, SPAG, PAG)
END IF

IF (ITPL.WU.1) THEN
    CALL FWN(AM, RI)
    CALL IFT(AM, RI, ISJP, DENDO)
END IF

IF (ITPL.EQ.1) THEN
    CALL SAT(AM, PH, NTDPTS1, NTTS2, IOUP, WTP, ALPHA)
    CALL FFT(AN, PH, JNOIN, INFILE)
END IF

CONTINUE

DO 9029 I=1, (NDIN-IFIX(FNIN*100.))
    RS(I)=COMPLEX(COSD(PH(K33)),SIND(PH(K33)))*(10**(-AM(K)/20.))
    CONTINUE

DO 9015 I=1, NMFIN
    ANST=FMIN1000.
    AIND=NINT(FINC*1000.)
    RETURN

END

SUBROUTINE SUP(ISUP, ACFN, YDES, WTP, ALPHA, GWID1, GWID2, DEMOD, PAG)
COMMON /BLKS/ARRA, NRPT
COMMON /BLK5/ARRA, NRPT
COMMON /BLK6/PLTDPTS, IP31, LP11, MOD1, FXE, FXS, FYE, FYS
COMMON /BLK7/PLTDPTS2, IP32, LP12, MOD2, TXE, TXS, TYS
DIMENSION ARRA(3D)
This array contains the commands.
ISUP=1  I Go into set-up mode
NA=1   I Command number

SUBROUTINE SUP(ISUP, ACFN, YDES, WTP, ALPHA, GWID1, GWID2, DEMOD, PAG)
53  FORMAT(1X,'COMMAND: ',I2,' ',@1)
   ACCEPT 55, ARRA(NA)
55  FORMAT(A3)
   CALL ESUP(ARRA, NA, IESUP)
   IF(IESUP.EQ.1) GOTO 51
   IF(ARRA(NA).EQ. 'END') GOTO 57
   IF(ARRA(NA).EQ. 'RPT') THEN
     TYPE *, 1
     TYPE 59
   END IF
   IF(IESUP.LE.I) SOTO 51
   IF(ARRA(NA).EQ. 'RPT') THEN
     I Repeat a section of commands
   TYPE *, 1
   TYPE 59
   FORMAT(1X,'TYPE NUMBER OF REPETITIONS ',@1)
   ACCEPT *, NRPT
   GOTO 57
   END IF
   NA=NA+1
   GOTO 51
C
C Input a series of control parameters
C
57  TYPE *, 1
   TYPE 61
61  FORMAT(1X,'Y OR N FOR AUTO FILE NAME
   ACCEPT 62,ACP-N
   FORMAT(1X,'Y OR N FOR BAD POINT REMOVAL
   ACCEPT 62, YDES
   TYPE 69
69  FORMAT(1X,'Y OR N FOR FREQ.GREATER THAN 20 GHZ
   ACCEPT 62, DEMO
   TYPE 71
71  FORMAT(1X,'Y OR N FOR PLOT AGAIN OPTION
   ACCEPT 62, PAG
   TYPE 68
68  FORMAT(1X,'Y OR N FOR PLOT SET-UP IN FPL
   ACCEPT 62, PLOTOPT1
   TYPE 7712
7712 FORMAT(1X,'Y OR N FOR PLOT SET-UP IN TPL
   ACCEPT 62, PLOTOPT2
   IF(PLOTOPT1.EQ. 'Y') THEN
     TYPE *, 1
     FPL PARAMETERS
   TYPE 7700
7700 FORMAT(1X,'Y OR N FOR ANOTHER PLOT ON OLD ONE
   ACCEPT 62, LPL1
   TYPE 7702
7702 FORMAT(1X,'INPUT NO. OF DIGITS RIGHT OF .XXX
   ACCEPT *, NOIG1
   TYPE 7704
7704 FORMAT(1X,'INPUT MIN, MAX, INC FREQUENCY
   ACCEPT *, FMX, FXE, FXG
   TYPE 7706
7706 FORMAT(1X,'INPUT MIN, MAX, INC FREQ. AMPLITUDE
   ACCEPT *, FMB, FYE, FYG
   END IF
   IF(PLOTOPT2.EQ. 'Y') THEN
     TYPE *, 1
     TPL PARAMETERS
   END IF

303
TYPE 7700
ACCEPT 62,LPL2
TYPE 7702
ACCEPT *, NDIG2

TYPE 7708
7708 FORMAT(IX, 'INPUT MIN,MAX,INC TIME', $)
ACCEPT *, TYS, TXE, TXS

TYPE 7710
7710 FORMAT(IX, 'INPUT MIN,MAX,INC TIME. AMPLITUDE', $)
ACCEPT *, TYS, TXE, TXS

END IF

TYPE 64
64 FORMAT(IX, 'Y OR N FOR HANNING OR BESSEL FILTER', $)
ACCEPT 62, WTP
IF (WTP.EQ.'N') THEN

TYPE 65
65 FORMAT(IX, 'INPUT ALPHA', $)
ACCEPT *, ALPHA

END IF

TYPE 67
67 FORMAT(IX, 'INPUT WINDOW START AND END POINTS nS', $)
ACCEPT *, GWID1, GWID2
RETURN

END

SUBROUTINE ESUP(ARRA, NA, IESUP)
! Find errors in commands
DIMENSION ARRA(30)
C=ARRA(NA)
IF (C.EQ.'TAR'.OR.C.EQ.'BTA'.OR.C.EQ.'SPH'.OR.C.EQ.'BSP') GOTO 570
IF (C.EQ.'WIN'.OR.C.EQ.'EXA'.OR.C.EQ.'EXI'.OR.C.EQ.'MLP') GOTO 570
IF (C.EQ.'CAL'.OR.C.EQ.'WRT'.OR.C.EQ.'FPL'.OR.C.EQ.'TPL') GOTO 570
IF (C.EQ.'TOM'.OR.C.EQ.'LAB'.OR.C.EQ.'RPT'.OR.C.EQ.'CHK') GOTO 570
IF (C.EQ.'END') GOTO 570
TYPE *, 'ENTER A VALID COMMAND'
IESUP=1
RETURN
570
IESUP=0
RETURN
END

SUBROUTINE CSUP(ARR, I, KRP, IEND, SPAG, KSTP)
COMMON /LKV/ARRA, NRPT
DIMENSION ARRA(30)
! Control subroutine for SUP
IF (KSTP.NE.0.AND.I.EQ.KSTP) THEN

KSTP=0
IEND=1
RETURN
END IF

C
1984
ARR=ARRA(I)
C
IF (ARR(I).EQ.'END') THEN
! Relinquish control set-up but
IEND=1
RETURN

RETURN

304
END IF

IF (I .GT. 1 .AND. ARRA(I) .EQ. 'END') THEN
    IEN=I
    ISUP=0
    RETURN
END IF

IF (ARR .EQ. 'LAB') THEN
    LAB (label) is starting point of
    NLAB=I+1
    I=I+1
    GOTO 1984
END IF

IF (SPAG .EQ. 'Y') THEN
    I
    This allow plots to be
    REDONE
    ARR=ARRA(I)
END IF

IF (NRPT .EQ. 0) RETURN

RElinquish SUP mode after
repetitions are done

RETURN

SUBROUTINE HLP(*)

TYPE 'LIST OF COMMANDS'
TYPE '*,'SUP'=SET UP A SERIES OF COMMANDS AND PARAMETERS'
TYPE '*,'TAR'=ENTER DATA INTO TARGET ARRAY'
TYPE '*,'BTA'=ENTER DATA INTO TARGET BACKGROUND ARRAY'
TYPE '*,'BSP'=ENTER DATA INTO CALIBRATION TARGET ARRAY'
TYPE '*,'EXA'=ENTER DATA INTO EXACT CALIBRATION TARGET ARRAY'
TYPE '*,'WIN'=CHANGE TARGET DISTANCE AND WINDOW WIDTH'
TYPE '*,'CAL'=CALIBRATE DATA'
TYPE '*,'CHK'=CHECK FILE STATUS AND PARAMETERS FOR CAL'
TYPE '*,'FOW'=GATE IN TIME DOMAIN'
TYPE '*,'FPL'=PLOT CALIBRATED DATA IN FREQUENCY DOMAIN'
TYPE '*,'TPL'=PLOT CALIBRATED DATA IN TIME DOMAIN'
TYPE '*,'WRT'=WRITE A CALIBRATED FILE'
TYPE '*,'HLP'=PRINT LIST OF COMMANDS'

RETURN

END
*EXIT* = EXIT FROM PROGRAM GRACEFULLY

*NOTE: MAX FILE NAME LENGTH = 12 CHARACTERS*

*NOTE: THE "CAL" COMMAND DOES NOT WRITE A FILE IN CAL61*

THIS IS ACHIEVED BY A SEPARATE "WRT" COMMAND

RETURN

END
THIS SUBROUTINE REMOVES LARGE NON-PERIODIC SPIKES FROM DATA IN
COMPLEX FORMAT. WRITTEN BY DONALD F. KIMBALL ON
JANUARY 10, 1983

INPUT TO SUBROUTINE REMOVE CONSISTS OF REAL AND IMAGINARY PARTS
CONTAINED IN A SINGLE COMPLEX VARIABLE "RECT", AND THE NUMBER OF DATA POINTS. THE OUTPUT UNITS OF THE
SUBROUTINE ARE THE SAME AS THE INPUT UNITS.

SUBROUTINE CREMOVE(RECT, NP, NREMOV, NSPIKE, FINC)

NREMOV = no. of pts. removed
NSPIKE = index no. of pts. removed

INCLUDE 'CALCOM.FOR'
DATA CFT/ 8.8865E+8 /
INTEGER NSPIKE(MD)
COMPLEX RECT(MD),CAVG,CSQAVG,CVAR,CSTDEV

NAV1 MUST BE AN EVEN NUMBER

NSPIKE(1)=0
NREMOV =0

Note: NAVG must be an even number

FIN = FINC * 1.E+9       | frequency increment
TZONE = 20.        | target zone in feet
SEN = 8.            | SEN standard deviations
ISKIP = (CFT/(TZONE * FIN)) - 1

ISKIP = no. of pts. to skip in average
ISKIP is derived from Shannon's theorem
ISKIP must be an odd number

IF((ISKIP/2)*2 .NE. ISKIP ) ISKIP = ISKIP + 1
TYPE *, 'ISKIP = ',ISKIP

TYPE *, 'IONE = ',IONE

dont use IONE points on either side test point

NAV1 = (2*CFT/(TZONE * FIN)) - ISKIP

NAV1, number of points to average over
this quantity is derived from the
the sinx/x filter.
NAV1 must be an even number

IF((NAV1/2)*2 .NE. NAV1 ) NAV1 = NAV1 + 1
IF(NAV1 .LT. 4) NAV1 = 4

TYPE*, 'NAV1 = ',NAV1

DO 606 J=1,NP

307
AVGD=0.  ! initialize the average summer
CAVG=0.0
C0AVG=0.0
C0AVGDB=0.
NUM_SUM = 0  ! number of points added up

COMPUTE AVERAGE OF (NAVG) POINTS AROUND POINTS J-1, J,
AND J+1, BUT NOT INCLUDING THOSE POINTS IN THE AVERAGE.

I = J - NAVG / 2 - INOLE
IF( I .LT. 1 ) I = 1
IDIS = NAVG - J  ! how close to end
IF( IDIS .LT. NAVG / 2 + INOLE )
I = I - NAVG / 2 + INOLE - IDIS
IF( IDIS .LE. INOLE ) I = I + INOLE - IDIS

TYPE *, ' J = ', J, 'SETTING I TO', I

IF( IABS( I - J ) .LE. INOLE ) GO TO 22 I don't average over middle

TYPE *, ' CALCULATING FOR J, I, NP', J, I, NP

CAVG = CAVG + RECT(I)
DB_NUM = 20 * LOG10( CABS( RECT(I) ) )
AVGDB = AVGDB + DB_NUM
C0AVGDB = C0AVGDB + DB_NUM ** 2
C0AVG = C0AVG + CMPLX( REAL(RECT(I)) ** 2, AIMAG(RECT(I)) ** 2)
NUM_SUM = NUM_SUM + 1

I = I + 1
IF( NUM_SUM .LT. NAVG ) GOTO 111

CAVG = CAVG / NUM_SUM
AVGDB = AVGDB / NUM_SUM
C0AVGDB = C0AVGDB / NUM_SUM
C0AVG = C0AVG / NUM_SUM

CALCULATE VARIANCE AND STANDARD DEVIATIONS

VARDB = C0AVGDB - AVGDB ** 2
CVAR = C0AVG - CMPLX( REAL(CAVG) ** 2, AIMAG(CAVG) ** 2)
IF( VARDB .LT. 0. ) VARDB = 0.
IF( REAL(CVAR) .LT. 0. ) CVAR = CMPLX( 0., AIMAG(CVAR) )
IF( AIMAG(CVAR) .LT. 0. ) CVAR = CMPLX( REAL(CVAR), 0. )
STDEVDB = SQRT(VARDB)
STDEV = CMPLX( SQRT(REAL(CVAR)), SQRT(AIMAG(CVAR)) )

SEE IF POINTS J-1, J, AND J+1 DEVIATES TOO FAR FROM THE AVERAGE BY
COMPARING THEM TO THE STANDARD DEVIATION

IF(J-1.LT.1) GO TO 202
IF(ABS(20 * LOG10(CABS(RECT(J-1)))) - AVGDB) .GT. SEN * STDEVDB) GOTO 101
IF(ABS(REAL(RECT(J-1) - CAVG)) .GT. SEN * REAL(STDEV)) GOTO 101

308
IF(ABS(AIMAG(RECT(J-1)-CAVG)).LT.SEN*AIMAG(CSTDEV)) GOTO 202
NREMOV=NREMOV+1
NSPIKE(NREMOV)=J-1
TYPE '*','POINT ','J-1,' REMOVED'
RECT(J-1)=CAVG
202 IF(ABS(20*LOG10(CABS(RECT(J)))-AVGDB).GT.SEN*STDEVDB) GOTO 303
IF(ABS(REAL(RECT(J)-CAVG)).GT.SEN*REAL(CSTDEV)) GOTO 303
IF(ABS(AIMAG(RECT(J)-CAVG)).LT.SEN*AIMAG(CSTDEV)) GOTO 404
303 NREMOV=NREMOV+1
NSPIKE(NREMOV)=J
TYPE '*','POINT ','J,' REMOVED'
RECT(J)=CAVG
404 IF(J+1.GT.NP) GO TO 808
IF(ABS(20*LOG10(CABS(RECT(J+1)))-AVGDB).GT.SEN*STDEVDB) GOTO 505
IF(ABS(REAL(RECT(J+1)-CAVG)).GT.SEN*REAL(CSTDEV)) GOTO 505
IF(ABS(AIMAG(RECT(J+1)-CAVG)).LT.SEN*AIMAG(CSTDEV)) GOTO 606
505 NREMOV=NREMOV+1
NSPIKE(NREMOV)=J+1
TYPE '*','POINT ','J+1,' REMOVED'
RECT(J+1)=CAVG
606 CONTINUE
TYPE '*','NREMOV,' POINTS REMOVED'
RETURN
END
SUBROUTINE FWNV(A,M,FH)

