METEOROLOGICAL DATA ANALYSES IN
SUPPORT OF THE ABRES PROGRAM

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Concord, Massachusetts 01742

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The Air Force Geophysics Laboratory (AFGL) has been tasked by the Space and Missiles System Organization (SAMSO) to provide meteorological analysis in support of the Advanced Ballistic Reentry Systems (ABRES) Program. The primary objectives of the work carried out under this contract have been to assist AFGL in (1) the analysis of surface, upper air, and satellite data leading to the computation of liquid water content profiles, (2) the analysis of special observational data obtained from aircraft flight experiments, (3) computer analysis of radar data, (4) development of display techniques for the AFGL Liquid Water.
Content Analyzer, (5) the evaluation of the uncertainties in the estimation of hydrometeor mass concentrations using radar and aircraft measurements, and (6) the assessment of thin cirrus cloud over the tropical Pacific.

The major results accomplished include (1) the development of improved computer programs for the real-time estimation of liquid water content profiles using radar measurements, (2) the completion of meteorological cross section analyses for Hong Kong, (3) the development of improved techniques for the collection, processing, and display of satellite meteorological data, (4) the establishment of the important error contributions to the uncertainties in the liquid water content estimates; the major contributors to the errors were found to be the radar calibration, polarization effects by the particles, length-to-mass conversion from the particle measurements, distribution estimates, and extrapolation limitations, (5) the establishment of procedures to derive liquid water content profiles from surface, upper air, and satellite data, (6) the development of improved techniques for the derivation of the environmental severity index using visible and infrared brightness values from satellites, and (7) the definition of the concentrations of ice particles which can be detected visually from the surface or from aircraft and by means of an infrared sensor from space.
ACKNOWLEDGMENTS

The research and technical support under this program has been carried out by several key individuals. Mr. Lloyd Perry and Mr. Peter Piechota provided the support for the AFGL Liquid Water Content Analyses (Task 1); Mr. Barry Mareiro worked with the McIDAS (Task 2); Ms. Shu-Lin Tung was responsible for the extensive analysis of aircraft and radar data and for the development of improved techniques for handling the data (Task 3); Messrs. Allan Bussey, Tony Lisa, Pieter Feteris and Ms. Mary Grace Fowler were responsible for the development of new techniques for the determination of liquid water content profiles using standard and special meteorological observations (Task 4); Dr. Robert Crane and Dr. Paul Smith completed the research on the difficult problems of evaluating the uncertainties in the estimation of hydrometeor mass concentration using SPANDAR and aircraft measurements (Task 5); and Dr. Hsiao-hua Burke carried out the theoretical study of the detectability of high altitude thin cirrus clouds by various remote sensors (Task 6). Coordination of work on these tasks was the responsibility of Ms. Mary Grace Fowler. The dedicated effort of all of these scientists is gratefully acknowledged and appreciated.

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1. INTRODUCTION

The Air Force Geophysics Laboratory (AFGL) has been tasked by the Space and Missiles System Organization (SAMSO) to provide meteorological analysis in support of the Advanced Ballistic Reentry Systems (ABRES) Program. The primary objectives of the work carried out under this contract by Environmental Research & Technology, Inc. (ERT) have been to assist AFGL in (1) the analysis of surface, upper air, and satellite data leading to the computation of liquid water content profiles, (2) the analyses of special observational data obtained from aircraft flight experiments, (3) computer analysis of radar data, (4) development of display techniques for the AFGL Liquid Water Content Analyzer, (5) the evaluation of the uncertainties in the estimation of hydrometeor mass concentrations using radar and aircraft measurements, and (6) the assessment of thin cirrus cloud over the tropical Pacific.

Throughout the three-year period of the contract, there were six major tasks which were identified as contributing to the support of the ABRES program. These tasks varied widely in their complexity and in their level of effort. This report is intended to provide a summary of each of these tasks.

As a result of the research and analyses on the program, three scientific reports have been published (Crane, 1978; Burke et al., 1978; and Bussey, 1979). These reports provide the details of the most significant scientific contributions accomplished under the contract.
2. ANALYSIS AND SUPPORT

2.1 Task 1 - Support for the AFGL Liquid Water Content Analyzer

Through a contract with Raytheon Company, AFGL developed an instrument to derive real-time estimates of the liquid water content using weather radar data. Support for the maintenance, operation, and for the development of new or improved computer programs for the water content analyzer was provided by ERT for two winter seasons at Wallops Island, Virginia. The analyzer was a key instrument in the assessment of the hydrometeor environment for the Sandia Air Force Materials Study (SAMS) Program.

