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ENHANCED SYSTEM FOR DETECTION OF RANDOMNESS IN SPARSE TIME SERIES DISTRIBUTIONS

TO WHOM IT MAY CONCERN:

BE IT KNOWN THAT FRANCIS J. O'BRIEN, JR, employee of the United States Government, citizen of the United States of America, and resident of Newport, County of Newport, State of Rhode Island, has invented certain new and useful improvements entitled as set forth above of which the following is a specification:

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DATE OF SIGNATURE

Attorney Docket No. 83995

ENHANCED SYSTEM FOR DETECTION OF RANDOMNESS IN SPARSE TIME SERIES DISTRIBUTIONS

STATEMENT OF GOVERNMENT INTEREST

The invention described herein may be manufactured and used by or for the Government of the United States of America for Governmental purposes without the payment of any royalties thereon or therefore.

CROSS REFERENCE TO RELATED PATENT APPLICATIONS

The present application is related to the following copending application: application of F. J. O'Brien, Jr. entitled "Detection of Randomness in Sparse Data Set of Three Dimensional Time Series Distributions," serial number 10/679,866, filed 6 October 2003 (Navy Case 83996).

BACKGROUND OF THE INVENTION

(1) Field of the Invention

The invention generally relates to signal processing/data processing systems for processing time series distributions containing a small number of data points (e.g., less than about ten (10) to twenty-five (25) data points). More particularly, the invention relates to a two-stage system for classifying the white noise degree (randomness) of a selected signal structure comprising a time series distribution composed of a highly sparse data set. As used herein, the term "random" (or "randomness") is defined in terms of a "random process" as measured by a selected probability distribution model. Thus, pure randomness, pragmatically speaking, is herein considered to be a time series distribution for which no function, mapping or relation can be constituted that provides meaningful insight into the underlying structure of the distribution, but which at the same time is not chaos.

(2) Description of the Prior Art

Recent research has revealed a critical need for highly sparse data set time distribution analysis methods and apparatus separate and apart from those adapted for treating large sample distributions. This is particularly the case in applications such as naval sonar systems which require that input time series signal distributions be classified according to their structure, i.e., periodic, transient, random or chaotic. It is well known that large sample methods often fail when applied to small sample distributions, but that the same is not necessarily true for small sample methods applied to large data sets.

Very small data set distributions may be defined as those with less than about ten (10) to twenty-five (25) measurement (data) points. Such data sets can be analyzed mathematically with certain nonparametric discrete probability distributions, as opposed to large-sample methods which normally employ continuous probability distributions (such as the Gaussian).

The probability theory discussed herein and utilized by the present invention is well known. It may be found, for example, in works such as P.J. Hoel et al., Introduction to the Theory of Probability, Houghton-Mifflin, Boston, MA, 1971, which is hereby incorporated herein by reference.

Also, as will appear more fully below, it has been found to be important to treat white noise signals themselves as the time series signal distribution to be analyzed, and to identify the characteristics of that distribution separately. This aids in the detection and appropriate processing of received signals in numerous data acquisition contexts, not the least of which include naval sonar applications. Accordingly, it will be understood that prior analysis methods and apparatus analyze received time series data distributions from the point of view of attempting to find patterns or some other type of correlated data therein. Once such a pattern or correlation is located, the remainder of the distribution is simply discarded as being noise. It is believed that the present invention will be useful in enhancing the sensitivity of present analysis methods, as well as being useful on its own.

Various aspects related to the present invention are discussed in the following exemplary patents:

U.S. Patent No. 6,068,659, issued May 30, 2000, to Francis J. O'Brien, Jr., discloses a method for measuring and recording the relative degree of pical density, congestion, or crowding of objects dispersed in a three-dimensional space. A Population Density Index is obtained for the actual conditions of the objects within the space as determined from measurements taken of the objects. The Population Density Index is compared with values considered as minimum and maximum bounds, respectively, for the Population Density Index values. The objects within the space are then repositioned to optimize the Population Density Index, thus optimizing the layout of objects within the space.