COMMON /BLK1/BUFF /BLK2/MP,FMIN,FINC /BLK3/IB
COMMON /BLK4/NDIM,ANST,AINC
BYTE BUFF(32200)
INTEGER*2 INFILE(7)
DIMENSION AM(5000),PH(5000)
CHARACTER*2 WT,LP
LOGICAL CE,WTY
EXTERNAL BID
DATA LP/'LP'/
PI=3.1415926
NDIM=NP
ANST=FMIN*1000.
AINC=NINT(FINC*1000.)
TYPE *,'
TYPE *,'STARTING TO WINDOW DATA TO REDUCE PRECURSORS IN TIME DOMAIN'
WRITE(6,*) 'INPUT T OF F TO CONSERVE SIGNAL ENERGY'
ACCEPT *, CE
CE=0.
IF(CE) THEN
S1=0.
S3=0.
DO 10 I=1,NDIM
S3=S3+10.**(AM(I)/20.*COSO(PH(I)))
10 S1=S1+10.**(AM(I)/10.)
END IF
WRITE(6,*) 'INPUT WINDOW TYPE, LP (LOW PASS) OR BP (BAND PASS)' ACCEPT , WT
IF(WT.EQ.'LP') THEN
INTLB=1
IF(INTLB.EQ.0) THEN
IW=NDIM
DT=PI/FLOAT(IW)
WRITE(6,*) 'INPUT T FOR HANNING OR F FOR KAISER-BESSSEL'
ACCEPT *, WT
IF(WT) THEN
DO 30 I=1,IW
WFACT=.5*(1.+COS( DT*I))
WRITE(6,*) 'INPUT T FOR HANNING OR F FOR KAISER-BESSSEL'
ACCEPT *, WT
IF(WT) THEN
DO 30 I=1,IW
WFACT=(COS(.5*DT*I)+.22*COS(1.5*DT*I))/1.22
AA=10.*AM(I)/20.*WFACT
IF(AA.LE.0.) AA=1.E-30
30 AM(I)=20.*ALOG10(AA)
ELSE
WRITE(6,*) 'INPUT ALPHA'
ACCEPT *, ALPHA
ALPHA=2.
TYPE *,'
TYPE *,'ALPHA=2 IN BESSEL WINDOWING FILTER'
PIA=ALPHA*PI
CON=BIO(PIA)
DO 31 I=1,IW
ARG=PIA*SQRT(1.-FLOAT(I)/FLOAT(IW))**2
WFACT=0.5*0.15/CON
AA=10.*AM(I)/20.*WFACT
IF(AA.LE.0.) AA=1.E-30
31 AM(I)=20.*ALOG10(AA)
END IF
ELSE
WRITE(6,*),'INPUT T FOR HANNING OR F FOR KAISER-BESSSEL'
C ACCEPT *, WTY
C IF(WTY) THEN
IWT=1
IF(WTY.EQ.0) THEN
IWT=2*IW-1
IF(FLOAT(NDIM)/2..EQ.FLOAT(IW)) THEN
DT=PI/FLOAT(IW)
II=0
DO 40 I=1,IW,2
II=II+1
WFACT=.5*(1.+COS(DT*I))
IF(WFACT.LE.1.E-4)
40 AM(IW+II)=20.*ALOG10(10.*AM(IW+II)/20.*WFACT)
ELSE
AM(IW+II)=20.*ALOG10(IW**AM(IW+II)/20.*WFACT)
END IF
ELSE
WRITE(6,*),'INPUT ALPHA'
C ACCEPT *, ALPHA
ALPHA=-2
TYPE *, 'ALPHA=-2 IN BESSSEL WINDOWING FILTER'
PIA=ALPHA*PI
CON=BIO(PIA)
IWT=NDIM/2
II=2*IW-1
IF(FLOAT(NDIM)/2..EQ.FLOAT(IW)) THEN
DT=1./FLOAT(IW)
II=0
DO 50 I=1,IW,2
II=II+1
ARG=-PIA*SQRT(1.-DT*I)**2
WFACT=0.15/CON
AM(IW+II)=20.*ALOG10(IW**AM(IW+II)/20.*WFACT)
50 AM(IW-I+1)=20.*ALOG10(10.*AM(IW-I+1)/20.*WFACT)
END IF
ELSE
WRITE(6,*),'INPUT T FOR HANNING OR F FOR KAISER-BESSSEL'
C ACCEPT *, WTY
C IF(WTY) THEN
IWT=1
IF(WTY.EQ.0) THEN
IWT=2*IW-1
IF(FLOAT(NDIM)/2..EQ.FLOAT(IW)) THEN
DT=1./FLOAT(IW)
II=0
DO 60 I=1,IW,2
II=II+1
ARG=PIA*SQRT(1.-DT*I)**2
WFACT=0.15/CON
AM(IW+II)=20.*ALOG10(IW**AM(IW+II)/20.*WFACT)
60 AM(IW-I+1)=20.*ALOG10(10.*AM(IW-I+1)/20.*WFACT)
ELSE
DT=1./FLOAT(IW)

DO 70 I=1, IW
    ARG=PIA*SQRT(1.-((DT*I)**2))
    WFACT=BID(ARG)/CON
    AM(IW+I+1)=20.*ALOG10(10.**[AM(IW+I+1)/20.]*WFACT)
70    AM(IW+I+1)=20.*ALOG10(10.**[AM(IW+I+1)/20.]*WFACT)
END IF
END IF
IF(CE) THEN
    S2=0.
    S4=0.
    DO 80 I=1,NDIM
        S4=S4+10**[(AM(I)/20.)*COSD(PH(I))]
        S2=S2+10**[(AM(I)/10.)]
        SC=SQRT(S1/S2)
        SCI=S1/S2
        WRITE(*,*) 'S1,S2,SC=',S1,S2,SC
        WRITE(*,*) 'S3,S4,SCI=',S3,S4,SCI
    DO 80 I=1,NDIM
80    AM(I)=20.*ALOG10(10.**[(AM(I)/20.)*SC]
END IF
RETURN
END
Subroutine GAT
Part of SQCL61.FOR, a group of FFT subroutines.
Developed by A. Dominik to perform gating in the time domain, modified by N.F. Chamberlain Jan. 1984 for use with CAL81.FOR.

SUBROUTINE GAT(AN, PH, NTOPTS1, NTOPTS2, ISUP, WTP, ALPHA)
COMMON /BLK1/BUFF /BLK2/NP, FNIN, FINC /BLK3/NB
COMMON /BLK4/NOM, ANST, AINC
BYTE BUFF(32200)
INTEGER*2 INFILE(15)
DIMENSION AM(5000), PH(5000)
LOGICAL WTP,CE
LOGICAL CE
EXTERNAL BID
PI=3.1415927
TYPE *, 'STARTING GATING ROUTINE IN TIME DOMAIN'
WRITE(6,*), 'INPUT T OR F TO CONSERVE SIGNAL ENERGY'
ACCEPT *, CE
CE=O
WRITE6,O), 'INPUT STARTING AND ENDING INDEX NUMBERS'
ACCEPT *, IS, IE
IS=2048-NTOPTS1
IE=2048+NTOPTS2
IST=16-1
IF(CE) THEN
S1=0.
S3=0.
DO 10 I=IS,IE
S3=S3+AM(I)
10 SI=SI+AM(I)*2
END IF
IF(ISUP.NE.1) THEN
WRITE6,O), 'INPUT Y FOR HANNING OR N FOR KAISER-BESSEL'
ACCEPT 11, WTP
11 FORMAT(A1)
END IF
DO 20 I=1,IS-1
AM(I)=O.
IF(WTP.EQ.'Y') THEN
NN=IE-IS+I
IN=NN/2
IN=2*IN-1
IF(FLOAT(NN)/2.,ED.FLOAT(IN)) THEN
DT=PI/FLOAT(IN)
DO 30 I=1,IN,2
WFAC=.5*(1.+COS(DT*I))
AM(I)=AM(I)+AM(I)*WFAC
30 AM(I)=AM(I)+AM(I)*WFAC
ELSE
DT=PI/FLOAT(IN)
DO 40 I=1,IN
WFAC=.5*(1.+COS(DT*I))
AM(I)=AM(I)+AM(I)*WFAC
40 AM(I)=AM(I)+AM(I)*WFAC

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ED IF
ELSE
IF(ISJP.NE.1) THEN
WRITE(6,*),'INPUT ALPHA'
ACCEPT *, ALPHA
END IF
PIA=PI*ALPHA
CON=BI0(PIA)
NM=IE-IS+1
IN=NN/2
IN=2*IN-1
IF(FLOAT(NM)/2..EQ.FLOAT(IN)) THEN
DT=1./FLOAT(IN)
DO 21 I=1,IN,2
ARG=PIA*SQRT(1.-(DT*1)**2)
WFACT=BIO(ARG)/CON
AM(IN-ISI+I)=AM(IN-ISI+I)*WFACT
21 AM(IN-ISI-I+1)=AM(IN-ISI-I+1)*WFACT
END IF
DO 50 I=IE+1,NDIM
50 AM(I)=0.
IF(IE) THEN
S2=0.
S4=0.
DO 80 I=IS,IE
S4=S4+AM(I)
80 S2=S2+AM(I)**2
SC0=SQRT(S1/S2)
SC1=S2/64
WRITE(6,*),'S1,S2,SC='S1,S2,SC
WRITE(6,*),'S3,S4,SC1='S3,S4,SC1
DO 70 I=IS,IE
70 AM(I)=AM(I)*SC
END IF
FUNCTION BID(X)
XX=X/2.
S=1.
S1=1.
FAC=1.
ACO=.00001
DO 10 I=1,100
FAC=FAC*I
S1=S1*XX
T=S
RETURN
END
S=S+(S/FAC)**2
TT=ABS((T-S)/S)
IF(TT.LT.ACC) GO TO 20
CONTINUE
20 BIO=S
IF(I.EQ.101) WRITE(6,'*') 'BESSEL FUNCTION DID NOT CONVERGE'
RETURN
END

C Subroutine IFT
C Part of BCL61. FOR, a group of FFT subroutines.
C Developed by A. Dominek to perform inverse Fourier
transformations in FTRAN. FOR, modified by N. F. Chamberlain for use in
CAL61. FOR
C
SUBROUTINE IFT(AM, PH, ISUP, DEMOD)
COMMON /BLK1/ BUFF, /BLK2/ NP, FMIN, FINC /BLK3/ IB
COMMON /BLK4/ NDIM, ANST, AINC
C INTEGER*2 INFILE(15)
BYTE BUFF(32000)
COMPLEX A(5000)
LOGICAL LUMP, LMG
DIMENSION AM(5000), PH(5000), S(1250)
PI=3.1415926
DTR=PI/180.
N=12
NS=2048
NST=2*NS
TNS=NST
NDIM=NP
ANST=FMINS*1000.
AINC=XINT(FINC*1000.)
IF(ISUP.NE.1) THEN
TYPE '*', 'STARTING INVERSE FOURIER TRANSFORM'
WRITE(6,'(A*)&') 'INPUT Y OR N FOR FRED. GREATER THAN 20 GHZ'
ACCEPT 5, DEMOD
FORMAT(A1)
END IF
IF(DEMOD.EQ.'N') THEN
M=NUMBER OF LEADING ZEROS
M=ANST/AINC+.1
N2=NDIM+N1
C CHECK FOR THE PROPER NUMBER OF DATA POINTS
IF(N2.GT.NS) WRITE(6,'(A*)&') 'TOO MANY SAMPLES', N2
IF(N2.GT.NS) NDIM=NS-N1
IF(N2.GT.NS) N2=NS
IF(M.GEQ.0.) GO TO 31
DO 30 I=1,M
C FILL IN LEADING ZEROS
30 A(I)=CMPLX(0.,0.)
C FILL IN DATA POINTS
31 DO 40 I=1,NDIM

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```
N1=N1+1
IF(N1.GT.NS) GO TO 51
ANG=PI(N1)*DTR
AMP=10.**[AN(I)/20.]
A(N1)=COMPLEX(COS(ANG),SIN(ANG))*AMP

C TRAILING ZEROS
NX=NS-N1-NDI
DO 50 I=1,NX
A(N2+I)=COMPLEX(0.,0.)
50 CONTINUE

C CONSTRUCT NEGATIVE FREQ. IMAGE
ANG=PI(N1)*DTR
C TRAILING ZEROS
NX=NS-N1-NDI
DO 70 I=1,NX
IF(FLAT(NDI)/2..E.FLATT(NDI/2)) NN=NDI-1
N2=NDI/2
IF(N2.GT.NS-1) WRITE(*,*) 'TOO MANY SAMPLES', N2
W=2.**PI*(ANST+N2*AINC)/1000.
ANG=PI(N1)*DTR
A(I)=0.**[AM(N2+1)/20.]*COMPLEX(COSD(ANG),SIND(ANG))
DO 70 I=2,NS
IF(N2.GT.NS-1) WRITE(*,*) 'TOO MANY SAMPLES', N2
W=2.**PI*(ANST+N2*AINC)/1000.
ANG=PI(N1)*DTR
A(I)=0.**[AM(N2+1)/20.]*COMPLEX(COSD(ANG),SIND(ANG))
DO 70 I=2,NS
IF(N2.GT.NS-1) WRITE(*,*) 'TOO MANY SAMPLES', N2
A(I)=COMPLEX(0.,0.)
70 CONTINUE
A(I)=COMPLEX(0.,0.)
END IF
CALL FORT(A,N,S,1,IERR)
IF(IERR.NE.0) TYPE *, 'ERROR IN FORT1',IERR

C DETERMINE TIME AXIS SCALING
FFMX=AINC
ANST=-1.E3/FFMX/2.
FMAX=FFMN
AINC=1.EB/FFMX/TNS
NDI=TNS
IF(DEMOD.EQ.'Y') THEN
A(I)=2.*A(I)
W0=2.*AINC*0.5
DO 76 I=1,NS-1
IF(N2.GT.NS-1) WRITE(*,*) 'TOO MANY SAMPLES', N2
A(I)=REAL(A(I))**COS(W0**I)-AIMAG(A(I))**SIN(W0**I)
76 A(I)=COMPLEX(AA,A0)
END IF