Documentation of the work completed under this task was given in two working reports. One report was entitled WETHRAD (Weather Radar System) and documented some of the operating system of the analyzer as well as the new programs developed during the 1976 winter season at Wallops Island. The second working report detailed all the changes made in the system during the 1977 winter season and provided an accurate description of all aspects of operating and using the analyzer. Work on the task was terminated at the end of FY 77 when there was no further requirement for use of the analyzer at Wallops Island.

2.2 Task 2 - Support for the McIDAS

This task involved support for the maintenance and modifications to the Man Computer Interactive Data Access System (McIDAS) at AFGL. The primary work accomplished was (1) the completion of the antenna installation including the tests for all drive systems and for printing accuracy, (2) the installation and testing of a sensitive receiver, (3) development of an improved antenna tracking circuit, and (4) modifications to the automated handling of satellite meteorological data in order to acquire information from the mid-west United States. Support on this program was provided for a six-month period and was terminated in July 1976.
2.3 Task 3 - Cloud Data Analysis

The major portion of the work in this task was the analysis of aircraft measurements of cloud and precipitation particles and the incorporation of these measurements with corresponding observations of the radar structure of the storm as observed with the radar systems at Wallops Island. Extensive studies were completed on relating the ice crystal area to ice crystal mass which was then used to determine the particle distributions and the cloud liquid water content of the ice and snow crystals. The final matching of the water contents with those derived from the radar measurements was accomplished through a series of analysis techniques which had been developed throughout the first year of the study.

During the third year of the program, the work concentrated on the mass versus size and mass versus area relationship from Knollenberg's ice crystal data to find the logarithmic regression equations and the standard errors to fit Knollenberg's linear functions. Work proceeded on the derivation of weighting functions of the average projection ratio channels and the equivalent circle ratio channels for eleven different ideal crystals.

Particle size data from several aircraft altitudes were processed and used to compute precipitation growth, and a computer program was set up to create the appropriate input for computations of ice crystal growth. Cloud data from twenty-one experimental tests carried out at Wallops Island and Kwajalein were processed.

A significant part of the effort on this task involved the development of computer programs to estimate the probable liquid water contents of clouds and precipitation using radiosonde data along with visible and infrared (IR) data from satellites. The liquid water content was then multiplied by the height and integration of this product was computed to obtain a parameter called the Environmental Severity Index (ESI). Using a data base of eleven radiosonde stations, the mean, median, and accumulated frequency tables of ESI were obtained.

Further development on the ESI studies involved taking an existing computer program (ANYPT), which extracted the mean atmospheric profiles of temperature, pressure, humidity and density over any point on the
globe, and modifying it to extract the mean temperature profiles over the NMC grid points covering the Eurasia area. This new program, entitled PROFIL, outputs the profiles generated by ANYPT Program ESRESI was then developed to computer wintertime ESI values; in this program the mean temperature profiles from program ANYPT were used to assist in the discrimination of ice and snow from clouds. Additional programs, called PROBMP and TSTATS, have been used operationally to derive statistics of ESI values over Eurasia for the months of May through September 1973.

During the last six months of the project, work involved the processing and analysis of aircraft and radar data from Kwajalein. Major accomplishments were as follows: (1) an organization of both crystal size relationships and crystal area relationships for various particle types was completed, (2) Knollenberg's 2-D crystal data were evaluated manually and a format was developed which made it possible to compare the results with a computerized version of the 2-D data analysis, (3) the processing of both the AFGL-2 and the satellite data comparison models for the three months of May, July and October 1973 was completed, and (4) a computer program to generate a non-dimensional particle-size distribution, which was subsequently used to obtain the liquid water content, was developed.

2.4 Task 4 - Cross Section Analysis and Liquid Water Content Assessment at Kwajalein

Bussey (1978) has prepared a scientific report which describes the methodology for cross section analysis and liquid water content assessment. Since the details of this task are included in the report, only the abstract is given below.

"Analyzing surface and upper air meteorological observations in the form of a time cross section is a concise form of simulating atmospheric conditions over a particular location. Using these observations and DMSP satellite data, procedures for estimating the liquid water content within regions of cloud and precipitation have been derived. The procedures for deriving the liquid water content profiles have evolved over a
five-year period, and it is the purpose of this report to describe the current methodology. The profiles of liquid water content are used as input for a determination of the environmental severity index (ESI). The advantage of the technique described is that estimates or a climatology of ESI can be obtained in any part of the world through an analysis of readily available meteorological observations."