U.S. Patent No. 5,506,817, issued April 9, 1996, to Francis J. O'Brien, Jr., discloses an adaptive statistical filter system for receiving a data stream comprising a series of data values from a sensor associated with successive points in time. Each data value includes a data component representative of the motion of a target and a noise component, with the noise components of data values associated with proximate points in time being correlated. The adaptive statistical filter system includes a prewhitener, a plurality of statistical filters of different orders, stochastic decorrelator and a selector. The prewhitener generates a corrected data stream comprising corrected data values, each including a data component and a time-correlated noise component. The plural statistical filters receive the corrected data stream and generate coefficient values to fit the corrected data stream to a polynomial of corresponding order and fit values representative of the degree of fit of corrected data stream to the polynomial. The stochastic decorrelator uses a spatial Poisson process statistical significance test to determine whether the fit values are correlated. If the test indicates the fit values are not randomly distributed, it generates decorrelated fit values using an autoregressive moving average methodology which assesses the noise components of the statistical filter. The selector receives the decorrelated fit values and coefficient values from the plural statistical filters and selects coefficient values from one of the filters in response to the decorrelated fit values. The coefficient values are coupled to a target motion analysis module which determines position and velocity of a target.

U.S. Patent No. 6,466,516 B1, issued October, 15, 2002, to Francis J. O'Brien, Jr. et al., discloses a method and apparatus for automatically characterizing the spatial arrangement among the data points of a three-dimensional time series distribution in a data processing system wherein the classification of said time series distribution is required. The method and apparatus utilize grids in Cartesian coordinates to determine (1) the number of cubes in the grids containing at least one input data point of the time series distribution; (2) the expected number of cubes which would contain at least one data point in a random distribution in said grids; and (3) an upper and lower probability of false alarm above and below said expected value utilizing a discrete binomial probability relationship in order to analyze the randomness characteristic of the input time series distribution. A labeling device also is provided to label the time series distribution as either random or nonrandom, and/or random or nonrandom within what probability, prior to its output from the invention to the remainder of the data processing system for further analysis.

U.S. Patent No. 6,397,234 B1, issued May 28, 2002, to Francis J. O'Brien, Jr. et. al., discloses a method and apparatus for automatically characterizing the spatial arrangement among the data points of a time series distribution in a data processing system wherein the classification of said time series distribution is required. The method and apparatus utilize a grid in Cartesian coordinates to determine (1) the number of cells in the grid containing at least-one input data point of the time series distribution; (2) the expected number of cells which would contain at least one data point in a random distribution in said grid; and (3) an upper and lower probability of false alarm above and below said expected value utilizing a discrete binomial probability relationship in order to analyze the randomness characteristic of the input time series distribution. A labeling device also is provided to label the time series distribution as either random or nonrandom, and/or random or nonrandom.

(3) Description of Another Department of the Navy Developments

A development in a related technological area made by the U.S. Department of the Navy is described in U.S. Patent No. 6,597,634 B1 issued July 22, 2003, to Francis J. O'Brien, Jr. et al, published as Publication No. US -2003-0043695-A1 on 6 March 2003, discloses a signal processing system to processes a digital signal converted from to an analog signal, which includes a noise component and possibly also an information component comprising small samples representing four mutually orthogonal items of measurement information representable as a sample point in a symbolic Cartesian four-dimensional spatial reference system. An information processing sub-system receives said digital signal and processes it to extract the information component. A noise likelihood determination sub-system receives the digital signal and generates a random noise assessment of whether or not the digital signal comprises solely random noise, and if not, generates an assessment of degree-of-randomness. The information processing system is illustrated as combat control equipment for undersea warfare, which utilizes a sonar signal produced by a towed linear transducer array, and whose mode operation employs four mutually orthogonal items of measurement information.

The above prior art and prior Department of the Navy development do not disclose a method which utilizes more than one statistical test to decide the structured properties of sparse data in order to maximize the likelihood of a correct decision in processing batches of the sparse data in real time operating submarine systems and/or other contemplated uses.

SUMMARY OF THE INVENTION

It is an object of the present invention to provide an improved two-stage method for analyzing sparse data.

It is yet another object of the invention to provide a two-stage method including an automated measurement of the spatial arrangement among a very small number of points, object, measurements or the like whereby an ascertainment of the noise degree (i.e., randomness) of the time series distribution may be made by conjoint methods of mathematical analysis.