C SHIFT 'O TIME' TO CENTER OF PLOT
DO 70 I=1,NS
```

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Subroutine FFT

Part of SCL61.FOR, developed by A. Dominski to perform forward Fourier transformation in FTRAN.FOR. Modified by N.F.Chamberlain for use in CAL53.FOR.

SUBROUTINE FFT (AM, I, J, NDIM, INFILE)
COMMON /BLK1/BUFF /BLK2/NP,FMIN,FINC /BLK3/IB
COMMON /BLK4/NDIM, ANST,AINC
INTEGER*2 INFILE(7)
BYTE BUFF(33200)
COMPLEX A(500)
DIMENSION AM(5000),PH(5000),B(1250)
LOGICAL INWN
CHARACTER*2 WT,LP
DATA LP/'LP'/
PI=3.141582
M=12
NS=2048
TNS=4096.
TYPE *,','
TYPE *, 'STARTING FORWARD FOURIER TRANSFORM'
SHIFT '0 TIME' TO THE LEFT END OF PLOT
DO 10 I=1,NS
K=NS+I
A(K)=CMPLX(AM(I),0.)
10 CONTINUE
NST=TNS
DO 20 I=1,NS
A(I)=CMPLX(AM(1)+I),0.)
20 CONTINUE
CALL FORT(A,M,S,-1,IERR)
WRITE(6,'(8X,2I5,2F5.0)') 'INPUT NDIM,ANST,AINC'
NOTE CAL53 WORKS TO ONE LESS DATA POINT THAN FTRAN!
NDIM=1601-1+200
ANST=0
AINC=10
ACCEPT *,NDIM,ANST,AINC
WRITE(6,'(8X,5X)') 'APPLY INVERSE WINDOW TO F'

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C ACCEPT, INWIN
NDIM=ND+INT(PWIN*1000.*0.1)
ANST=0
AINC=INT(FINC*1000.)
INWIN=0
IF(INWIN) THEN
WRITE(*,*) 'INPUT WINDOW TYPE, LP (LOW PASS) OR BP (BAND PASS)'
ACCEPT 1,WT
FORMAT(1A2)
IF(WT.EQ.'LP') THEN
IN=NDIM
DT=PI/FLOAT(IN)
DO 15 I=1,IN
WFACT=.5*(1.+COS(DT*(I-.5)))
AM(I)=20.*ALOG10(CABS(A(I+1))/WFACT)
PH(I)=ATAN2D(AIMAG(A(I+1)),REAL(A(I+1)))
ELSE
NN=INT(ANST/AINC+.1)
IN=NDIM/2
IF(INWIN.EQ.'BP') THEN
DT=PI/FLOAT(IN)
DO 30 I=1,IN
WFACT=.5*(1.+COS(DT*(I-.5)))
AM(I+1)=20.*ALOG10(CABS(A(2*NN+I+1))/WFACT)
AM(I-I)=20.*ALOG10(CABS(A(2*NN-I+1))/WFACT)
PH(I+1)=ATAN2D(AIMAG(A(2*NN+I+1)),REAL(A(2*NN+I+1)))
PH(I-I)=ATAN2D(AIMAG(A(2*NN-I+1)),REAL(A(2*NN-I+1)))
ELSE
AM(I+1)=20.*ALOG10(CABS(A(2*NN+I+1)))
PH(I+1)=ATAN2D(AIMAG(A(2*NN-I+1)),REAL(A(2*NN-I+1)))
END
DO 40 I=1,IN
WFACT=.5*(1.+COS(2.*DT*I))
AM(I+1)=20.*ALOG10(CABS(A(2*NN+I+1))/WFACT)
AM(I-I)=20.*ALOG10(CABS(A(2*NN-I+1))/WFACT)
PH(I+1)=ATAN2D(AIMAG(A(2*NN+I+1)),REAL(A(2*NN+I+1)))
PH(I-I)=ATAN2D(AIMAG(A(2*NN-I+1)),REAL(A(2*NN-I+1)))
END IF
ENDIF
ELSE
NN=INT(ANST/AINC+.1)
DO 50 I=1,NN
AM(I)=20.*ALOG10(CABS(A(I)))
PH(I)=ATAN2D(AIMAG(A(I)),REAL(A(I)))
ENDIF
C CALL LAB(INFILE)
JNDIM=NDIM
RETURN
END
Subroutine RPL

Part of a group of plotting subroutines called PLOTV.FOR; originally developed by A. Dominek to achieve time domain plots in FTRAN.FOR, modified for use with CAL51.FOR by N.F. Chamberlain, Jan. 1984.

SUBROUTINE RPL(INFILE, AM, SPAG, PAG)

COMMON /BLK1/BUFF /BLK2/NP,FMIN,FINC /BLK3/IB
COMMON /BLK4/KONM, ANST, AINC
COMMON /BLK7/PLOTOPT2, IP32, LPL, NOIS, XB, XS, YE, YS
COMMON /PLT1/YB, GSY, GDX, ND1, ND2, TH(5000), ISYM
BYTE BUFF(33200)
REAL*4 AN(5000);
INTEGER*2 LINE1(30), LINE2(30), PARA(30), INFILE(15)

LOGICAL LPL

EQUIVALENCE (LINE1(1),BUFF(1)), (LINE2(1),BUFF(61)),
1(PARA(1),BUFF(121))

DATA LMASK'/88888888X/

YB = YB

C

IF(LPL) THEN

IF(LPL.EQ.'Y') THEN
ISYM = ISYM - 1
AINC = AINC * 1.E-5
DO 31 I = ND1, ND1 + ND2 - 1
AMT = AMT / 2
IF(AMT .LT. YB) AMT = YB
31
AMT(I - ND1 + 1) = AMT
CALL STRYP(AMT, YB, GSY, TH, GDX, ND2, 1, ISYM)
IF(PLOTOPT2.EQ.'Y') THEN
TYPE*,','
WRITE(*,*) 'INPUT Y OR N FOR ANOTHER CURVE ON SAME PLOT
ACCEPT 32, LPL
END IF
32
FORMAT(A1)

IF(LPL.EQ.'N') THEN
CALL PLOT(0, 0, 999)
CALL PLOTHEW(IN)
CALL LIB$SPAWN('VPLTO')
END IF
ELSE
ISYM = -1
AINC = AINC * 1.E-5
C
CALL VPLTO(0, 0, IDUM)
CALL PLOT(7, 1.75, -3)
XL = 8.
ANST = ANST + AINC * NOIM
TYPE*, ',
IF (PLOTOPT2.EQ.'Y') THEN
TYPE*, ', 'BEGINNING/END TIME AND STEP TIME SIZE'
ACCEPT *, XB, XS
END IF
N02=INT((XE-XB)/AINC)+1
TH(1)=XB
AMAX=1000.
AMIN=1000.
AINC=ABS(AINC)
DO 20 I=1,NDIN
   IF(AM(I).LT.AMIN) AMIN=AM(I)
20   IF(AM(I).GT.AMAX) AMAX=AM(I)
   DO 21 I=1,ND2
   TH(I)=TH(I-1)+AINC
21   ND1=INT((XB-AMST)/AINC)+1
   TYPE *,
   TYPE 25,AMIN,AMAX
F0RMAT(' Minimum is',F17.3,/, ' Maximum is',F17.3)
YL=6.
IF(PLOTOPT2. NE. 'Y') THEN
   TYPE *
   TYPE , 'INPUT BEGINNING/END MAGNITUDE AND STEP MAGNITUDE SIZE'
   ACCEPT *,YB,YE,YS
   END IF
   IF(PLOTOPT2. NE. 'Y') THEN
   TYPE *
   TYPE , 'INPUT THE NUMBER OF DIGITS TO THE RIGHT OF .XXX'
   ACCEPT *,NDIG
   END IF
   DO 30 I=ND1,NDI+ND2-1
      AMT=AM(I)
      IF(AMT.LT.YB) AMT=YB
30   AMT(I-ND1+1)=AMT
   SMX=X
   DSMX=X/XL
   XL=XL/XD
   SPX=XS*DXL
   IPX=ABS(XS/XD+0.5)
   CALL FFAXIS(0.,0.,16TIME IN NANOSEC, -16, XL, 90., SMX, DSMX, 1SPX, NDIG, 1., 0)
   YD=YE-YB
   SYM=YB
   SYM=YD/YL
   SYL=YL/YD
   SPY=YS*YL
   IPY=ABS(YD/YS+0.5)
   CALL FFAXIS(0.,0.,16IMPULSE RESPONSE,16, YL, 90., SYM, SYM, 1SPY, -1, 1, 0)
   YLS=IPYS*SPY
   IP33=2*IP32
   CALL STRYP(AMT,-YB,DSY,TH,0.,0.,DSX,ND2,1.,SYM)
   CALL GRID(YLG,0.,IPY,SPY,IPX,SPX,LMASK)
   CALL SYMBOL(-6.6,0.,10,%FILE,90.,80)
   CALL SYMBOL(-6.8,7.,10,%FILE,80.,IP33)
   CALL SYMBOL(-6.4,0.,10,%FILE,90.,80)
   CALL SYMBOL(-6.2,0.,10,%FILE,90.,80)
   IF(PLOTOPT2. NE. 'Y') THEN
   TYPE *,

320
WRITE(6,*) 'INPUT Y OR N FOR ANOTHER CURVE ON SAME PLOT'
ACCEPT 92, LPL
END IF
IF(LPL.EQ.'N') THEN
CALL PLOT(0.,0.,088)
CALL PLOTNOW(ING)
CALL LIB*SPAWN('VPLLOT')
END IF
END IF
AINC=AINC*1.55
C
IF(PAG.EQ.'Y') THEN
TYPE *, '1'
TYPE 1985
FORMAT(1X,'Y OR N FOR PLOT AGAIN ',@)
ACCEPT 1986,SPAG
1986 FORMAT(A1)
END IF
C
RETURN
END
C
Subroutine APP, part of the plotting package PLOTV.FOR.
C
Developed by A. Vomina to achieve amplitude and phase plots in the
frequency domain in FTRAN.FOR, modified by N.F. Chamberlain Jan. 1984
for use with CEL61.FOR
C
SUBROUTINE APP(INFILE,RS,SPAG,PAG)
C
COMMON /BLK1/BUFF /BLK2/NP,FMIN,FINC /BLK3/IB
COMMON /BLK4/NDIM,ANST,AINC
COMMON /BLK5/PLOTOPT1,IP31,LPL,NDIG,XB,XE,YS,YE,YS
BYTE BUFF(32200)
DIMENSION AM(5000),PH(5000),AM(2050)
COMPLEX*8 RS(5000)
INTEGER*2 LINE1(30),LINE2(30),PARA(30),INFILE(15)
COMMON/PLTf/)BB,DSX, YP, YBA, OYP, OYB, YEA, TH(2050),ISYI
EQUIVALENCE (LINE1(I),BUFF(I)),(LINE2(I),BUFF(12I))
1(PARA(I),BUFF(12I))
LOGICAL LPLOT
DATA LMAK/'**9888888888**'/
C
DO 9015 I=1,NP
AM(I)=2.0*LOG10(CABS(RS(I)))
PH(I)=ATAN2(AIMAG(RS(I)),REAL(RS(I)))
9015 CONTINUE
IF(LPL.EQ.'Y') THEN
ISYM=ISYM-1
XB1=2.*XB-ANST
XB1=XB1/1000.
TH(I)=XB/1000.
AINC=AINC/1000.
DO 5 I=2,NDIM
5 CONTINUE
TH(I)=TH(I-1)+AINC
DO 10 I=1,NDIM
   AMT=AM(I)
   IF(AMT.GT.YEA) AMT=YEA
   IF(AMT.LT.YBA) AMT=YBA
10  AMT(I)=AMT
   CALL PLOT(-.55,-.75,-3)
   CALL PLOT(.55,5.55,-3)
   CALL STRYP(TH,XB1,DSX,PH,YBP,DSYP,NDIM,1.,ISYN)
   CALL PLOT(0.,-4.8,-3)
   CALL STRYP(TH,XB1,DSX,AMT,YBA,DSYA,NDIM,1.,ISYN)
   IF(PLOTOPT1.NE.'Y') THEN
      TYPE *, 'WRITE(6,*)' INPUT Y OR N FOR ANOTHER CURVE ON SAME PLOT'
      ACCEPT 32, LPL
      END IF
32  FORMAT(A1)
   IF(LPL.EQ.'N') THEN
      CALL PLOT(0.,0.,999)
      CALL PLOTNOW(10M)
      CALL LIB6SPAW('VPLOT')
      END IF
   ELSE
      CALL VPLOTS(0,0,10M)
      CALL PLOT(.55,5.55,-3)
      ISYN=1
   END IF
   C DO PHASE PLOT FIRST
   C CALCULATE THE X AXIS INFORMATION
   XL=6.
   XB2=ANST
   XE2=XB2+AINC*(NDIM-1)
   IF(XB2.GT.XE2) THEN
      XX=XB2
      XB2=XX
      XE2=XX
   END IF
   XB1=XB2
   TYPE *, '
   TYPE *, 'THE INITIAL AND FINAL FREQUENCIES ARE', XB2/1000.,XE2/1000.
   IF(PLOTOPT1.NE.'Y') THEN
      TYPE *, '
      TYPE *, 'INPUT THE INITIAL, FINAL AND STEP SIZE FREQUENCIES IN GHZ'
      ACCEPT *, XE,XE
      END IF
   XB3=XB1*1000.
   XE3=XE*XE
   XB3=XE3
   XB3=XB3
   XB1=XB1+XB3
   XB1=XB1/1000.
   TYPE *, 'INPUT STEP FREQUENCY SIZE'
   C ACCEPT *,XB3
   TH(1)=XB3/1000.
   C PMAX=PH(1)
   C PHN=PH(1)
DO 20 I=2,NDIM
TH(I)=TH(I-1)+AINCC
C IF(PH(I).LT.PMIN) PMIN=PH(I)
C IF(PH(I).GT.PMAX) PMAX=PH(I)
IF(AM(I).LT.AMIN) AMIN=AM(I)
20 IF(AM(I).GT.AMAX) AMAX=AM(I)
X0=X0+X3
X3=X3/1000.
DEX=EX/10X/1000.
DNL=NX/XD
SPX=X0*DX
IPX=ABS(XD*X3+0.5)
CALL FFAXIS(0.,0.,16FREQUENCY IN GHZ,-16,0.,SMX,DEX,
16PX,2,1.,0)
C CALLULATE THE Y AXIS INFORMATION
YL=4.
C TYPE *, 'THE MINIMUM AND MAXIMUM PHASE VARIATION IS', PMIN, PMAX
C TYPE *, 'INPUT BEGINNING/END PHASE AND STEP MAGNITUDE SIZE'
C ACCEPT *,YB,YE,YS
YB2=180.
YB2=90.
YB2=90.
SY=YS-YB
SMY=YB2
DSY=YO/QL
DYL=YL/WD
SPY=YS2*DYL
IPY=ABS(YO/YS2+0.5)
CALL FFAXIS(0.,0.,16PHASE IN DEGREES,16,0.,SMY,DSY,
16PY,-11.1,0)
YL=SIPY*SPY
C PLOT THE CURVE, GRID AND IDENTIFICATION INFORMATION
YPB=YP2
DSY=DSY
IPB=IP31
CALL STRYPI(TH,X3,DSY,YPB,IPB,SMY,1.,ISYM)
CALL GRID(O.,0.,IPX,SPX,IPY,SPY,LMASK)
CALL SYMBOL(.5,4,7,10,6FILE:.,0.,8)
CALL SYMBOL(12.4,7,10,%REF(INFILE),0.,IP31)
CALL SYMBOL(.5,4,8,10,%REF(LINE1),0.,80)
CALL SYMBOL(.5,4,9,10,%REF(LINE2),0.,80)
CALL SYMBOL(.5,4,10,10,%REF(Para),0.,80)
CALL PLOT(.5,-4.8,9)
C PLOT THE MAGNITUDE NOW
CALL FFAXIS(0.,0.,16FREQUENCY IN GHZ,-16,0.,SMX,DEX,
16PX,2,1.,0)
TYPE *,'
TYPE 25,AMX,AMAX
25 FORMAT(' Minimum is',F10.3,' DB',/,' Maximum is',F10.3,' DB')
IF(PLOTOPT.1.E.'Y') THEN
TYPE *,'
TYPE *, 'INPUT BEGINNING/END MAGNITUDE AND STEP MAGNITUDE SIZE'

323
ACCEPT *, YB, YE, YS
END IF

IF(PLOTOPT. NE.'Y') THEN
  TYPE '!
  TYPE 'INPUT THE NUMBER OF DIGITS TO THE RIGHT OF .000'
  ACCEPT *, NDIG
END IF

DO 30 I=1, NDIG
  AMT=AMT(I)
  IF(AMT.GT.YE) AMT=YE
  IF(AMT.LT.YB) AMT=YB
  AMT(I)=AMT
  YD=YE-YB
  SINY=YB
  DSY=VD/YL
  DYL=YL/YD
  SPY=Y*YYL
  IPY=ABS(YD/YS+0.5)
  CALL FFAIX(0.,0.,15H1NAGNITUDE IN DB, 15, YL, 90., 15Y, 01, NDIG, 1., 0)
  YLS=IPY*SPY
  YEA=YE
  0BA=YS
  DSYA=DSY
  CALL STRYP(TH, XH, DEX, AMT, YBA, DSYA, NDIG, 1., ISYM)
  CALL GRID(0., 0., IPX, SPX, IPY, GPY, LPSK)
  TYPE 'IS THIS THE LAST PLOT, T TO F'
  ACCEPT *, LPL
  IF(LPL) THEN
    CALL PLOT(0., 0., 999)
    CALL PLOTNOW(IMG)
  ELSE
    CALL PLOT(0., 0., -999)
  END IF
END IF

IF(PLOTOPT. NE.'Y') THEN
  TYPE '!
  WRITE(6, #) 'INPUT Y OR N FOR ANOTHER CURVE ON SAME PLOT'
  ACCEPT 32, LPL
END IF

IF(LPL.EQ.'N') THEN
  CALL PLOT(0., 0., +999)
  CALL PLOTNOW(IMG)
END IF

CALL LIBSPAWN('VPLOT')
END IF

RETURN
END
Program Name: CATLOG.FOR

Modified from B3.FOR (by J. Chen) during 1994 by Neil Chamberlain.

Program now automatically fully scales a directory of calibrated data files, in conjunction with FILESORT.COM.

This program makes use of 2-18 GHz calibrated data and generates a full-scale data file.

REAL*4 FA(1601), A218(1601), P218(1601)
INTEGER*2 LINEX(24)
COMPLEX*8 TA(1601), CA(1601), R(1601)
CHARACTER INFILE*31, DIR_FILE*11, ID*48, T_POL*2, F_INC*4
CHARACTER ELEV*2, S_FAC*4, W_0*3, ORIG*7, RFilee*10, T_SHIP*2, T_ASP*2

COMMON BUFF, NDM, ANST, AINC
COMMON /BLK2/ T_SHIP, T_ASP

TYPE ' ', 'Input routine ...'

TYPE ' ', 'Input polarization type (FV, RH, FX)'
ACCEPT 7, T_POL
FORMAT(2A)

TYPE ' ', 'Input target elevation (15.27 deg)'
ACCEPT 7, ELEV

TYPE ' ', 'Input frequency increment MHz'
ACCEPT *, F_INC

TYPE ' ', 'Input frequency increment again'
ACCEPT 9, F_INC
FORMAT(A4)

TYPE ' ', 'Input a 7 char. description of source data'
ACCEPT 10, ORIG
FORMAT(A7)

TYPE ' ', 'Input directory file name'
ACCEPT 5, DIR_FILE
FORMAT(A11)
PRINT 1771, 'Directory file name ', DIR_FILE
PRINT 1772
FORMAT(0)
1771 FORMAT(1X, A2, A11)

TYPE ' ', ' }
TYPE *, 'Input number of files in directory'
ACCEPT *, NLEN

End of input routine

Open directory file so that file headers may be read

OPEN (UNIT=2, NAME = DIR_FILE, READONLY, TYPE='OLD', ERR=999)

DO 1111, I_MAIN=1, NLEN
  IF(I_MAIN.EQ.1) THEN
    TYPE *, ' '  
    TYPE 15
    FORMAT(1X, 'Starting scaling procedure')
    END IF
  READ(I,12) INFIL
  FORMAT(A)
  INFIL=INFIL(20:29)
  TYPE*, 'File', I_MAIN, ', INFIL
  FORMAT(1X, A4, I3, A4, A10)
  CALL READ(INFILE, A218, P218)

Convert data from dB and deg to napers and radians.
Note that data is taken and stored in the former format.
Chan stores his data in dB and rad and this format is necessary
for use in DBPLT, the full scale plotting routine.

PI=4.*ATAN(1.)
DTR=PI/180.
DO 929 I=1, 1601
  A218(I)=10.**(A218(I)/20.)
  P218(I)=P218(I)*DTR
929 CONTINUE

IF(T_SHIP.ED.'MO'.OR.T_SHIP.ED.'LB') THEN
  SF=2400.
  WD=0.25
  S_FAC(I:4)='2400'
  W_D(I:3)='1/4'
  ELSE
    SF=1200.
    WD=0.5
    S_FAC(I:4)='1200'
    W_D(I:3)='1/2'
  END IF

Compile a file name for the scaled target

  FILE(I:2)=T_SHIP(I:2)
  FILE(3:4)=T_PDL(I:2)
  FILE(5:6)=T_ASP(I:2)
  FILE(7:8)=ELEV(I:2)
Compile a header to describe the scaled target