2.5 Task 5 - Evaluation of Uncertainties in the Estimation of Hydrometeor Mass Concentrations Using SPANDAR Data and Aircraft Measurements

A major effort under the project was the detailed assessment of uncertainties or errors in the radar and aircraft estimates of cloud and precipitation liquid water contents along the trajectories of reentry test vehicles. Crane (1978) has described the complete results of the work completed under this task. Since this was a major and important part of the overall project, a more detailed description of the task is provided than the descriptions for the previous tasks.

2.5.1 Objectives of the Error Analysis

The Air Force Geophysics Laboratory (AFGL) was tasked by the Space and Missiles Systems Organization (SAMSO) to forecast, sample, and report the ice, snow, and water particle environment for erosion testing of payloads launched from the NASA Wallops Flight Center. The mass concentrations of hydrometeors along the payload trajectory were estimated using radar data obtained from the NASA Space Range Radar (SPANDAR) at Wallops Island, VA. The radar reflectivity-to-mass concentration conversion relationships required to interpret the radar data were obtained from aircraft measurements made in the vicinity of the payload trajectory prior to launch.

The objective of the uncertainty (or error) analysis reported herein is to review the radar and aircraft measurement procedures and provide estimates of both the possible (maximum) and probable (most likely) error bounds (uncertainties) for the mass concentration estimates. This report does not provide error bound estimates for a particular launch but considers the kinds of uncertainties that could occur and provides estimates of their expected values. This analysis only considers uncertainties
that can occur when the equipment is operated correctly and does not malfunction. In some cases, it is possible to detect operator or equipment error but often it is not. Possible areas for improvement are pointed out in the report but conscientiousness was expected of those involved in the measurement program and is not questioned further.

2.5.2 Brief Description of the Measurement Program

Mass concentrations of ice, snow, and water particles encountered by the test vehicle as it passed along its trajectory are required to evaluate the susceptibility of nose tip and side wall materials to erosion. In situ observations of mass concentration are not possible during vehicle traversal. Remote observations by radar were obtained immediately prior to, during, and after vehicle passage to best describe the hydrometeor environment. These observations were converted to mass concentration estimates using reflectivity-to-mass relationships derived from aircraft samples of the hydrometeors prior to launch.

Measurements with SPANDAR were made to support the Sandia Air Force Materials Study (SAMS) program during the 1975-1976 time period (SAI, 1976) and to support the Materials Screening Vehicle (MSV) program during the 1976-1977 time period (Cockayne and Fletcher, 1976). For this report, the differences between the two measurement programs are best described by noting that the SAMS observations were made close to the radar (5-25 km range) while the MSV observations are made at significantly longer ranges (~150 km).

Four separate types of radar observations were required for the erosion measurement program: elevation scans over disdrometers and rain gauges to provide primary radar calibration during a mission, sphere tracks to provide secondary radar calibration, link-mode observations with aircraft flights made just before launch to provide reflectivity-to-mass conversion relationships, and volume scan observations along the expected vehicle trajectory immediately preceding, during, and after vehicle traversal through the erosion test region. These measurements were supported by three digital processor systems attached to the radar each of which must be separately calibrated. The radar operator (the John Hopkins Applied Physics Laboratory-APL) maintained and operated a minicomputer processing system that controlled the antenna and was used
to record the link-mode observations. AFGL maintained and operated the
two other processors, one to provide primary data recording for post
test analysis and the second to provide on site, near-real-time liquid
water content estimates for mission control. The second processor also
provided backup data recording for post test analysis.

The aircraft measurement program entailed the use of two aircraft
operated by AFGL to obtain hydrometeor size, phase, and type distributions.
Size distributions were obtained using PMS 1-dimensional (1-D) cloud and
precipitation particle probes (Knollenberg, 1970). Crystal habit or
particle phase and type information were obtained using PMS 2-D cloud and
precipitation particle probes. The aircraft also recorded temperature,
humidity, air speed and time data to aid in post test analysis.

Link-mode observations encompassed the simultaneous use of one of the
aircraft and the radar to provide particle distribution and reflectivity
measurements. SPANDAR automatically tracked the aircraft using its conical
scan tracking system and sampled the reflectivity values one range
interval closer to the radar than the interval occupied by the aircraft.
In post analysis the aircraft and radar data were statistically compared.
Although the data were taken simultaneously, the data sets were obtained
from significantly different sampling volumes. The sampling volume for
the aircraft probe is determined by the instrument capture area times
the distance traversed by the aircraft as it obtained a sample. The
radar sampling volume is determined by the antenna beam cross section
generated as it rotated about the aircraft position while scanning to
track the aircraft times the distance along the beam defined by the pulse
length of the radar and subsequent radar integration.