It is yet another object of the invention to provide a method and apparatus useful in naval sonar systems which require acquired signal distributions to be classified according to their structure (i.e., periodic, transient, random, or chaotic) in the processing and use of those acquired signal distributions as indications of how and from where they were originally generated.

Further, it is an object of the invention to provide a method and apparatus capable of labeling a time series distribution with (1) an indication as to whether or not it is random in structure, and (2) an indication as to whether or not it is random within a probability of false alarm of a specific randomness calculation.

These and other objects, features, and advantages of the present invention will become apparent from the drawings, the descriptions given herein, and the appended claims. However, it will be understood that above listed objects and advantages of the invention are intended only as an aid in understanding certain aspects of the invention, are not intended to limit the invention in any way, and do not form a comprehensive or exclusive list of objects, features, and advantages.

With the above and other objects in view, as will hereinafter more fully appear, a feature of the invention is the provision of conjoint random process detection methods and subsystem for use in a naval sonar signal processing/data processing system. In a preferred embodiment, the random process (white noise) detection subsystem includes an input for receiving a time series distribution of data points expressed in Cartesian coordinates. This set of data points will be characterized by no more than a maximum number of points having a value (amplitude) between a maximum and a minimum value received within a preselected time interval. A hypothetical representation of a white noise time series signal distribution in Cartesian space is illustratively shown in FIG. 1. The invention is specifically adapted to analyze both selected portions of such time series distributions, and the entirety of the distribution depending upon the sensitivity of the randomness determination which is required in any particular instance.

The input time series distribution of data points is received by a display/operating system adapted to accommodate a pre-selected number of data points N having a value (amplitude for sonar signals and the like) within certain limits within a pre-selected time interval. The display/operating system then creates a virtual window around the input data distribution, and divides the geometric area of the virtual window into a grid consisting of cells each having the same geometric shape and an equal enclosed area. Ideally, the grid fills the entire area of the window, but if it does not, the unfilled portion of the window is disregarded in the randomness determination.

An analysis device then examines each cell to determine whether or not one or more of the data points of the input time series distribution is located therein. Thereafter, a counter calculates the number of occupied cells. Also, the number of cells which would be expected to be occupied in the grid for a totally random distribution is statistically predicted by a computer device according to known Poisson probability process and binomial theory equations, and application of the Central Limit Theorem, constituting the test of randomness. In addition, the statistical bounds of the predicted value are calculated based upon a known distinct discrete binomial criteria.

A comparator is then used to determine whether or not the actual number of occupied cells in the input time series distribution is the same as the statistically predicted number of cells for a random distribution. If it is, the input time series distribution is characterized as random. If it is not, the input time series distribution is characterized as nonrandom.

Thereafter, the characterized time series distribution is labeled as random or nonrandom, and/or as random or nonrandom within a pre-selected probability rate of the expected randomness value prior to being output back to the remainder of the data processing system. In the naval sonar signal processing context, this output either alone, or in combination with overlapping similarly characterized time series signal distributions, will be used to determine whether or not a particular group of signals is white noise. If that group of signals is white noise, it commonly will be deleted from further data processing. Hence, it is contemplated that the present invention, which is not distribution dependent in its analysis as most prior art methods of signal analysis are, will be useful as a filter or otherwise in conjunction with current data processing methods and equipment.

In the above regards, it should be understood that the statistical bounds of the predicted number of occupied cells in a random distribution (including cells occupied by mere chance) mentioned above may be determined by a second calculator device using a so-called probability of false alarm rate. In this case, the actual number of occupied cells is compared with the number of cells falling within the statistical boundaries of the predicted number of occupied cells for a random distribution in making the randomness determination. This alternative embodiment of the invention has been found to increase the probability of being correct in making a randomness determination for any particular time series distribution of data points by as much as 60%. For instance in one version of this alternative embodiment, the method may comprise one or more method steps such as, for example only, creating a virtual window having a two-dimensional area containing a distribution of data points of the sparse data for a selected time period and/or subdividing substantially the entirety of the area of the virtual window into a plurality k of cells wherein each of the plurality k of cells have the same polygonal shape and define the same area value.