```c
ID(1:8) =RFILE(1:8)
ID(9:9) = ' '  
ID(10:11)=TASP(1:2)
ID(12:12)= ' '  
ID(13:15)= 'DES'  
ID(16:16)= ' '  
ID(17:19)= 'SF='  
ID(20:23)=0_FAC(1:4)  
ID(24:24)= ' '  
ID(25:28)= 'FIND='  
ID(30:33)=F_INC(1:4)  
ID(34:34)= ' '  
ID(35:37)= 'WD='  
ID(38:40)=M_D(1:3)  
ID(41:41)= ' '  
ID(42:48)=ORIG(1:7)

TYPE = ' '  
TYPE 33,ID  
PRINT 33,ID  
33  
FORMAT(1X,4BA)  
PRINT ' ', ' '  
```

Calculate full scale frequencies according to scale factor

```c
DO 46 II=1,1801
FA(II)=(2.+II-1)/100.*SF*1000.
46 CONTINUE
```

CREATE A FULL-SCALE COMPLEX TARGET

```c
DO 3540 IJ=1,1801
CA(IJ)=COMPLX(SF*A21(IJ)*COS(P218(IJ)),SF*A218(IJ)*SIN(P218(IJ)))
3540
```

STORE THE INPUT-FILE NAMES IN A BUFFER

```c
DO 987 IK=1,24
LINE2(IK)= ' '  
987
```

SMOOTH THE FULL-SCALE DATA FILE BY CONVOLVING

```c
THE FILE WITH A HAMMING WINDOW
```

CALCULATE PARAMETERS FOR THE CONVOLUTIONS, WHERE,

```c
FLOW IS THE LOWEST FULL-SCALE FREQUENCY 
FHIGH IS THE HIGHEST FULL-SCALE FREQUENCY 
FINC IS THE FULL-SCALE FREQUENCY INCREMENT 
```