The final link in the chain of observations and analyses required to
produce mass concentration estimates from SPANDAR data was the analysis
of the PMS 1-D probe observations to provide equivalent melted sphere
size distributions from the observed particle shadow lengths. Given the
equivalent melted sphere sizes and postulates about particle shape and
orientation, the mass and reflectivity values may be calculated for use
in generating the reflectivity-to-mass conversion relationships. The
PMS probe does not directly sample the equivalent melted sphere diameter;
it only records an estimate of the maximum size (length) of the projec-
tion of each particle onto a horizontal line perpendicular to the direc-
tion of aircraft motion as the particle passes through the sampling head. The probe does not provide any information on the other physical dimensions required for the calculation of a volume estimate or on the particle density required for the calculation of a mass estimate. Except for the PMS 2-D probe, the 1-D probe provides more relevant information about hydrometeors than any other instrument currently available.

The length-to-melted diameter (mass) conversion relationships for the PMS probes have been statistically determined using a two step process. First, the response of the particle measurement system to idealized particle shapes was simulated to provide corrections for habit, orientation, and size (Knollenberg, 1975). Second, a large number of observations of naturally occurring crystals were analyzed to determine the expected relationship between the maximum (horizontal) crystal dimension and particle mass. Separate relationships were developed for each crystal habit. To reduce the sensitivity of the reflectivity-to-mass conversion relationships to possible uncertainties associated with the identification of crystal habit and with the use of a limited number of observations to generate the required length-to-mass relationships, the PMS data were used only to calculate a multiplier to be used with the square root of the radar reflectivity values for the estimation of mass concentration (the \( k \) factor) and not to provide direct estimates of either mass or reflectivity (\( k \) is equal to the ratio of mass to the square root of reflectivity). The resultant reduction in sensitivity to sizing error however, depends upon the breadth of the particle size distribution.

Reduced to bare essentials, the radar was used to provide an estimate of the radar reflectivity factor, \( Z \), and the aircraft measurements were used to provide the relationship between \( Z \) and the mass concentration as expressed by an equivalent liquid water content, \( M \). The liquid water content was determined by

\[
M = k \sqrt{Z}
\] (1-1)

Uncertainties in the estimation of liquid water content arise from two basic causes: equipment calibration and data interpretation. The total uncertainty for an \( M \) estimate is caused by both aircraft system and radar system uncertainties. In general, the largest contribution to the
2.5.3 Total Uncertainty Budget

The possible uncertainty sources and their contribution to the total liquid water content estimation uncertainty budget are listed in Table 1. The uncertainties are separated into radar and aircraft categories and further segregated into calibration and interpretation classes. The uncertainties for each category are listed by contribution to $M$ using $M = \sqrt{Z}$. These categories are then subdivided into identifiable subclasses which are described in detail in this report. Uncertainty estimates are provided for SAMS and MSV missions. The difference between the two are due to differences in distance between the radar and the path of the vehicle through the erosion test region. In addition, uncertainty estimates are provided for observations at 4 heights. The crystal habits were assumed to be bullet rosettes at a 10 km height, small snow at 6 km, and large snow at 3 km height and wet snow at 2 km height (just above the melting layer).

The measurement uncertainties are assumed to be multiplicative and are given both in decibels and in percent (increase) of the estimated value. Three types of uncertainties are treated, bias (not subject to change over the entire observation period), slowly varying (not expected to change during a mission), and random (expected to change from one sample to the next). Slowly varying uncertainties are caused by such problems as not precisely setting or monitoring power levels in the radar set, not estimating the correct calibration constant, or not estimating the correct particle habit and size distribution for a given height region. Bias errors are of two possible sources: uncertainties in the values of absolute constants such as antenna gain that may exist even after a number of measurements to determine their values, and the use of incorrect values for constants in the calibration or data analysis procedures. Bias errors caused by the use of incorrect constants or procedures may be
### TABLE 1b