```c
FLOW=2./SF*1000.  
FHIGH=18./SF*1000.  
FLOW=INT(FLOW*10.)/10.+1  
FHIGH=INT(FHIGH*10.)/10.  
```
I=1
K=0
ICNT=1601

TYPE *,',':
TYPE ',',FA(1)='FA(1)',FA(ICNT)='FA[ICNT]
TYPE ',',FLOW='FLOW,RHIGH='RHIGH
TYPE ',',ICNT=','ICNT,NPT=','NPT
TYPE ',',SF='SF

PICK UP THE DESIRED FREQUENCY AND POSITION THE
CENTER OF THE HAMMING WINDOW AT THAT FREQUENCY.
ESTIMATE THE DATA VALUE OF THE DESIRED FREQUENCY BY
TAKING THE WEIGHTED AVERAGE OF THE NEIGHBORING DATA POINTS
COVERED BY THE HAMMING WINDOW, WITH THE WEIGHTINGS DETERMINED
THE HAMMING WINDOW.

DO 7720 RF=FLOW,RHIGH,FINC

IF(RF.GE.FA(I).AND.RF.LE.FA(I+1))GO TO 7730
I=I+1
IF(I.GT.ICNT)GO TO 4812
GO TO 7740

7730 K=K+1
CALL INTER(R,CA,FA,ICNT,NS,IK,WD,RF)

7720 CONTINUE

OUTPUT THE SMOOTHED FULL-SCALE DATA FILES

4812 CALL PDATA(RFILE,R,K,FLOW,FINC,SF,ID,LINE2)

1111 CONTINUE
CLOSE(UNIT=2,DISP='SAVE',ERR=1001)
GOTO 9090

699 TYPE *,'Open error - file not found'
GOTO 11
1001 TYPE *,'Close error'
STOP
9090 END

SUBROUTINE PDATA(RFILE,CA,NPT,FLOW,FINC,SF,ID,LINE2)

THIS SUBROUTINE WRITES A FILE ON A STORAGE UNIT

INTEGER*2 LINE(24)
CHARACTER ID*48,RFILE*10
REAL*4 AM(801),RM(801)
COMPLEX*8 CA(801)

DO 210 I=1,N:
IF(CABS(CA(I)).EQ.0.)CA(I)=(1.E-15,1.E-15)

328
SUBROUTINE INTER(R, CA, ICNT, NS, I, WD, RF)
COMPLEX*8 R(1601), CA(1601)
REAL*4 FA(1601)
WEI=0
IG=1
XTMP=0,
YTMP=0,
RF=RF-WD/2,
RFF=RF+WD/2.
20 IF(IS.GT.ICNT)GO TO 10
IF(FA(IS).GT.RFH)GO TO 1D
HANG=-.5*4.4*COS(3.141592*T/WD)
XTMP=XTMP+HANG*REAL(CA(IS))
YTMP=XTMP+HANG*AIMAG(CA(IS))
WEI=WEI+HANG
IG=IG+1
GO TO 20
10 IG=IG-1
15 IF(IS.LT.IJ)GO TO 30
IF(FA(IS).LT.RFL)GO TO 30
T=FA(IS)-RF
HANG=-.5*4.4*COS(3.141592*T/WD)
XTMP=XTMP+HANG*REAL(CA(IS))
YTMP=XTMP+HANG*AIMAG(CA(IS))
WEI=WEI+HANG
IG=IG-1
GO TO 15
30 R(K)=CMPLX(XTMP/WEI, YTMP/WEI)
RETURN
END

SUBROUTINE PHCR(RPH, FA, IBASE, PH)
REAL*4 PH(201), FA(201)
TWOPI=6.*ATAN(1.)
PI=TWOPX/2.

30 IF(DPH.LT.PI) GO TO 20
DPH=DPH+TWOPX
GO TO 30

20 IF(DPH.GT.-PI) GO TO 40
DPH=DPH-TWOPX
GO TO 20

40 FBASE=FA(IBASE)
DO 10 I=1,201
10 PH(I)=PH(I)-(DPH-FA(I)/FBASE)
RETURN
END

C
C
C THIS SUBROUTINE ESTIMATES THE VALUES OF
C PHASE RETURNS AT THE ENDS OF A DATA FILE,
C USING A LINEAR INTERPOLATION
C
C SUBROUTINE EPH(PHM,PH,NS,NF,NI)
REAL*4 PH(201)
X=0
Y=0
XY=0
X2=0
DO 10 I=NS,NF,NI
X=X+FLOAT(I)
Y=Y+PH(I)
XY=XY+FLOAT(I)*PH(I)
10 X2=X2+FLOAT(I)*FLOAT(I)

XN=(NS-NF)*NI+1
DELTA=(XN*X2-X*X)
A=(XN*XY-X*Y)/DELTA
B=(X2*Y-X*XY)/DELTA
PHM=A*NS+B
RETURN
END

C
C
C THIS SUBROUTINE READS A FILE AND EXTRACTS ASPECT AND SHIP-TYPE
C INFORMATION FROM THE HEADER
C
C SUBROUTINE REA(INFILE,AN,PH)
COMMON BUFF,NDIM,ANST,AINC
COMMON /BLK2/ T_SHP,T_ASP
C
BYTE BUFF(3200)
INTEGER*2 LINE2(30),PARAM(30)
REAL*4 AP(4000),AM(1601),PH(1801)
CHARACTER*2 T_SHP,T_ASP,SHIP(6),ASP(14)
CHARACTER*1 NUM(10),LINE1(60),INFILE*10,CFILE*60
C
DATA SHIP/"AD","LR","LB","KH","KX","MD"/
DATA ASP/"00","10","15","20","30","40","45","50","60","80","90"/
DEFINE BUFFER STRUCTURE

EQUIVALENCE(LINE1(I),BUFF(11),(LINE2(I),BUFF(61))
1,PARAM(1),BUFF(121)),(AP(1),BUFF(361))

READ A FILE
CALL TR(INFILE)

Here begins the name and aspect extractor

DO 2222 I=1,100
CFILE(I:I)=LINE(I)
2222 CONTINUE
K=0
DO 8888 I=7,30

IF (K.NE.0) GOTO 5577
DO 8888 J=1,10
IF (CFILE(I:I).EQ.NUM(J)) THEN
K=I
GOTO 8888
END IF
8888 CONTINUE

5577 IF(K.NE.0) GOTO 8888
DO 7788 L=1,10
IF(CFILE(I:I).EQ.NUM(L)) THEN
GOTO 8888
END IF
7788 CONTINUE
IF(CFILE(I:I),EQ.11) THEN
L_STOP=I-1
GOTO 9999
END IF

9999 CONTINUE
DO 6699 IC=7,20
IF(CFILE(IC:IC+2).EQ. 'DES') THEN
N_FIND=IC+3
GOTO 7755
END IF
6699 CONTINUE

7755 DO 4444 I=N_FIND,N_FIND+8
IF(CFILE(I:I).EQ. 'A') THEN
ISHIP=1
GOTO 3333
END IF
IF(CFILE(I:I).EQ. 'M') THEN
ISHIP=6
GOTO 3333
4444 CONTINUE

331
END IF
IF(CFILE(I:1+1).EQ. 'LR') THEN
  ISHIP=2
GOTO 3333
END IF
IF(CFILE(I:1+1).EQ. 'DB') THEN
  ISHIP=3
GOTO 3333
END IF
IF(CFILE(I:1+1).EQ. 'KH') THEN
  ISHIP=4
GOTO 3333
END IF
IF(CFILE(I:1+1).EQ. 'XX') THEN
  ISHIP=5
GOTO 3333
END IF

CONTINUE

C
3333 IF(CFILE(K:L_STOP).EQ. '0') IASP=1
IF(CFILE(K:L_STOP).EQ. '10') IASP=2
IF(CFILE(K:L_STOP).EQ. '15') IASP=3
IF(CFILE(K:L_STOP).EQ. '50') IASP=4
IF(CFILE(K:L_STOP).EQ. '40') IASP=5
IF(CFILE(K:L_STOP).EQ. '45') IASP=6
IF(CFILE(K:L_STOP).EQ. '50') IASP=7
IF(CFILE(K:L_STOP).EQ. '50') IASP=8
IF(CFILE(K:L_STOP).EQ. '50') IASP=9
IF(CFILE(K:L_STOP).EQ. '50') IASP=10
IF(CFILE(K:L_STOP).EQ. '50') IASP=11
IF(CFILE(K:L_STOP).EQ. '100') IASP=12
IF(CFILE(K:L_STOP).EQ. '170') IASP=13
IF(CFILE(K:L_STOP).EQ. '90') IASP=14

C
TYPE ",","'
C
TYPE ",', 'Information from source file'
C
TYPE 105,CFILE,ASP(IASP)," DEG ",SHIP(ISHIP)
PRINT 105,CFILE,ASP(IASP)," DEG ",SHIP(ISHIP)
C
TYPE 106,LINE2
C
TYPE 106,PARAM
105 FORMAT(X,A6O,A2,A5,A2)
106 FORMAT(X,3OA2)
C
C_TYPE 1:2=SHIP(ISHIP){1:2}
C_TYPE 1:2 =ASP(IASP){1:2}
C
C
GET NUMERICAL INFORMATION FROM THE THIRD LINE
OF THE HEADER
C
CALL DDDE(NDIM,ANGT,AINC)
C
DIVIDE AN AMP-PHASE ARRAY INTO
AN AMP ARRAY AND A PHASE ARRAY
C

332
DO 189 NN=1, NDIM
AM(NN)=AP(NN)
189 IF(AM(NN).GT.40)AM(NN)=40.
PH(NN)=AP(NN)

C
C CHECK FOR BAD DATA POINTS, I.E., AM(I).GT.995
C
CALL ERRF(AM, PH, NDIM)
RETURN
END

SUBROUTINE TRCINFILE
COMMON BUFF, NDIM, ANST, AINC
INTEGER*2 INFILE(15)
CHARACTER *10 INFILE
DIMENSION AM(1601), PH(1601)
BYTE BUFF(33200), TBUFF(512)
INCLUDE 'SYSLIBRARY: FORIOSDEF'
IB=1
ICNT=0

OPEN(UNIT=8, NAME=INFILE, READONLY, TYPE='OLD', IOSTAT=IERR, ERR=8100)

IF(IB.EQ.1)LEN=512-8*4
IF(IB.GT.1)LEN=512-2*B*4

READ(I, 80, END=80) TBUFF
READ(8, 80, END=80) TBUFF

FORMAT(512A1)

BUFF(ICNT+I)=TBUFF(I)
IB=IB+1
ICNT=ICNT+ LEN
GO TO 82

DO 86 I=1, LEN
BUFF(ICNT+I)=TBUFF(I)
GO TO 391

ELIMINATE BLANK SPACES IN BETWEEN EACH CHARACTER
IN A FILE HEADER

DO 40 I=1, IB0
BUFF(I)=BUFF(2*I-1)
GO TO 331

8100 IF(IERR.EQ.0, FORMAT_11000) THEN
TYPE 1112, INFILE
 ELSE IF(IERR.EQ.0, FORMAT_11000) THEN
TYPE 1113, INFILE

333
1113 FORMAT(' FILE : ',A10, ' WAS BAD, ENTER NEW FILENAME')
ELSE
  TYPE *, 'UNRECOVERABLE ERROR, CODE=', IERR
  STOP
ENDIF
ACCEPT 1114, INFILE
1114 FORMAT(A10)
GO TO 8106
331 CLOSE(UNIT=8, DISP='SAVE')
RETURN
END

C

SUBROUTINE DCDE
COMMON Buff, NDIM, ANST, AINC
DIMENSION AM(1601), PH(1601)
BYTE Buff(33200)
INTEGER*4 IMIN, IINC, NDIM
C
NO OF DATA POINTS IS STORED IN FOUR CHARACTERS, AND
STARTING ANGLE AND ANGLE INC. IN 5 CHARACTERS
C
CHARACTER*3 CNL1
CHARACTER*4 CNL
CHARACTER*5 CFF, CINC
CHARACTER*1 EQH, TCAS
DATA CNL(1) = '0', 'A'/
EQUIVALENCE (BUFF(123), CNL), (BUFF(131), CFF), (BUFF(140), CINC)
EQUIVALENCE (BUFF(123), TCAS), (BUFF(124), CNL1)
C
CONVERT CHARACTERS INTO THEIR NUMERICAL EQUIVALENTS
IF(ECH.EQ.TCAS) THEN
  DECODE(3, 102, CNL1) NDIM
ELSE
  DECODE(4, 101, CNL) NDIM
ENDIF
100 FORMAT(15)
101 FORMAT(14)
102 FORMAT(13)
DECODE(5, 100, CFF) IMIN
DECODE(5, 100, CINC) IINC
ANST = FLOAT(IMIN)
AINC = FLOAT(IINC)
RETURN
END

C

SUBROUTINE ERRF(AM, PH, NDIM)
DIMENSION AM(1601), PH(1601)
COMPLEX C1, C2, CD
DO 1 I = 1, NDIM
IF(PH(I).GT.90.) WRITE(6,2) I,AM(I),PH(I)
1 CONTINUE
2 FORMAT(I1X,16HEXERROR AT DATA PT,1I4,4H1MAG=,1F10.4,4H1PHI=,1F10.4)
C CHECK LEFT END POINT
IF(AM(I).GT.100.) THEN
  DO 200 I=2,NDIM
  IF(AM(I).LE.100. .AND. AM(I+1).LE.100.) THEN
  A1=10.**(AM(I)/20.)
  C1=CMPLX(A1*COSD(PH(I)),A1*SIND(PH(I)))
  A2=10.**(AM(I+1)/20.)
  C2=CMPLX(A2*COSD(PH(I+1)),A2*SIND(PH(I+1)))
  CD=C1-C2
  AD=AIMAG(CD)
  DO 212 II=1,I-1
  RD=REAL(CD)+AD*II
  AC=AIMAG(CD)+AD*II
  AM(II)=20.*LOG10(SQRT(RC*RC+AC*AC))
  PH(II)=ATAN2D(AC,RC)
  212 CONTINUE
  GO TO 200
  ELSE
  END IF
  END IF
C CHECK RIGHT END POINT
211 IF(AM(NDIM).GT.100.) THEN
  DO 220 I=1,NDIM
  J=NDIM-I
  IF(AM(J).LE.100. .AND. AM(J-1).LE.100.) THEN
  A1=10.**(AM(J)/20.)
  C1=CMPLX(A1*COSD(PH(J)),A1*SIND(PH(J)))
  A2=10.**(AM(J-1)/20.)
  C2=CMPLX(A2*COSD(PH(J-1)),A2*SIND(PH(J-1)))
  CD=C1-C2
  AD=AIMAG(CD)
  DO 222 II=J+1,NDIM
  RD=REAL(CD)+AD*(II-J)
  AC=AIMAG(CD)+AD*(II-J)
  AM(II)=20.*LOG10(SQRT(RC*RC+AC*AC))
  PH(II)=ATAN2D(AC,RC)
  222 CONTINUE
  GO TO 220
  ELSE
  END IF
  END IF
C CHECK INTERIOR POINTS
221 DO 230 I=2,NDIM-1
  IF(AM(I).GT.100.) THEN
  DO 240 K=I+1,NDIM
  IF(AM(K).LE.100.) THEN

335
A1=10.**(AM(I-1)/20.)
C1=CMPLX(A1*COSD(PH(I-1)),A1*SIND(PH(I-1)))
A2=10.**(AM(K)/20.)
C2=CMPLX(A2*COSD(PH(K)),A2*SIND(PH(K)))
CD=(C1-C2)/(K-I+1)
RD=REAL(CD)
AD=AIMAG(CD)
DO 241 II=I,K-1
  RC=REAL(C1)+RD*(II-I+1)
  AD=AIMAG(C1)+AD*(II-I+1)
  AM(II)=2D.*LOG10(SQRT(RC*RC+AC*AC))
  PH(II)=ATAN2D(AC,RC)
241 CONTINUE
  GO TO 230
ELSE
  END IF
240 CONTINUE
ELSE
  END IF
230 CONTINUE
RETURN
END
C THIS IS THE END
C
C                                                                                       

Program name: TAFCAL.FOR

Time And Frequency Classification Algorithm (TAFCAL)

Originally developed by J. Chen
in two parts as FREQ.FOR and TIME.FOR
Modified by N. Chamberlain May 1984 to incorporate
classification at various polarizations, elevations and an
extended range of aspects (with the capability of intro-
ducing known errors in aspect or elevation)

COMPLEX C(110,30), NC(110,30), FFT(32)
REAL D(110,110), A(110,30), P(110,30), W(110,30)
REAL AN(110,30), PN(110,30), WN(110,30)
REAL ML(30), PX(30), PY(10,30), PYT(30), SAP(110)
REAL SP(110), SN(110), Y_PLOT(10,50), FFT(8)
INTEGER IPOL(5), IELEV(3), SNR, SNR_MAX, SNR_MIN
CHARACTER POL(5)*1, ELE*5, TLNE*5, T_CHAR*4, TCON_EA(10)*1, TCON_EL(10)*1
CHARACTER LINE(11)*70, PAR_CH*2, PAR(10)*2, AYN(10)*1, YN*1
CHARACTER CON_EA*4, CON_EL*1

COMMON /BLK1/ IPOL, IELEV, NS, C, A, P, IRA, ID, IGRAPH
COMMON /BLK2/ FFT, FFTT

Initiate header variables with blanks, so they can be searched

DO I=1,70
LINE(I)(J:J)= ' ' END DO
END DO

IGRAF=1 ! Index for counting graphs
IER_AG=0 ! Initialize error in aspect index
IBIG_LOOP=1 ! This index enables the total number of runs, frequency and/or time to counted for plotting

Start input routine

TYPE "," Default to target directory
TYPE "," Do frequency before time if doing both
TYPE "," Print classification distances ? Y or N
ACCEPT 3, DIS
FORMAT (A1)
IF (DIS.EQ.'Y') THEN
IDIS=1
ELSE
IDIS=0
END IF

TYPE "," Input 1 OR 0 for new normalizing constant in AN feature
ACCEPT *, INR

Parameters are entered by means of subroutines in order to make the
changing of parameters simpler and more flexible later on in program

CALL DOM(IDFT1, IDFT2, IQ, IGRAPH, &22) \ Select classification domain
CALL PZ(JPOL, LINE, IQ, IGRAPH, &1) \ I/P polarization parameters
CALL ELN(IJELEV, LINE, IQ, IGRAPH, &9) \ I/P elevation parameters
LINE(I:30) = ELEV ASSUMED KNOWN
IF(IJELEV EQ 2) CALL EURI(LINE, IQ, IGRAPH, &10) \ a priori ele know
IF(IJELEV .GT. 1 .AND. IERAS .NE. 0) CALL ERE(IERAS, LINE, IQ, IGRAPH, &22)
CALL ASPI(NASPA, MAXASP, INCASP, LINE, IQ, IGRAPH, &4) \ a priori asp.
CALL AK(IAK, LINE, IQ, IGRAPH, &5) \ select asp. knowledge
IF(IAK .GT. 1 .AND. IERAS .NE. 0) CALL ERA(IERAS, LINE, IGRAPH, &6)
CALL FRE(FMIN, FINC, NF, LINE, IQ, IGRAPH, &7) \ select frequencies
IF(IDFT2 .NE. 0) CALL RALL(LINE, IQ, IGRAPH, &8)
CALL NOM(I3, L36, NEX, IQ, IGRAPH, &554) \ select no expt & r.n seeds
IF(IDFT2 .NE. 1) CALL NAG(JNAG, NAG, &555)
CALL FEY(IY, IY, IR, IQ, IGRAPH, &656)

End of input routine  

Call data files  

Estimate signal powers for freq. and time classification

PAVE=0.
DO 910 I=1, NS
DO 910 J=1, NF
PAVE=PAVE+A(I,J)*A(I,J)
CONTINUE

PAVE=PAVE/(FLOAT(NF)*FLOAT(NS)) \ Pave=sum(A**2/ns/nf)
PAVE=10.**ALOG10(PAVE) \ dB
IN_PAVE=INT(PAVE) \ convert to nearest integer
IN_MIN=IN_PAVE-30 \ +/- 30 dB is a useful range
IN_MAX=IN_PAVE+30 \ for classification
IN_INC=10  \ this gives us 7 values of S/N

IF(IBIG_LOOP .AND. IGRAPH .EQ. 1) THEN \ first run
TYPE *, ' 
TYPE *, 'Ave power [dB] = ', IN_PAVE
TYPE *, 'Suggested MIN, MAX, INC of noise power for +/- 30 dB S/N '
TYPE *, IN_MIN, IN_MAX, IN_INC

Specify minimum, maximum and increment of
additive gaussian noise powers, once signal power is known

TYPDE *, ' 
TYPE *, 'Input MIN, MAX, INC of the noise power [dB]'
ACCEPT *, IV1, IV2, IVT

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IV1=IN_MIN
IV2=IN_MAX
IVT=IN_INC
SNR_MAX=INT(PAVE)-IV1  \ I Find limits of SNR
SNR_MIN=INT(PAVE)-IV2
ELSE
IV1=IN_PAVE+SNR_MIN  \ Calculate Noise on subsequent runs to
IV2=IN_PAVE+SNR_MAX  \ provide constant SNR range
ENDIF

TOTAL
IF(I0.FT1.ED.1) THEN  \ IF(ITF2.ED.1) GOTO TIME DOMAIN
V
ELSE
V
ENDIF

DO 98237 I=1,NF
98237 WL(I)=300./((FMHZ(I)+1)*FINC)
C Compute wavelengths

PI=4.*ATAN(I.)
TWOP=2.*PI
NF=NF-1
C
DO 1434 J=1,NF-1
IF([RA.ED.1) A(I,J)=A(I,J)/A(I,J+1)  \ Compute relative amplitude
C
PC=P(I,J)
DP=P(I,J)-P(I,J+1)
IF(DP.GT.PI)PC=PC-TWOP
IF(DP.LT.-PI)PC=PC+TWOPI
C
W(I,J)=WL(J)*PC-WL(J+1)*P(I,J+1)  \ Compute relative phase
C
IF([RA.ED.0)CALL VARN(A,NF,NS,VA,1)  \ Estimate variances of amplitudes
IF([RA.ED.1)CALL VARN(A,NF1,NS,VA,1)  \ Estimate variances of W's
C
RK=SQRT(VA/(RR*VW))  \ Calculate normalization constant RK
C
PRINT OUT
PRINT *, 'No of targets = ',NS
PRINT *, 'VAR(AMP) = ',VA
PRINT *, 'VAR(W) = ',VW
PRINT *, 'Ave. signal power (dB) = ',PAVE
PRINT 18221,RR
18221 FORMAT(1X,'VAR(A)/VAR(W) = ',E12.5)
PRINT 18222,IVW,IVA
18222 FORMAT(' IVW = ',I1,' IVA = ',I1)
PRINT *,'
C
ELSE  \ TIME DOMAIN
C
CALL FORT(FFFT,5,SFFT,0,IFERR)

Calculate square roots of total signal powers

DO 88274 I=1,NS
SP(I)=0
DO 8824 J=1,NF
  SP(I)=SP(I)+ABS(C(I,J))**2
8824
SP(I)=SORT(SP(I))

PRINT *, 'No of targets ',NS
PRINT *, 'Ave signal power (dB) = ',PAVE

C

END

CALL Calculate 90% confidence interval for 30% error
calculate into header line
C
C
5757 CON_IN=165.0*SQRT(0.7*0.3/NS/NEX)
IF(IG.EQ.1.AND.IGRAPH.GT.1) THEN
  DO I=1,70
    LINE(I:I)="'
  END DO
END IF
IF(IG.EQ.1) THEN
  ENCODE(29,1038,LNE(9))CON_IN
END IF
1038 FORMAT('90% CI (95%) +/- ','F3.1','%')
ELSE
  DO I=70,20,-1
    IF(LINE(I:I).NE."") THEN
      LS=I
      GOTO 52
    END IF
  END DO
52 ENCODE(4,1037,LNE(9)(LB+5:LB+8))CON_IN
1037 FORMAT('F3.1','%')
END IF