**MSV Missions (150 km Range): Contributions to Liquid Water Content Uncertainties**

<table>
<thead>
<tr>
<th>CATEGORY</th>
<th>2 km Height - Large Snow</th>
<th>3 km Height - Large Snow</th>
<th>6 km Height - Small Snow</th>
<th>10 km Height - Ice Crystals</th>
</tr>
</thead>
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<tr>
<td>Radar - $\sqrt{2}$</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Calibration</td>
<td>0.2 S 0.5* 12</td>
<td>0.2 S 0.5* 12</td>
<td>0.2 S 0.5* 12</td>
<td>0.2 S 0.5* 12</td>
</tr>
<tr>
<td>Interpretation</td>
<td>- 0.3* 7</td>
<td>- 0.2* 5</td>
<td>- 0.1* 2</td>
<td>- 0.1* 2</td>
</tr>
<tr>
<td>Attenuation</td>
<td>- - 5.0 216</td>
<td>- - 2.0 58</td>
<td>- - 0.1 2</td>
<td>- - 0.1 2</td>
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<tr>
<td>Polarization</td>
<td>- - 1.4 38</td>
<td>- - 0.8 20</td>
<td>- - 0.6 15</td>
<td>- - 0.5 12</td>
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<tr>
<td>Range Errors</td>
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<td>- - 0.1 2</td>
<td>- - 0.1 2</td>
<td>- - 0.1 2</td>
</tr>
<tr>
<td>Aircraft - $\kappa$</td>
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<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Calibration</td>
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<td>- - 0.1 2</td>
<td>- - 0.1 2</td>
<td>- - 0.1 2</td>
</tr>
<tr>
<td>Interpretation</td>
<td>- - 0.2 5</td>
<td>- - 0.2 5</td>
<td>- - 0.3 7</td>
<td>- - 0.8 20</td>
</tr>
<tr>
<td>Length-to-Mass</td>
<td>- - 1.2 32</td>
<td>- - 1.2 32</td>
<td>- - 1.4 38</td>
<td>- - 3.6 140</td>
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<tr>
<td>Crystal Habit</td>
<td>- - 2.2 66</td>
<td>- - 0.1 2</td>
<td>- - 0.1 2</td>
<td>- - 0.1 2</td>
</tr>
<tr>
<td>Truncation</td>
<td>- - 2.9 95</td>
<td>- - 0.3 7</td>
<td>- - 0.2 5</td>
<td>- - 0.1 2</td>
</tr>
<tr>
<td>Sampling</td>
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<td>- - 1.0 26</td>
<td>- - 1.0 26</td>
<td>- - 1.0 26</td>
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<tr>
<td>$\kappa / \tau$</td>
<td>- - 6.6 357</td>
<td>- - 2.7 86</td>
<td>- - 2.0 58</td>
<td>- - 3.9 145</td>
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<tr>
<td><strong>TOTAL</strong></td>
<td>0.2 S 6.6 357</td>
<td>0.2 S 2.7 86</td>
<td>0.2 S 2.0 58</td>
<td>0.2 S 3.9 145</td>
</tr>
</tbody>
</table>

*Correlated from one height to the next

*Random errors have been removed by averaging
### TABLE 1a

SAMS Missions (25 km Range):
Contributions to Liquid Water Content Uncertainties

<table>
<thead>
<tr>
<th>CATEGORY</th>
<th>2 km Height - Large Snow</th>
<th>3 km Height - Large Snow</th>
<th>6 km Height - Small Snow</th>
<th>10 km Height - Ice Crystals</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>E. Bias</td>
<td>Slowly Vary</td>
<td>Random</td>
<td>E. Bias</td>
</tr>
<tr>
<td>Radar - $\sqrt{E}$</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Calibration</td>
<td>0.2 5</td>
<td>0.5* 12</td>
<td>0.2 5</td>
<td>0.2 5</td>
</tr>
<tr>
<td>Interpretation</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>Attenuation</td>
<td>0.3* 7</td>
<td>0.05* 1</td>
<td>-</td>
<td>0.3* 7</td>
</tr>
<tr>
<td>Polarization</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>Side Lobes</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
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<tr>
<td>Refraction</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>Range Errors</td>
<td>0.3** 7</td>
<td>0.1** 2</td>
<td>-</td>
<td>0.4** 10</td>
</tr>
<tr>
<td>Aircraft - $\kappa$</td>
<td>-</td>
<td>0.1 2</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>Calibration</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
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<tr>
<td>Interpretation</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>Length-to-Mass</td>
<td>-</td>
<td>0.2 5</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>Crystal Habit</td>
<td>-</td>
<td>1.2 32</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>Truncation</td>
<td>-</td>
<td>2.2 66</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>Sampling</td>
<td>-</td>
<td>2.9 95</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>$k/\tau$ Extrapolation</td>
<td>-</td>
<td>1.0 26</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>Total - $M = k\sqrt{\tau}$</td>
<td>0.8 20</td>
<td>4.0 151</td>
<td>-</td>
<td>0.9 23</td>
</tr>
</tbody>
</table>

*Uncertainty may be removed by reprocessing the data
*Correlated from one height to the next
*Random errors have been removed by averaging
eliminated but bias errors caused by uncertainties in absolute constants cannot be. Uncertainties that can be eliminated by reprocessing the available data are identified by asterisks (*) in the dB columns of Table 1.

Slowly varying uncertainties are statistical in nature and are assumed to have a log-normal distribution (log-normal since the uncertainties are multiplicative). The uncertainties for any category are considered to have a log-normal distribution which is described by the standard deviation of the logarithm (decibel value). The composite uncertainty is the sum of a number of independent uncertainties, hence, is calculated using the sum of the variances for each category (expressed in decibels).

The bias uncertainties are not random and must be treated in a different manner. Following Crane (1971), the maximum possible uncertainty for each category is summed to provide an estimate of the total maximum possible uncertainty. Again the logarithm of the uncertainty is used (decibel value) since the uncertainties are multiplicative. The maximum possible uncertainty is interpreted as being the equivalent of the 3-σ or three standard deviation value. To express the bias uncertainty as an expected uncertainty, one third the maximum possible value is used for entry in Table 1. The one third value is labeled E Bias for expected bias value to indicate that it does not correspond to the full maximum possible bias error value. The total error bounds for the measurements are obtained by combining the expected uncertainties using the sums of the squares of the statistically estimated components. The resultant error bounds are summarized in Table 2.

2.5.4 Uncertainties in the Estimate of the Environmental Severity Index

The analysis of uncertainties presented in this report primarily address the problem of estimating mass concentration values. Rain erosion analysts use the mass concentration estimates to model integrated effects. Their interest is not in the error at a point but in the error of integrals such as the Environmental Severity Index (ESI). ESI is defined as

\[
ESI = \int_{0}^{\infty} M(h) \, dh
\]  

(1-2)
### TABLE 2

**MOST PROBABLE UNCERTAINTY BOUNDS FOR LIQUID WATER CONTENT ESTIMATES**

<table>
<thead>
<tr>
<th>Mission</th>
<th>Height (km)</th>
<th>68 Percent Errors</th>
<th>Bounds percent</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td>dB</td>
<td>factor</td>
</tr>
<tr>
<td>SAMS</td>
<td>3.0</td>
<td>±1.9</td>
<td>(1.55)±1</td>
</tr>
<tr>
<td></td>
<td>6.0</td>
<td>±2.2</td>
<td>(1.66)±1</td>
</tr>
<tr>
<td></td>
<td>10.0</td>
<td>±2.5</td>
<td>(1.78)±1</td>
</tr>
<tr>
<td>MSV</td>
<td>3.0</td>
<td>±2.7</td>
<td>(1.86)±1</td>
</tr>
<tr>
<td></td>
<td>6.0</td>
<td>±2.0</td>
<td>(1.58)±1</td>
</tr>
<tr>
<td></td>
<td>10.0</td>
<td>±2.3</td>
<td>(1.70)±1</td>
</tr>
</tbody>
</table>
where $M(h)$ is the particle mass concentration at height $h$ (along the trajectory). The uncertainty (error) in ESI, $\Delta E$, is linearly related to the error in mass concentration $\Delta M$ by

$$ESI + \Delta E = \int_0^\infty (M(h) + \Delta M(h))dh$$

or

$$\Delta E = \int_0^\infty \Delta M(h)dh$$

The errors in mass concentration are multiplicative and best expressed by the ratio $E_m = \Delta M/M$ (in decibels or percent). Therefore,

$$\Delta E = \int_0^\infty \left( \frac{\Delta M}{M} \right)Mdh = \int_0^\infty E_m Mdh$$

(1-4)

The bias component of the error is constant over space ($h$) and does not vary in time or from mission to mission. The bias error in ESI, $\Delta E_B$, therefore, is given by

$$\Delta E_B = (E_m)_B \int_0^\infty Mdh = (E_m)_B \cdot ESI$$

(1-5)

or

$$(E_E)_B = (E_m)_B$$

The ESI or any other linear integral function of $M$ therefore has a bias error equal in percent (or decibels) to the bias error in $M$.