C Encode number of targets into a header line
C
IF(IG.EQ.1.AND.IGRAPH.GT.1) THEN
  DO I=1,70
    LINE(I:I)="'
  END DO
END IF
IF(IG.EQ.1) THEN
  ENCODE(21,1040,LNE(9))NS
END IF
1040 FORMAT('NO OF TARGETS ','I6')
ELSE
  DO I=70,21,-1
    IF(LINE(I:I).NE."") THEN
      LS=I
      GOTO 52
    END IF
  END DO
52 ENCODE(4,1037,LNE(9)(LB+5:LB+8))
1037 FORMAT('F3.1','%')
END IF


ENCODE(5,1039,TLINE)NS
LINE(8)(L8+6:L8+10)=TLINE(1:5)
END IF

ENCODE(15,1039,TLINE)NS
LINE(8)(L8+6:L8+10)=TLINE(1:5)
END IF

 Encode classification features into a header

 IF(IYA.EQ.D.AND.IYW.ER.Oj THEN
TLINE(1:4)=TLINE(1:4)="W"
ELSE IF(IYA.51.1.AND.IYW.Em.0) THEN
TLINE(1:4)=TLINE(1:4)="W"
ELSE IF(IYA.51.1.AND.IYW.Em.0) THEN
TLINE(1:4)=TLINE(1:4)="W"
ELSE IF(IYA.51.1.AND.IYW.Em.0) THEN
TLINE(1:4)=TLINE(1:4)="W"
END IF

IF(I2a...AND.IGRAPH.6T.1) THEN
DO I=1,70
LINE(10)(I:I)="I"
END DO
END IF

IF(I2a...AND.IGRAPH.6T.1) THEN
DO I=1,70
LINE(10)(I:I)="I"
END DO
END IF

IF(I2a...AND.IGRAPH.6T.1) THEN
DO I=1,70
LINE(10)(I:I)="I"
END DO
END IF

Vary the noise power from IV1 to IV2 dB

J_COUNT=NINT(FLOAT(IV2-IV1)/FLOAT(IVT))+1
TYPE *, *
TYPE *, Count down to and of classification run'
IRMIS=0

DO BD IV=IV1,IV2,IVT
IRMIS=IRMIS+1
TYPE *,J_COUNT
J_COUNT=J_COUNT-1
RNIS=0  \* Initialize the number of false alarms

DO 888 IEX=1,NEX  \* Perform classifications NEX times

IF(IDIS.EQ.1)PRINT *, '-------------------------------------------------------------------'

Add noise to the original data files to generate test targets

VM=10.**(FLOAT(IY)/10.) \* 1 dB to nepers

CALL MN(VM,IS3,IS4,C,NC,NF,NS,IYA,IYW,IEX,SAP,SP) \* \* CALL MN

IF(IQFT1.EQ.1) THEN \* \* \* IF(IQFT2.EQ.1) THEN GOTO TIME DOMAIN CLASS.

Calculate the noisy amplitude and phase returns

DO 821 I=1,NS
DO 822 J=1,NF
AN(I,J)=ABS(NC(I,J))
PC=AN(I,J)
DP=AN(I,J)-PI
IF(DP.GT.PI)PC=PC-TWOPI
IF(DP.LT.-PI)PC=PC+TWOPI
WM(I,J)=WM(I,J)*PC-WL(I,J)-PI
IF(IPM.EQ.0)GOTO 398

Calculate the noisy W's

DO 821 I=1,NS
DO 822 J=1,NF-1
PC=PN(I,J)
DP=PN(I,J)-PI
IF(DP.GT.PI)PC=PC-TWOPI
IF(DP.LT.-PI)PC=PC+TWOPI
WM(I,J)=WM(I,J)*PC-TH(I,J)*PM(I,J)

Calculate the noisy relative amplitudes

DO 349 IR1=1,NS
DO 349 IR2=1,NF-1
AM(IR1,IR2)=AN(IR1,IR2)/AN(IR1,IR2+1)

Compute the nearest neighbour distances between any two classes

CALL DIST_F(IYA,IYW,IYA,A,W,AN,WM,NS,RR,RK,D,JNAJ,ANAG)

342
ELSE \------------------------------------------------------------------------------------------------------------------------------------- TIME DOMAIN CLASS.

Calculate the square roots of the total powers of the noisy targets

DO 8221 I=1,N
SNP(I)=0
DO 8212 J=1,NF
SNP(I)=SNP(I)+ABS(NC(I,J))**2
8212 CONTINUE
8221 SNP(I)=SQRT(SNP(I))

Compute cross coefficients between any two targets

NB=INT(FMIN/FINC)

CALL DIST_T(C,NC,SP,SNP,NF,NS,D,NB)

END IF

Do classification and compute number of false alarms

If the aspects are assumed known then
consider only those classes whose aspect angles
are the same as that of the target, otherwise
consider all classes

KASP=(MAXASP-MINASP)/INCASP+1 ! No of aspect angles
IF(IAK.EQ.0)KASP=1 ! aspects when aspect is known

Find the nearest neighbour to the test targets
or a class which has the max correlation coefficient
with a target

IF(IEK.EQ.1.AND.IELEV(3).GT.1) THEN ! Elevation known
If elevation is known and there is more than 1 elevation, then
consider only those classes whose elevations are known

IPP=2 ! Elev known
ELSE
IPP=1 ! Elev unknown
END IF

DO IP=1,IPP ! Split target array into 2 parts if elev known

IKX1=NS*(IP-1)/2+1 ! --> ( 1 then NS/2 + 1 ) or 1
NS1=NS*IP/IPP ! --> ( NS/2 then NS ) or NS

DO 7263 IKX=1,KASP ! 1--No of aspects
IEPO=1
IK1=IKX+IKX1-1
7171 DO 740 IT=IK1,NS1,KASP ! 1,2,--NS/2 & NS/2+1,2,--NS
DIN=10000000000.
DIN=10.

IF(IER_EL.EQ.1) THEN
  NS2=NS/IP
  IKX2=[IP-IP]*NS/2+1
ELSE
  NS2=NS1
  IKX2=IKX1
END IF

IK2=IKX+IKX2-1
DO 860 IU=IK2,NS2,KASP
  Index for searching noiseless targets

IF(IER_EL.EQ.1) THEN
  Introduce a deliberate error in aspect. There must be 2 or 3 aspects to do this. If last shift up by INCASP, if last shift down by INCASP, if middle shift up and down by INCASP. This simulates an error of +/- INCASP degrees.
  IF(IKX.EQ.1) I=IU+1
  IF(IKX.EQ.3.OR.(KASP.EQ.2.AND.IKX.EQ.2)) I=IU-1
  IF(IKX.EQ.2.AND.KASP.EQ.3) I=IU+(-1)**IEXPD
ELSE
  I=IU
END IF

D(I,IT) is the distance between noiseless target I and noisy target IT

IF(D(I,IT).GE.DMIN) GO TO 860
DMIN=D(I,IT)
IMIN=I
ELSE
IF(D(I,IT).LE.DMAX) GO TO 860
DMAX=D(I,IT)
IMAX=I
END IF

860 CONTINUE

If a target is misclassified increase the number of false alarms by one

IF(IER_EL.EQ.1) THEN
If a deliberate error is made in aspect, alter IMIN,IMAX to account for this. So if algorithm picks S1 @ 10 deg as the closest neighbour to S1 @ 0 deg, then this is considered a correct classification.
IF(IQFT1.EQ.1) IMINMAX=IMIN
IF(IQFT2.EQ.1) IMINMAX=IMAX
IF(IKX.EQ.1) IMINMAX=IMIN-1
IF(IKX.EQ.3.OR.(IKX.EQ.2.AND.KASP.EQ.2)) IMINMAX=IMINMAX+1
IF(IKX.EQ.2.AND.KASP.EQ.3.AND.IEXPD.EQ.1) IMINMAX=IMINMAX+1
IF(IIXE.2.AND.KASP.EQ.3.AND.IEXP0.EQ.2) IMINMAX=IMINMAX+1
IMIN=IMINMAX
IMAX=IMINMAX
END IF

C
IF(IER_EI.EQ.1.AND.IEK.EQ.1) THEN

C Allow for deliberate error made in elevation in the same way as was
done for aspect

C IF(IQFT1.EQ.1) IMINMAX=IMIN
IF(IQFT2.EQ.1) IMINMAX=IMAX
IF(IP.EQ.1) IMINMAX=IMINMAX-NS/2
IF(IP.EQ.2) IMINMAX=IMINMAX+NS/2
IMIN=IMINMAX
IMAX=IMINMAX

C END IF
C
IF(IDIS.EQ.1.AND.IDIS.EQ.1)PRINT *, 'TARGET :',IT, ' MIN = ',IMIN
IF(IDIS.EQ.1.AND.IDIS.EQ.1)PRINT *, 'TARGET :',IT, ' MAX = ',IMAX
IF(IDIS.EQ.1)PRINT 730,(NINT(D[I1,IT]),I1=IK2,NS2)
C IF(IDIS.EQ.2)WRITE(1,730)(D[I1,IT],I1=IK2,NS2)
C
730 FORMAT(1X,38IS,/)C

C IF(IMIN.NE.IT.AND.IQFT1.EQ.1)RMIS+1.
C Increment errors
C IF(IMIN.NE.IT.AND.IQFT2.EQ.1)RMIS+1.
C CONTINUE
C
IF(IIXE.2.AND.IER_EI.EQ.1.AND.IEXP0.EQ.1.AND.KASP.EQ.3) THEN
IEXP0=2
GOTO 7171
! If no of asps = 3, repeat for middle aspect
END IF
C
7171 CONTINUE
! number of aspects loop
END DO
! number of elevations loop
C
888 CONTINUE
! number of experiments loop
C
C Calculate the maximum likelihood estimate
C of the error probability
C (Account for extra classifications done when there are 3 aspects
C and a deliberate error is introduced.)
C
C IF(IER_EI.EQ.1.AND.KASP.EQ.3) RMIS=RMIS/FLOAT(NEX*[NS+6])**100.
C IF(IER_EI.EQ.0.OR.KASP.EQ.2) RMIS=RMIS/FLOAT(NEX**NS))**100.
C
C SNR=INT(PAVE)-IV
! Signal to noise ratio
C IF(IDIS.EQ.0) PRINT 1113,SNR,RMIS
C
1113 FORMAT(1 SIGNAL TO NOISE RATIO = ',I3,1; % ERRORS = ',F5,1)
C
Y_PLOT(IBIG_LOOP,J,COUNT+1)=RMIS
! plotting array
C IF(IV.EQ.2) THEN
PRINT *, ' *********************************************
PRINT *, ' END IF

345
CONTINUE

IF(IER_AS.EQ.1) THEN
  IF(IQ.EQ.1.OR.IGRAPH.EQ.1) THEN
    TYPE *, 'Continue to have error in aspect? Y or N'
    ACCEPT 12986, CONERA
  END IF
  IF(CONERA.EQ.'N') IER_AS=0
  END IF

IF(IER_EL.EQ.1) THEN
  IF(IQ.EQ.1.OR.IGRAPH.EQ.1) THEN
    TYPE *, 'Continue to have error in elevation? Y or N'
    ACCEPT 12986, CONERE
  END IF
  IF(CONERE.EQ.'N') IER_EL=0
  END IF

IF(YN.EQ.'S'.OR.YN.EQ.'G'.OR.YN.EQ.'R') GOTO 1117

12986 FORMAT(A1)
TCONERA(IQ:1:1)=CONERA(I:1)  ! Store previous entries for subsequent runs
ELSE
  CONERA(I:1)=TCONERA(IQ:1:1)
END IF
IF(CONERA.EQ.'N') IER_AS=0
END IF

1115 FORMAT(A1)
AYN(IQ)=YN
ELSE
  YN=AYN(IQ)
END IF

IF(YN.EQ.'S'.OR.YN.EQ.'G'.OR.YN.EQ.'R') GOTO 1117

1117 "Routine for changing parameters"

IF(IGRAPH.EQ.1.OR.IQ.GT.INUMG) THEN
  TYPE *, 'Type P to change current parameters'
  TYPE *, 'Type R to do new graph retaining old parameters'
  TYPE *, 'Type G to do new graph with new parameters'
  ACCEPT 1115, YN
END IF

IF(AYN.EQ.'S'.OR.AYN.EQ.'G'.OR.AYN.EQ.'R') GOTO 1117

"Routine for changing parameters"

IF(IGRAPH.EQ.1.OR.IQ.GT.INUMG) THEN
  TYPE *, 'Type P to change polarization'
END IF
TYPE *, , 'AM to change rel/ amplitude'
TYPE *, , 'FR to change frequencies'
TYPE *, , 'EL to change elevations'
TYPE *, , 'EK to change un/known elev'
TYPE *, , 'EE to change no/error in elev'
TYPE *, , 'AS to change aspects'
TYPE *, , 'AK to change un/known aspect'
TYPE *, , 'AE to change no/error in aspect'
TYPE *, , 'NE to change number of expts'
TYPE *, , 'DO to change domains'
TYPE *, , 'FE to change NN features'
TYPE *, , 'NA to use Nails algorithm'

C
1118 TYPE *, , 'Input change'
ACCEPT 1116,PAR_QH

1116 FORMAT(A2)
PAR(IQ)(1:2)=PAR_QH(1:2)
ELSE
PAR_QH(1:2)=PAR(IQ)(1:2)
END IF
C
IF(PAR_QH, EQ, 'PO') CALL FZM(IPOL,LIN, IQ1, IGRAPH, &5656)
IF(PAR_QH, EQ, 'FR') CALL FRE(FMIN, FINC, NF, IQ, IGRAPH, &5656)
IF(PAR_QH, EQ, 'AM') CALL RAL(INA, IQ, IGRAPH, &5656)
IF(PAR_QH, EQ, 'EL') THEN
CALL ELN(IELEV, IQ, IGRAPH, &7070)
7070 IF(IELEV(3).EQ.2) CALL EKJ(IEK, IQ, IGRAPH, &7072)
7072 IF(IEK.EQ.1.AND.IER_AS.EQ.0) CALL ERE(IER_EL, IQ, IGRAPH, &5656)
GOTO 5856
END IF
IF(PAR_QH, EQ, 'EK') THEN
CALL EKI(IEK, IQ, IGRAPH, &7074)
7074 IF(IEK.EQ.1.AND.IER_AS.EQ.0) CALL ERE(IER_EL, IQ, IGRAPH, &5757)
GOTO 5757
END IF
IF(PAR_QH, EQ, 'EE'.AND.IAK.EQ.1) THEN
CALL ERA(IER_EL, IQ, IGRAPH, &5757)
ELSE IF(PAR_QH, EQ, 'EE'.AND.IAK.EQ.0) THEN
TYPE *, 'Elevation must be known in order to introduce error'
GOTO 1118
END IF
IF(PAR_QH, EQ, 'AK') THEN
CALL AKJ(IAK, IQ, IGRAPH, &7076)
7076 IF(IAK.EQ.1.AND.IER_EL.EQ.0) THEN
CALL ERA(IER_AS, IQ, INCASP, IGRAPH, &5757)
GOTO 5757
END IF
IF(PAR_QH, EQ, 'AE'.AND.IAK.EQ.1.AND.IER_EL.EQ.0) THEN
CALL ERA(IER_AS, IQ, INCASP, IGRAPH, &5656)
ELSE IF(PAR_QH, EQ, 'AE'.AND.IAK.EQ.0) THEN
TYPE *, 'Aspect must be known in order to introduce error'
GOTO 1118
END IF
IF(PAR_CH.EQ.'AS') THEN
CALL ASP(NINASP, MAXASP, INCASP, LINE, IQ, IGRAPH, &7078)
END IF

IF(NINASP.NE.MAXASP) THEN
CALL ARI(IAK, LINE, IQ, IGRAPH, &7080)
END IF

IF(IAK.EQ.1.AND.IER_EL.EQ.0) THEN
CALL ERA(IER_A3, LINE, IQ, INCASP, IGRAPH, &72)
END IF

END IF
GOTO 5556
END IF

IF(PAR_CH.EQ.'NE') CALL NDS(IS3, IS4, NEX, IQ, IGRAPH, &5757)
END IF

IF(PAR_CH.EQ.'FE') CALL FEA(IYA, IYW, RR, IQ, IGRAPH, IB, LINE, &505)
END IF

IF(IPFT1.EQ.1) CALL FE(IYA, IYW, RR, IQ, IGRAPH, IB, LINE, &505)
END IF

IF(PAR_CH.EQ.'NA'.AND.IPFT1.EQ.1) CALL NAG(JNAS, ANAG, &5757)

CALL IXN(FTI, IQFT2, IQ, XC, QPH, &525)

IF(QFT2.BEQ.1) STOP
END IF

END IF

END IF

GOTO 1118

Plotting routine

TYPE *, 'SNR_MIN', 'SNR_MAX', 'SNR_INC', 'SNR_MIN', 'SNR_MAX', IVT
NDS=IBIG_LOOP-1
CALL V_PLOT_Y_PLOT, NDS, IRNIS, LINE, IGRAPH, SNR_MIN, SNR_MAX, IVT

IF(YN.EQ.'S') THEN
STOP
END IF

IF(YN.EQ.'G') GOTO 11111

End of first graph... continue if desired

INUMG=IQ
IBIG_LOOP=1
IQ=1
IGRAPH=IGRAPH+1
DO I=1,70
LINE(111)(I:1)=" Type 
END DO

TYPE *, 'Type PO to change polarization'
TYPE *, 'AM to change rel/ampplitude'
TYPE *, 'FR to change frequencies'
TYPE *, 'EL to change elevations'
TYPE *, 'EK to change un/known elev'
TYPE *, 'EE to change no/error in elev'
TYPE *, 'AS to change aspects'
TYPE *, 'AK to change un/known aspect'
TYPE *, 'AE to change no/error in aspect'
TYPE *, 'NE to change number of expts'
TYPE *, 'FE to change features'
TYPE *, 'DO to change domains'
TYPE *, 'EX to exit from change session'
C

8000 TYPE *,"'
    TYPE *,'Input change'
    ACCEPT 1116,GRA_CH
C

   IF(GRA_CH.EQ.'EX') GOTO 5656
   IF(GRA_CH.EQ.'DD') CALL DDM(IQFT1,IQFT2,IQ,IGRAH,&8000)
   IF(GRA_CH.EQ.'CP') CALL CPN(IPOL,LINE,IQ,IGRAH,&8000)
   IF(GRA_CH.EQ.'FR') CALL FRE(FMIN,FINC,NF,LINE,IQ,IGRAH,&8000)
   IF(GRA_CH.EQ.'AM') CALL RAF(INR,IGRAH,&8000)
   IF(GRA_CH.EQ.'EL') THEN
     CALL ELN(IIELEV,LINE,IQ,IGRAH,&8000)
   END IF
   IF(IIELEV(3).EQ.2) CALL ENL(IIEK,LINE,IQ,IGRAH,&8020)
   8020 IF(IIEK.EQ.1 .AND.IER-AS.EQ.0) CALL ERE(IER-EL,LINE,IQ,IGRAH,&8000)
     GOTO 8000
   END IF
   IF(GRA_CH.EQ.'EK') THEN
     CALL ENL(IIEK,LINE,IQ,IGRAH,&8030)
   8030 IF(IIEK.EQ.1 .AND.IER-AS.EQ.0) CALL ERE(IER-EL,LINE,IQ,IGRAH,&8000)
     GOTO 8000
   END IF
   IF(GRA_CH.EQ..'EE'.AND.IIEK.EQ.1) THEN
     CALL ERE(IER-EL,LINE,IQ,IGRAH,&8000)
   ELSE IF(GRA_CH.EQ..'EE'.AND.IIEK.EQ.0) THEN
     TYPE *,'Elevation must be known in order to introduce error'
     GOTO 8000
   END IF
   IF(GRA_CH.EQ.'AK') THEN
     CALL AAE[IAX,LINE,IQ,IGRAH,&8040)
   8040 IF(IAX.EQ.1 .AND.IER_EL.EQ.0) THEN
     CALL ERA(IER-AS,LINE,IQ,INCASP,IGRAH,&8000)
   END IF
     GOTO 8000
   END IF
   IF(GRA_CH.EQ..'AE'.AND.IAX.EQ.1 .AND.IER_EL.EQ.0) THEN
     CALL ERA(IER-AS,LINE,IQ,INCASP,IGRAH,&8000)
   ELSE IF(GRA_CH.EQ..'AE'.AND.IAX.EQ.0) THEN
     TYPE *,'Aspect must be known in order to introduce error'
     GOTO 8000
   END IF
   IF(GRA_CH.EQ.'AS') CALL ASP(MINASP,MAXASP,INCASP,LINE,IQ,IGRAH,&8000)
   IF(GRA_CH.EQ.'NE') CALL NOS(IS3,IS4,NEX,LINE,IGRAH,&8000)
   IF(GRA_CH.EQ.'FE') THEN
     IS=1
     CALL FEA(IYA,IYW,RR,IQ,IGRAH,I9,LINE,&8050)
   8050 IS=0
     GOTO 8000
   END IF
C
   STOP
   END
C
C
-----------------------------------------------------------------------------------
END OF TAFCAL.MAI
SUBROUTINES

This subroutine calculates the nearest neighbour distances between any two classes in the data base.

SUBROUTINE DIST_F (IRA, IYW, IYA, A, W, AN, WN, NS, RR, RK, D, JNAG, ANAG)

REAL D(110,110), A(110,30), W(110,30), AN(110,30), WN(110,30), D_T(20)
REAL MEAN, SIGMA

DO 20 I=1, NS
DO 20 IT=1, NS
D(I,IT)=0.
DO 20 K=1, NF-1

C

D_T(K)=0.0
C

IF(IYW.EQ.1) D(I,IT)=D(I,IT)+((W(I,K)-WN(IT,K))*RK)**2
IF(IYA.EQ.1) D(I,IT)=D(I,IT)+((A(I,K)-AN(IT,K))*RK)**2
IF(IYW.EQ.1) D_T(K) = (W(I,K)-WN(IT,K))*RK**2
IF(IYA.EQ.1) D_T(K) = (A(I,K)-AN(IT,K))*RK**2

CONTINUE
C

IF ABSOLUTE AMPLITUDES ARE USED THEN TAKE INTO ACCOUNT THE LAST FREQUENCY.

C

IF(IYA.EQ.1.AND.IRA.EQ.0)D(I,IT)=D(I,IT)+
(1*(A(I,NF)-AN(IT,NF))**2
C

IF(IYA.EQ.1.AND.IRA.EQ.0) THEN
D_T(K) = (A(I,NF)-AN(IT,NF))**2
L_E=NF
ELSE
L_E=NF-1
END IF
C

Eliminate bad distances.

C

find mean and variance.
C

D_TSUM=0.0
D_TSQA=0.0
DO H=1, L_E
C

D_TSUM=D_TSUM + D_T(M)
D_TSQA=D_TSQA + D_T(M)**2
END DO
C

MEAN = D_TSUM/L_E
SIGMA = SQRT((D_TSQA/L_E)-(MEAN)**2)

350
This subroutine calculates the time domain correlation coefficients for a set of noise contaminated targets.

SUBROUTINE DIST_T(C, NC, SP, SNP, NF, NS, D, NB)

REAL D(110,110), SP(110), SNP(110), S(8)
COMPLEX C(110,30), NC(110,30), RF(30), F(32)

DO 20 I=1,NS
    DO 20 IT=1,NS
        DO 23 K=1,NF
            RF(K)=C(I,K)*CONJG(NC(IT,K))
            CONTINUE
        END
    END
    CONTINUE
    CALL IDFT(RF,NF, NB, STM)
    CONTINUE
    CALL IFFT(STM)
    CONTINUE
    D(I,IT)=STM/(2.*SP(I)*SNP(IT))
    CONTINUE
RETURN
END

This subroutine calculates the time domain correlation coefficients for a set of noise contaminated targets.
This subroutine adds gaussian noise to the original
data files to generate test targets

SUBROUTINE NN(VR,IS3,IS4,C,NC,NF,NS,IYA,IYW,IEX,SAP,SP)

COMPLEX NC(110,30),C(110,30)
REAL SAP(110),SP(110)

The Gaussian noise is additive and has variance VR
and mean 0.

Equal amounts of noise are added to the real
and imaginary parts of the data points.

VR1=VR/2.

DO 10 I=1,NS
Reset the seeds for each test target.

IS3=IS3-4
IS4=IS4-6
DO 10 J=1,NF

Generate two Gaussian numbers.

CALL GAUSS(IS3,0.,VR1,S3)
CALL GAUSS(IS4,0.,VR1,S4)

Generate a test target

NC(I,J)=CMPLX(REAL(C(I,J))+S3,AIMAG(C(I,J))+S4)

If the phase returns are unknown, set them to zero.

IF(IYA.BLE.1.AND.IYW.BGE.0)NC(I,J)=CMPLX(ABS(C(I,J))+S3,S4)
RETURN
END

This subroutine reads a data file generated by
the CATLOG.FOR scaling program

SUBROUTINE DATA(MINASP,MAXASP,INCAPSP,FMIN,FMAX,FINC,NF,LINe)

COMPLEX C(110,30)
REAL*4 A(110,30),P(110,30),W(110,30),AM(500),PH(500)
CHARACTER *2 FPOL,FEL,FLNAM(5),ASP(23),AIR(4),SHIP(6)
CHARACTER INFILE*14
CHARACTER*(*),LINE(11)
INTEGER IPOL(5),IELEV(3)

COMMON /BLK1/ IPOL,IELEV,NS,C,A,P,IRA,IQ,GRAPH
DATA ASP/"00",'05','10','15','20','25','30','35','40','45',
          '50','55','60','65','70','75','80','85','90','95',
          '201','251','301','351','401','451','501','551','601',
          '651','701','751','801','851','901','951',
C
DATA SHIP/"AD","LR","LB","LX","KH","NO"/
DATA AIR/"B7","CD","C5","OC"/
C
ICBASE=0  ! Counter to count each file accessed
C
IF(IQ. M.1 AND. IGRAPH.BM.1) THEN
  IAS=-1
ELSE IF(AQ.EU.'A13 THEN
  IAS=0
ELSE
  GOTO 669
END IF
END IF
C
IF(IAS.EQ.0) I NST=4
IF(IAS.EQ.1) I NST=6  ! There are 6 ships available for classif.
C
DO 772 I=1,INST
  IF(IAS.EQ.0) IFLNAM(I)=AIR(I)  ! Assign target name to temp variable
  IF(IAS.EQ.1) IFLNAM(I)=SHIP(I)
C
IF(INCASPB.0) INCASP=1
IBC=(MAXASP-MINASP)/INCASP
C
ITPOLV=0
ITPOLH=1
ITPOLX=1
C
Start main loop for collecting target data

C
DO IP=1,IPOL(5)
  FPOL(1:1)="F"
  IF(IPOL(1).EQ.ITPOL.0) THEN
    FPOL(2:2)="V"
  ELSE IF(IPOL(2).EQ.ITPOL.1) THEN
    FPOL(2:2)="H"
END IF
C
353
ITPOLX=0
GOTO 5050
ELSE IF(IPOL(3)*ITPOLX.EQ.1) THEN
FPOL(2:2)="'X'
ITPOLX=0
ENDIF

5050
IEL15=1
IEL27=1

C
DO IE=1,IELEV(3)
  IF(IELEV(1)*IEL15.EQ.1) THEN
    FEL(1:2)="'15'
    IEL15=0
    GOTO 5050
  ELSE IF(IELEV(2)*IEL27.EQ.1) THEN
    FEL(1:2)="'27'
    IEL27=0
    END IF
C

5050
DO 40 I=1,NST
DO 111 IAP1=MINASP,MAXASP,INCASP
  IBASE=IBASE+1
  Convert aspect angle to an index so that file name can be formed
  IF(IAS,GT.1) THEN
    IF(IAP1.GT.90) THEN
      IF(IAP1.GT.100) IASP=21
      IF(IAP1.GT.170) IASP=22
      IF(IAP1.GT.180) IASP=23
    ELSE
      IASP=NINT(FLOAT(IAP1)/5)+1
    END IF
  ELSE
    IASP=IAP1/15+1
  END IF
  Compile filename: R:***FV#15
  ** = Target name
  FH 27
  ## = Target aspect
  FX
  F = Full scale
  V,H,X = Polarizations
  INFILE(1:2)="R:"
  INFILE(3:4)=FLNAM(I)
  INFILE(5:6)=FPOL
  INFILE(7:8)=ASP[IASP]
  INFILE(9:10)=FEL
  INFILE(11:12)=".D"
  INFILE(13:14)="AT"
  INFILE=INFILE(3:14)
C
TYPE 6060,INFILE(1:8)
C
PRINT 6060,INFILE(1:8)
6060
FORMAT(1X,A8)
CALL RT(INFILE,AM,PH,TFMIN,TFINC,GF)

Note the data files
are already scaled by the scale factor

IMAX=(FMAX-TFMIN)/TFINC
IMIN=(FMIN-TFMIN)/TFINC
IC=FINC/TFINC
NF1=0
DO 2111 J=IMIN,IMAX,IC

Convert amplitude from dcm**2 to nepers m**2
by antilogging and division by 100

NF1=NF1+1
A(IBASE,NF1)=10.**((AM(J)/20.)/100..
2111 P(IBASE,NF1)=PH(J)
111 CONTINUE
40 CONTINUE
END DO
END DO

End main loop for collecting target data

NS=IBASE
NS=Total number of targets = #pol x fasp x #elev x #target(#6 for ship)

Split up A,P matrix into 2 parts to allow V/H to be done

IF(IPOL(4).EQ.1) THEN
NS=NS/2
DO I=1,NS
DO J=1,NF1
AI,J=AI,J/A[I+NS,J]
P[I,J]=P[I,J]-P[I+NS,J]
END DO
END DO
END IF

DO 1366 I=1,NS
DO 1366 J=1,NF
1366 C[I,J]=CMPLX(A[I,J]*COS(P[I,J]),A[I,J]*SIN(P[I,J]))
IF(NF.LE.NF1)GO TO 1435
DO 1437 I=1,NS
P[I,NF]=0
1437 C[I,NF]=0
1435 RETURN
END

If it is not possible to supply all of the frequencies requested
fill the remaining array points with zeros

A[I,NF]=0
1437 C[I,NF]=0
1435 RETURN
END
This subroutine calculates the variance of a set of numbers

SUBROUTINE VARN(A, NF, NS, VA, INC)

REAL A(110,30)

RMEAN=0
DO 10 I=1,NS
DO 10 J=1,NF,INC
10  RMEAN=RMEAN+(NS*(NF/INC))
VA=0
DO 20 I=1,NS
DO 20 J=1,NF,INC
20  VA=VA+(A(I,J)-RMEAN)**2
VA=VA/(NS*(NF/INC))
RETURN
END

This subroutine generates a Gaussian random number

SUBROUTINE GAUSS(IS, AM, V, S)

A=0.0
DO 50 I=1,12
A=A+RANI(IS)
50  CONTINUE
S=(A-6.)*SQRT(V)+AM
RETURN
END

This subroutine calculates the inverse Fourier transform of a set of complex data points and finds the maximum of the results. It assumes that the data points are symmetric around zero frequency

SUBROUTINE IDIFT(RF, NF, NB, STM)

COMPLEX F(32),RF(50)
REAL S(8)

COMMON /BLK2/S,F

MP=32
DO 40 I=1,MP
F(I)=(0.,0.)
K=0
DO 50 I=NB+1,NB+NF
K=K+1
F(I)=RF(K)
50  F(NP+2-I)=CRNJG(RF(K))
CALL a standard IFFT in the scientific library

CALL FORT(F,5,6,2,IFERR)
IF(IFERR.NE.0)PRINT *, 'ERR IN IFFT'

Find the maximum of the outputs of the IFFT

600 IF(FREAL(F(I)).GT.STM)STM=FREAL(F(I))
RETURN
END

+------------------------------------------+

Originally developed by J.Chen
Modified by N.Chamberlain May 1984 for inclusion in TAPCAL

This subroutine plots the results of classifications.

SUBROUTINE V_PLOT(YT, NDS, NPT, LINE, IGRAPH, INX1, INX2, INX3)

REAL X(50), Y(50), YR(10,50)
CHARACTER*(*) LINE(11)

IF(IGRAPH.EQ.1) THEN
  TYPE 'INPUT MIN, MAX, AND INC FOR S/N'
  ACCEPT *, XMIN, XMAX, XINC
ELSE
  XMIN=FLOAT(INX1)
  XMAX=FLOAT(INX2)
  XINC=FLOAT(INX3)
END IF

NPT=(XMAX-XMIN)/XINC+1.
J=0
DO 100 FM=XMIN, XMAX, XINC
J=J+1
100 X(J)=FM

DSX=(XMAX-XMIN)/5.
DSY=100./8.
SPAX=5./FLOAT(NPT-1)
SPAY=6./10.

CALL VPLOTS(0,0,0)
CALL PLOT(1.5,1.5,-3)
CALL NEWPEN(3)
CALL AXIS(0.,0., 'POST-PROCESSING S/N IN DB', -25,5., 0., XMIN, DSX, SPAX, 0)
CALL AXIS(0.,0., 'MISCLASSIFICATION PERCENTAGE', 25,8.80,0., DSY, SPAY, 0)
CALL NEWPEN(1)
CALL GRID(0.,0., NPT-1, SPAX, 10, SPAY, '33333333')

DO 60 I=1,NDS
DO 70 J=1,NPT
Y(J)=YT(I,J)
60 CONTINUE
70 CONTINUE

357
CONTINUE
ISYM=1
CALL NEWPEN(3)
CALL STRYP(X,XMIN,DSX,Y,0.,DSY,NPT,0.5,ISYM)
C
DO 8876 I=1,11
IF(LINE(I)(1:10) .EQ. ' ') GOTO 8876
HEI=HEI-FL0AT(I-1)*0.2
ILEN=LEN(LINE(I))
CALL NEWPEN(2)
CALL SYMBOL(0.0,HEI,0.1,%REF(LINE(I)),0.0,ILEN)
8876 CONTINUE
C
CALL PLOT(0.0,0.0,998)
CALL PLOTHOW(IMS)
TYPE *, 'Plot again? Y or N'
ACCEPT 99,PAG
FORMAT(A1)
IF(PAG.EQ.'Y') GOTO 5
CALL LIB$SPAWN('8 RENAME [NFC4348.FSB]*.MET [NFC4348.MET]*.**
&,'NL:' , 'NL:')
RETURN
END
C
This subroutine inputs the required polarization parameters.
C
SUBROUTINE PZN(IPOL,LINE,IGRAPH,*
C
INTEGER IPOL(5)
CHARACTER POL(5)*1,TLINE*5,TPOL(10)*5
CHARACTER*(*) LINE(11)
C
DO I=1,5
IPOL(I)=0 ! Reset pol. control indices
END DO
C
IF(IGRAPH.EQ.1.OR.IGRAPH.1) THEN
TYPE *
TYPE *, 'Input polarization(s) V,H,X'
ACCEPT 7,POL
7 FORMAT(5(A1))
DO I=1,5
TPOL(IGRAPH)(I)=POL(I)
END DO
ELSE
DO I=1,5
POL(I)=TPOL(IGRAPH)(I)
END DO
END IF
C
Assign pol. control parameters
DO I=1,5  
IF(POL(I).EQ.'V') IPOL(1)=1  
IF(POL(I).EQ.'H') IPOL(2)=1  
IF(POL(I).EQ.'X') IPOL(3)=1  
TLINE(I:I)=POL(I)  
END DO  
IF(IQ.EQ.1.AND.IGRAPH.GT.1) THEN  
DO I=1,70  
LINE(2)(I:I)=' '  
END DO  
END IF  
IF(IQ.EQ.1) THEN  
LINE(2)(1:16)="POLARIZATION"  
LINE(2)(20:25)=TLINE(1:5)  
ELSE  
DO I=70,20,-1  
IF(LINE(2)(I:1).NE.' ') THEN  
L2=I  
GOTO 63  
END IF  
END DO  
END IF  
LINE(2)(L2+6:L2+10)=TLINE(1:5)  
END IF  
IF(POL(1).EQ.'V'.AND.POL(2).EQ.'X'.AND.POL(3).EQ.'H') IPOL(4)=1  
IPOL(5)=IPOL(1)+IPOL(2)+IPOL(3)  ! Check for valid inputs  
IF(IPOL(5).EQ.5,5,11  
TYPE *,'\'You have selected'  
TYPE *,'Vertical polarization'  
TYPE *,'Horizontal polarization'  
TYPE *,'Cross polarization'  
TYPE *,'V/H polarization'  
RETURN 1  
END  
359

This subroutine inputs the required elevation parameters.

SUBROUTINE ELN(IQELEV,LINE,IQ,IGRAPH,*  
INTEGER IQELEV,LINE,LINE,IQ,IGRAPH,*  
CHARACTER LINE*5,TLINE*5,TEL(10)*5

359
CHARACTER*(*) LINE(11)

IF(IGRAPH.EQ.'1',OR.IO.EQ.'1') THEN
  TYPE 'Input elevation angle(s) 15,27'
  ACCEPT 9,ELE
  FORMAT (AS)
  TEL(IQ)(1:5)=ELE(1:5)
ELSE
  ELE(1:5)=TEL(IQ)(1:5)
END IF

IF(IO.EQ.'1',AND.IGRAPH.GT.'1') THEN
  DO I=1,70
    LINE(4)(I:1)=I
  END DO
END IF

IF(IO.EQ.'1',AND.IGRAPH.GT.'1') THEN
  LINE(4)(1:19)=ELEVATION (DEG.)
  LINE(4)(20:25)=ELE(1:5)
ELSE
  DO I=70,20,-1
    IF(LINE(4)(I:1).NE.' ') THEN
      IEND_L=I
    END IF
  END DO
END IF

L4=IEND_L
LINE(4)(L4+5:L4+10)=ELE(1:5)
END IF

C
C Set elevation control parameters
C
  DO I=1,3
    ILEV(I)=0
  END DO

C
  IF(ELE(1:2).EQ.'15',OR.ELE(4:5).EQ.'15') ILEV(1)=1
  IF(ELE(1:2).EQ.'27',OR.ELE(4:5).EQ.'27') ILEV(2)=1
  ILEV(3)=IELEV(1)+IELEV(2)
  IF(IELEV(3)) 13,13,15

C
C Print out
C
  IF(IELEV(1).EQ.1) PRINT *, '15 deg. elevation'
  IF(IELEV(2).EQ.1) PRINT *, '27 deg. elevation'

C
RETURN 1

C

This subroutine inputs the required a priori information for elevation

SUBROUTINE EQUICK(LINE,IO,IGRAPH,*)
CHARACTER(*) LINE(11)
CHARACTER TEK(10)*1,EK*1

C
IF(IQ.EQ.1.OR.IGRAPH.EQ.1) THEN
  TYPE *, '
  TYPE *, 'Elevation known --> Type K'
  TYPE *, 'Elevation unknown --> Type U'
  ACCEPT 10, EK

  FORMAT(A1)
  TEK(IQ)(1:1)=EK(1:1)
  ELSE
  EK(1:1)=TEK(IQ)(1:1)
  END IF

C
IF(EK.EQ.'K') THEN
  IEK=1
  PRINT*, 'Elevation assumed known'
  ELSE IF(EK.EQ.'U') THEN
    IEK=0
    PRINT*, 'Elevation assumed unknown'
    ELSE
      IEK
      END IF
      GOTO 1
      END IF

C
IF(IQ.GT.1.AND.IEK.GT.0) THEN
  LINE(3)(1:3)=/'UNKNOWN'
  ELSE
    LINE(3)(1:3)=/'KNOWN'
    END IF
    RETURN
    END

C
This subroutine inputs error in elevation
SUBROUTINE ERE(IER_EL,LIN, IQ, IGRAPH, *)

CHARACTER(*) LINE(11)
CHARACTER TLIE(*),TERM(10)*1,ER_EL*1

C
IF(IQ.EQ.1.OR.IGRAPH.EQ.1) THEN
  TYPE *, '
  TYPE *, 'Type Y or N for error in elevation'
  ACCEPT 10, ER_EL

  FORMAT(A1)
  TERM(IQ)(1:1)=ER_EL(1:1)
  ELSE
  ER_EL(1:1)=TERM(IQ)(1:1)
  END IF

C
IF(ER_EL.EQ.'Y') THEN
IER_EL=1
PRINT *, 'Error of 15 deg introduced in elevation'
ELSE IF (IER_EL .EQ. 'N') THEN
IER_EL=0
PRINT *, 'No error in elevation angle'
ELSE
GOTO 1
END IF
C
IF (LINE[11][1:3].EQ. ',.AND.IER_EL .EQ. 1) THEN
 ENCODE(20,20,LIN[11])IQ
20 FORMAT('ELEV ERR IN CURVE', I3)
ELSE IF (LINE[11][1:3].NE. ',.AND.IER_EL .EQ. 1) THEN
 ENCODE(30,30,TLINE)IQ
30 FORMAT(I3)
DO I=70,16,-1
 IF (LINE[I][1:1].NE. ',) THEN
L11=I
 GOTO 40
END IF
END DO
END IF
C
RETURN 1
END
C
-------------------------------------------------------------------------------------
This subroutine selects frequency or time domain algorithms
C
SUBROUTINE DOM(IQFT1,IQFT2,IQ,IGRAPH,*)
C
CHARACTER IQFT(10)*1,QFT*l
C
IF (IQ.EQ. 1.OR.IGRAPH .EQ. 1) THEN
1 TYPE ' ',
 TYPE ' ', Input Classification domain F or T
 ACCEPT 2,QFT
2 FORMAT (A1)
IQFT[IQ][1:1]=QFT[1:1]
ELSE
QFT[1:1]=IQFT[IQ][1:1]
END IF
IF
IF (QFT .EQ. 'F') THEN
IQFT1=1
IQFT2=0
PRINT *, 'Classification in frequency domain'
ELSE IF (QFT .EQ. 'T') THEN
IQFT2=1
IQFT1=0
PRINT *, 'Classification in time domain'
ELSE
GOTO 1
END IF
END
This subroutine inputs required aspect angles.

SUBROUTINE ASP(MINASP, MAXASP, INCASP, LINE, IGRAPH, *)

CHARACTER TLINE*12
CHARACTER(*) LINE(11)
INTEGER THINA(10), TMAXA(10), TINCA(10)

Specify target aspect angles

IF(IO.EQ.1.OR.IGRAPH.EQ.1) THEN
  TYPE *, 'Input MIN, MAX, INC aspect angles'
  ACCEPT *, MINASP, MAXASP, INCASP
  THINA(IQ)=MINASP
  TMAXA(IQ)=MAXASP
  TINCA(IQ)=INCASP
ELSE
  MINASP=THINA(IQ)
  MAXASP=TMAXA(IQ)
  INCASP=TINCA(IQ)
END IF

IF(IO.EQ.1.AND.IGRAPH.GT.1) THEN
  DO I=1,70
    LINE(8)(I:1)=' ';
  END DO
  ENDDO
END IF
ENDE  

ENCOD(30,1020,LIN(6)) MINASP, MAXASP, INCASP

1020 FORMAT('MIN, MAX, INC ASPECT',3I4)

ELSE

ENCOD(12,1025,TLINE)MINASP, MAXASP, INCASP

1025 FORMAT(3I4)

DO I=70,30,-1
  IF(LIN(8)(I:1).NE. ' ') THEN
    IEND_L=I
    GOTO 53
  END IF
  END DO

53  LS=IEND_L
  LINE(6)(LS+1:LS+14)=', '
  LINE(6)(LS+2:LS+14)=TLINE(1:12)
END IF

PRINT 2849,MINASP, MAXASP, INCASP

2849 FORMAT(' MIN, MAX, AND INC of the aspects ....',3I4)
This subroutine inputs required frequencies.

SUBROUTINE FRE(FMIN, FINC, NF, LINE, IQ, IGRAPH, *)

CHARACTER TLINE*8
CHARACTER*(*) LINE(11)
REAL THINF(10), TINCF(10)
INTEGER TINF(10)

Specify frequencies

IF(IQ.EQ.1.AND.IGRAPH.GT.1) THEN
   TYPE = 1
   TYPE *= 'Input FMIN, FINC and no. of frequencies'
   ACCEPT *, FMIN, FINC, NF
   TINF(IQ)=FMIN
   TINCF(IQ)=FINC
   TINF(IQ)=NF
   ELSE
      FMIN=TINF(IQ)
      FINC=TINCF(IQ)
      NF=TINF(IQ)
   END IF

IF(IQ.EQ.1.AND.IGRAPH.GT.1) THEN
   DO I=1,70
      LINE(?)(I:I)= ' ' 
   END DO
   END IF

IF(IQ.EQ.1) THEN
   ENCODE(20, 1080, LINE(7))NF  
   ELSE
   ENCODE(3, 1085, TLINE)NF     
   IF(IQ.EQ.1) THEN
   ENCODE(3, 1085, TLINE)NF
   ELSE
   ENCODE(3, 1085, TLINE)NF
   DO I=70,20,-1
      IF(LINE(7)(I:I),' ' ) THEN
         IEND L=I
      BOTO 83
      END IF
      END DO
      L7=IEND L
      LINE(7)(L7+6:L7+8)=TLINE(1:8)
   END IF

FMAX=FMIN+(NF-1)*FINC
PRINT 2889, FMIN, FMAX, FINC, NF
2889 FORMAT (' FMIN, MAX, INC & No. of the freqs ...... ',F7.3, I4)

RETURN 1
This subroutine inputs required a priori information for aspect angle