The random error is due primarily to sampling errors associated with the radar measurements. This component of error is assumed to be uncorrelated from sample to sample and can be reduced by averaging. The integral for ESI can be reduced to a summation over samples

$$ESI = \int_0^\infty Mdh = \sum_{j=1}^J \left( \sum_{i=1}^{I_j} \frac{M_{ij}}{I_j} h_j \Delta h \right) = \sum_{j=1}^J \bar{M}_j h_j \Delta h$$

(1-6)

where $I_{ij}$ samples are averaged for the $j^{th}$ layer and $J$ layers are summed to provide an ESI estimate. The error is again given by
\[
\Delta E = J \sum_{j=1}^{J} \sum_{i=1}^{I_j} (E_m)_{ij} \bar{M}_j h_j \Delta h
\]

Since the errors are assumed to be statistically independent from sample to sample, the expected square error is given by

\[
(\Delta E_R)^2 = J \sum_{j=1}^{J} \sum_{i=1}^{I_j} (m_{Rj})^2 (\bar{M}_j h_j \Delta h)^2 = J \sum_{j=1}^{J} \frac{[(E_m)_{Rj}]^2}{I_j} (\bar{M}_j h_j \Delta h)^2
\]

and the rms error, \( \Delta E_R \) is given by

\[
\Delta E_R = \sqrt{\sum_{j=1}^{J} \frac{[(E_m)_{Rj}]^2}{I_j} \bar{M}_j^2 h_j^2 \Delta h}
\]

(1-7)

If \( \bar{M}_j \) varies inversely with height, \( \bar{M}_j h_j = \text{constant} = M_0 h_0 \) and if \( (E_m)_R \) and \( I_j = I \) are not dependent on \( h \), the rms error would be given by

\[
\Delta E_R = \frac{(E_m)_R M_0 h_0 h}{\sqrt{J I}}
\]

(1-8)

where \( JI \) is the number of samples. In general, \( \bar{M}_j \) does not vary inversely with height and the reduction in sampling error is not as great as expressed by (1-8).

Intermediate between the random errors and bias errors are the slowly varying errors. These errors are correlated at each height level but may be uncorrelated from one height to the next. Errors which are caused by uncertainties in setting radar parameters are correlated from one height to another but vary from one mission to another. Errors associated with the incorrect choice of crystal habit are correlated over the depth of a layer with a particular habit but uncorrelated between layers with different habits. The slowly varying error, therefore, is given by

\[
\Delta E_s = \sqrt{\sum_{j=1}^{J} [(E_m)_{sj}]^2 \bar{M}_j^2 h_j^2 \Delta h^2 + \sum_{j=1}^{J} \sum_{k=1}^{J} [(E_m)_{sj} (E_m)_{sk}]_{jk} \bar{M}_j \bar{M}_k h_j h_k \Delta h^2}
\]

(1-9)
where $\rho_{jk}$ is the correlation coefficient between observations at levels $j$ and $k$. This can be approximately reduced to the sum over independent layers, typically 2 to 4, representative of each habit where the correlation coefficient $\rho_{ik}$ is 1 within the layer and zero from layer to layer:

$$\Delta E_s = \sqrt{\sum_{j=1}^{J^*} (\bar{\varepsilon}_m s_j)^2 \bar{\mu}_j h_j^2 \Delta h^2}$$

(1-10)

where $J^*$ is the number of independent layers (layers with different habit).

Referring to Table 1, the bias errors and the slowly varying errors which are correlated from height to height are also errors in ESI. The random errors may be reduced by summing over many samples, hence, do not contribute to the error in estimating ESI. The slowly varying errors which are uncorrelated from layer-to-layer (habit-to-habit) can be reduced only if the mass concentration is inversely related to height otherwise, the effect of the weighting by $\bar{\mu}_j h_j$ must be taken into account when computing the error. Using the sample $M(h)$ distribution given in Figure 1, the errors for ESI computed using the error estimates given in Table 1 are given in Table 3.

**TABLE 3**

<table>
<thead>
<tr>
<th></th>
<th>Bias (dB)</th>
<th>Slowly Varying (dB)</th>
<th>Total (dB)</th>
<th>Percent</th>
</tr>
</thead>
<tbody>
<tr>
<td>SAMS</td>
<td>1.0</td>
<td>1.8</td>
<td>±2.1</td>
<td>-38, +62</td>
</tr>
<tr>
<td>MSV</td>
<td>0.2</td>
<td>2.4</td>
<td>±2.4</td>
<td>-42, +74</td>
</tr>
</tbody>
</table>

*ESI value calculated for measurements at heights above 3 km.
Figure 1  Mass Concentration Profile, M(h), for MSV-2 Test, 20 March 1976
2.5.5 Summary of Results

Error bounds were estimated for typical SAMS or MSV missions. The detailed uncertainty budget is summarized in Table 1. Table 2 presents the total error budget for mass concentration estimates at three different heights and Table 3 presents the ESI uncertainty bounds for measurements at heights above 3 km. Observations at a height of 2 km are subject to large possible uncertainties due to the presence of the melting layer and have been excluded from the ESI uncertainty analysis. In the case of the MSV observations, the inclusion of the 2 km height value in the ESI calculation would increase the slowly varying component of uncertainty from 2.4 to 3.4 dB. A similar but smaller increase would occur for the SAMS missions, from 1.8 to 2.2 dB.

The total uncertainties for mass concentration estimation at heights above 3 km are of the order of 2 dB for the SAMS and MSV missions being slightly higher for the latter. The uncertainties in ESI have a similar magnitude. The ESI uncertainties are not smaller than the mass concentration uncertainties because of the correlations of the slowly varying uncertainties within a region with a predetermined crystal habit and the mass height weighting of each of these regions in the calculation of ESI.

The error budgets given in Table 1 reveal that the major source of uncertainty in the estimation of mass concentration or ESI value arises from the interpretation of the aircraft particle size measurements. Although the use of the $\kappa$ factor in the analysis of aircraft data has significantly reduced the uncertainties associated with distribution measurements, crystal habit uncertainties and $\kappa/Z$ extrapolation from aircraft observations to shot time remain as the dominant cause of uncertainty. For the MSV missions and at a height of 6 km, radar data interpretation contributes 16 percent and the aircraft data interpretation contributes 78 percent of the total uncertainty budget.

Problems still remain in the realm of data interpretation. Uncertainties in the interpretation of the aircraft and radar data exist because of the uncertainties associated with crystal orientation. These problems were mentioned in passing but additional measurements and analyses are required to better define their contributions to the total error budget. The reported tendency for the wide arm precipitation probe
to still undersize particles after the inclusion of recommended corrections also requires additional experimental verification and analysis.

2.6 Task 6 - A Study of Thin Cirrus Over the Tropical Pacific

As part of the overall project, Burke, et al (1978) have carried out a theoretical study of the detectability of high altitude (>10 km) thin cirrus clouds by various remote sensors over the tropical Pacific. Defining the existence and probability of occurrence of thin cirrus clouds is important because of the requirement that there be a natural transition from laminar to turbulent flow over reentering objects, which normally occurs around 10 km. Microscopic roughening of the RV surface caused by impact of isolated hydrometeors can be sufficient to trigger transition and thereby to invalidate the experiment. According to one estimate, there will be an intolerable risk of premature transition above a 10 km height whenever particles larger than 80 microns occur in concentrations of more than 1 m$^{-5}$. A "severe-clear" case or the absence of any detectable ice particles is therefore required above 10 km.

Because of the sparsity of direct measurements, it is difficult to define the atmospheric environment in the vicinity of reentering vehicles for the purpose of evaluating natural atmospheric effects on them. The attempt of identifying cirrus clouds along reentry paths is further complicated by the multiple of their possible crystal sizes, shapes, orientations, numbers and densities. Consequently, a sensitivity study of various remote sensing techniques for quantitative measurements of thin cirrus clouds is required.

The advantage of using remote sensors is that they provide continuous information on cirrus properties in the vicinity of actual reentries. Using in situ measurements in the reentry area, on the other hand, can cause possible dangers from the reentering object. However, remote sensors provide only certain quantities that are often difficult to relate uniquely to standard atmospheric parameters. Assumptions have to be made in order to retrieve all the cirrus cloud quantities: height, thickness, ice water content, crystal size and concentrations, etc. Theoretical estimates, based on optical properties of ice crystals and radiative transfer computations, are carried out to predict the minimum detectable number density of ice crystals for various remote sensors.
The standard tropical atmosphere is used as ambient conditions to simulate Kwajalein atmosphere. The various remote sensors discussed in the Burke, et al. report include (1) eyeball detection at surface, (2) eyeball detection from aircraft altitude, and (3) infrared (6.7 μm wavelength) sensor from space.

The conclusions of the study were as follows:

- from theoretical computations it has been shown that thin cirrus cloud detected from aircraft observations or infrared sensors on-board satellites can be invisible from the ground;
- the required concentration of less than 1 m$^{-3}$ of ice particles greater than 80 μm is too low a threshold (by a factor of $10^3$) for virtually all the existing sensors; and
- a probability study provided a statistical review of the severe clear cases; such severe clear cases are not established from the threshold of 1 m$^{-3}$ of particles greater than 80 μm but from observations of cirrus combined with various meteorological criteria.
3. REFERENCES