```
SUBROUTINE AAI(IAK,LINL,IGRAPH,*)
CHARACTER(*1) LINE(11)
CHARACTER TAK(10)*1,AK*1

IF(IG.EQ.1.OR.IGRAPH.EQ.1) THEN
  TYPE 0,' Aspect known  --> TYPE K'
  TYPE 0,' Aspect unknown --> TYPE U'
  ACCEPT 19,AK
END IF

IF(AK.EQ.'K') THEN
  IAK=1
  PRINT*, ' Aspect assumed known'
ELSE IF(AK.EQ.'U') THEN
  IAK=0
  PRINT*, ' Aspect assumed unknown'
ELSE
  GOTO 1
END IF

IF(IAG.EQ.1.AND.IAK.EQ.1) THEN
  LINE(5)(1:30)=ASPECT ASSUMED KNOWN
ELSE IF(IAG.EQ.0.AND.IAK.EQ.1) THEN
  LINE(5)(1:30)=ASPECT ASSUMED UNKNOWN
END IF
IF(IAG.GT.1.AND.IAK.EQ.0) LINE(5)(21:41)=/' UNKNOWN'
IF(IAG.GT.1.AND.IAK.EQ.1) LINE(5)(21:41)=/' KNOWN'
RETURN 1
```

This subroutine selects relative amplitude feature.

```
SUBROUTINE RAL(IRA,IGRAPH,*)
CHARACTER TRA(10)*1,RA*1

IF(IG.EQ.1.OR.IGRAPH.EQ.1) THEN
  TYPE 0,' Use relative amplitude ? Y or N'
  ACCEPT 17,RA
END IF
```
SUBROUTINE NS(IS3, IS4, NX, M, IGRP, IGRA, IGRAPH)

INTEGER TNEX(10)

C This subroutine inputs seeds for random number generator, and number of experiments.

C C

C Input two arbitrary large odd integers which will serve as seeds for generating random numbers

IF(IGRA.EQ.1) THEN
  TYPE 'Input two arbitrary large odd integers'
  ACCEPT *IS3, IS4
  TYPE 'Input number of experiments'
  ACCEPT *NX
  TNEX(IQ)=NX
ELSE
  NEX=TNEX(IQ)
END IF

PRINT 30, IS3, IS4
PRINT 20, NEX

C C

C This subroutine selects error in aspect angle

SUBROUTINE ERA(IER, AS, LINE, IQ, INCASP, IGRAPH)

C C

C

C

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Note there must be 2 or 3 aspects to use the error subroutine.

IF (IQ.EQ.1.OR.IGRAPH.EQ.1) THEN
  TYPE *, 'Type Y or N for error in aspect'
  ACCEPT 10, ER_AS
  IF (IQ.EQ.0) THEN
    TYPE *, 'Error subroutine for error in aspect'
  ELSE
    ER_AS(1:1)=TER_A(IQ)(1:1)
  END IF
END IF

IF (ER_AS.EQ.'Y') THEN
  IER_AS=1
  PRINT *, 'Error introduced in aspect', INCASP, 'deg'
ELSE IF (ER_AS.EQ.'N') THEN
  IER_AS=0
  PRINT *, 'No error in aspect'
ELSE
  GOTO 1
END IF

IF (LINE(11)(1:3).NE.1 .AND. IER_AS.EQ.1) THEN
  ENCODE(20,20,LINE(11))
ENDIF

IF (LINE(11)(1:3).NE.' ' .AND. IER_AS.EQ.1) THEN
  ENCODE(20,20,LINE(11))
ENDIF

DO I=70,18,-1
  IF (LINE(11)(I:1).NE. ' ') THEN
    L11=I
    GOTO 40
  END IF
END DO

LINE(11)(L11+5:L11+8)=LINE(11)(1:3)
END IF

RETURN 1

END

This subroutine selects nearest neighbour features.

SUBROUTINE FEA(IY, IYW, RR, IQ, IGRAPH, ISO, LINE, *)

CHARACTER*(*) LINE(11)
CHARACTER YWA(3)*1, TWA(10)*2, T_CHAR*4, TLINE*5
REAL TRR(10)

IF (IQ.EQ.1.OR.IGRAPH.EQ.1) THEN

367
TYPE *,'Input required features A,W'
ACCEPT 557,YWA
557 FORMAT(3A1)
DO I=1,3
TWA(IG)(I:I)=YWA(I)
END DO
ELSE
DO I=1,3
YWA(I)=TWA(IG)(I:I)
END DO
RR=TRR(IQ)
END IF

C
Assign feature control variables

C
IYW=0
IYA=0
DO I=1,3
IF(YWA(I).EQ.'A') IYA=1  
IF(YWA(I).EQ.'W') IYW=1
END DO
IF((IYA+IYW).NE.0) RETURN
C
RETURN 1
END

C
This subroutine forms the input routine for an algorithm designed
to reduce the effects of impulsive noise. This algorithm is still
in an experimental stage and needs further development.

C
SUBROUTINE NAS(JNAG,ANAG,*)
C
CHARACTER YNAG*1
C
1 TYPE *,'Use Neil's Algorithm (Y or N)' 
ACCEPT 10,YNAG
10 FORMAT(A1)
IF(YNAG.EQ.'Y') THEN
JNAG=1
TYPE *,'Input value for standard deviation test (2)'

368
ACCEPT *,,ANAG
PRINT *, 'Nella Algorithm used'
PRINT *, ANAG
ELSE IF (YNAG.EQ. 'N') THEN
  JNAG=0
ELSE
  GOTO 1
END IF
C
RETURN 1
END
C
C

END
C
C THIS SUBROUTINE READS A FILE
C
SUBROUTINE TRT(INFILE,AN,RI,FKIN,FINC,SF)
    INTEGER*2 LINE2(24)
    INTEGER*4 ID(12)
    CHARACTER INFILE*(12)
    REAL*4 AM(B31),PH(B31)
    INCLUDE 'SYSLIBRARY:FORIOSDEF'
    0106 OPEN(UNIT=5, NAME=INFILE, TYPE='OLD', READONLY, IOSTAT=IERR, ERR=9100)
    30 FORMAT(12A,24A,4A)
    READ(5,30)(ID(I),I=1,12),LINE2,NP,FKIN,FINC,SF
    40 FORMAT(128A)
    GO TO 331
    9100 IF(IERR.EQ.9100) THEN
        TYPE 1112, INFILE
    1112 FORMAT(3A,'FILE ',A14,' WAS NOT FOUND',/,,1A9)
        ENTER FILENAME AGAIN'
        ELSE IF(IERR.EQ.9100) THEN
            TYPE 1113, INFILE
    1113 FORMAT(2A,'FILE ',A14,' WAS BAD, ENTER NEW FILENAME')
        ELSE
            TYPE 3,'UNRECOVERABLE ERROR, CODE=',IERR
            STOP
    ENDIF
        TYPE 3,'INPUTFILENAME'
        ACCEPT 1114, INFILE
    1114 FORMAT(A)
        GO TO 0106
    331 CLOSE(UNIT=5,DISP='SAVE')
RETURN
END

C
C THIS SUBROUTINE DECODES PARAMETERS READ
C FROM THE DATA FILES, WHICH ARE CODED IN ASCII CODES
C
SUBROUTINE DCDE(NP,FMIN,FINC)
    COMMON BUFF
    BYTE BUFF(5656)
    INTEGER*4 IMINI,INCI,NP
C
C NO OF DATA POINTS IS STORED IN THREE CHARACTERS, AND
C STARTING FREQ, AND FREQ INC. IN 5 CHARACTERS
C
CHARACTER*3 CHL
CHARACTER*5 CFF,CINC
EQUIVALENCE (BUFF(124),CHL),(BUFF(131),CFF),(BUFF(140),CINC)
C
C CONVERT CHARACTERS INTO THEIR NUMERICAL EQUIVALENTS
C
DECODE(3,100,CHL)NP
    100 FORMAT(I5)
DECODE(5,100,CFF)IMINI

FMIN=FLOAT(IMINI)/100.
FINC=FLOAT(INCI)/1000.
RETURN
END
This procedure sorts through a directory and compiles a file of desired filenames. This file later becomes the source of a control routine in CATLOG to enable large batches of files to be scaled automatically.

```
ON ERROR THEN GOTO DONE
ON CONTROL_Y THEN GOTO DONE

INQUIRE F_TYPE "Input file type "%; ""
INQUIRE D_OUT "Input output directory filename "

DIRECTORY/VERSIONS=1/COLUMNS=1-
/NODATE/NOFILES/NOHEADINGS-
/NOTAILING/OUTPUT='D_OUT' 'F_TYPE'

WRITE SYSOUTPUT D_OUT
WRITE SYSOUTPUT "="

DIRECTORY/VERSIONS=1/COLUMNS=1/NOFILES/NOHEADINGS 'F_TYPE'

DEFINE USER_MODE SYSINPUT SYSCOMMAND
RUN FC:NAMEFILE

IF .NOT. @STATUS THEN GOTO DIR_ERR
DONE:
EXIT
DIR_ERR:
WRITE SYSOUTPUT "Error: 'F#MESSAGE(@STATUS)'
EXIT
```

371
| Command procedure : LINKCAL61.COM  |
| Linking routine for CAL61.FOR       |
| LINK CAL61, SUBCAL61, FORT, SUBREMOVE, RCLIB/LIBRARY, -  |
| PLOTV, PNV, 'PLOTLIB'               |

| Command procedure : LINKTAF.COM      |
| Linking routine for TAFCAL.FOR       |
| LINK TAFCAL, TRT, PF: FORT, 'PLOTLIB' |

| Command procedure : LINKCAT.COM       |
| Linking routine for CATLOG.FOR        |
| LINK CATLOG, RDFL                     |