Additional steps may comprise determining a quantity Θ wherein Θ represents an expected proportion of the plurality k of cells which will be nonempty in a random distribution. When Θ is less than a pre-selected value, then the method may comprise utilizing a Poisson distribution to determine a first mean of the data points. When Θ is greater than the pre-selected value, then the method may comprise utilizing a binomial distribution to determine a second mean of the data points. The method may further comprise computing a probability p from the first mean or the second mean, depending on whether Θ is greater than or less than the pre-selected value. Other steps may comprise determining a false alarm probability α based on a total number of the plurality of k cells. By comparing p with α, the method may be utilized to then determine whether to characterize the sparse data as noise or signal.

In one example, the pre-selected amount discussed above is equal to 0.10 such that if $\Theta \leq 0.10$, then the Poisson distribution is utilized, and if $\Theta > 0.10$, then the binomial distribution is utilized. Also, in one embodiment, the step of determining a probability of false alarm rate α comprises setting the alarm rate α equal to 0.01 when the total number of the plurality of k cells is greater than 25, and/or determining a probability of false alarm rate α comprises setting the alarm rate α equal to 0.05 when the total number of the plurality of k of cells is greater than or equal to 5 and less than or equal to 25 and/or determining a probability of false alarm rate α comprises setting the alarm rate α equal to 0.10 when the total number of the plurality of k cells is less than 5.

The above and other novel features and advantages of the invention, including various novel details of construction and combination of parts will now be more particularly described with reference to the accompanying drawings and pointed out by the claims. It will be understood that the particular device and method embodying the invention is shown and described herein by way of illustration only, and not as limitations on the invention. The principles and features of the invention may be employed in numerous embodiments without departing from the scope of the invention in its broadest aspects.

BRIEF DESCRIPTION OF THE DRAWINGS

Reference is made to the accompanying drawings in which is shown an illustrative embodiment of the apparatus and method of the invention, from which its novel features and advantages will be apparent to those skilled in the art, and wherein:

FIG. 1 is a hypothetical depiction in Cartesian coordinates of a representative white noise (random) time series signal distribution;

FIG. 2 is a hypothetical illustrative representation of a virtual window in accordance with the invention divided into a grid of square cells each having a side of length δ , and an area of δ^2;

FIG. 3 is a block diagram representatively illustrating the method steps of the invention;

FIG. 4 is a block diagram representatively illustrating an apparatus in accordance with the invention; and

FIG. 5 is a table showing an illustrative set of discrete binomial probabilities for the randomness of each possible number of occupied cells of a particular time series distribution within a specific probability of false alarm rate of the expected randomness number.

DESCRIPTION OF THE PREFERRED EMBODIMENT

Referring now to the drawings, a preferred embodiment of the method and apparatus of the invention will be presented first from a theoretical perspective, and thereafter, in terms of a specific example. In this regard, it is to be understood that all data points are herein assumed to be expressed and operated upon by the various apparatus components in a Cartesian coordinate system. Accordingly, all measurement, signal and other data input existing in terms of other coordinate systems is assumed to have been re-expressed in a Cartesian coordinate system prior to its input into the inventive apparatus or the application of the inventive method thereto.

The invention starts from the preset capability of a display/operating system 8 (FIG. 4) to accommodate a set number of data points N in a given time interval ≅t. The value (amplitude) of each data point in each time series distribution falls within limits which may be expressed as ≅Y= max (Y) - min (Y). A representation of a time series distribution of random sonar input data points 4 is shown in FIG. 1. A subset of this overall time series data distribution would normally be selected for analysis of its signal component distribution by this invention.

For purposes of mathematical analysis of the signal components, it is assumed that the product/quantity given by ≅t * ΔY = [max(t) - min(t)] * [max(Y) - min(Y)] will define the window "geometric area" with respect to the quantities in the analysis subsystem. The sides of the ≅t * ΔY window are drawn parallel to the time axis and amplitude axis, respectively, although other window shapes may be employed (such as a convex polygon) without departure from the invention in its broadest aspects. Then, for substantially the total area of the display region, a Cartesian partition is superimposed on the region with each partition being a small square of side δ (see, FIG. 2). The measure of δ will be defined herein as:

$$\delta = (\Delta t * \Delta Y / k)^{1/2} \qquad (1)$$

The quantity k represents the total number of small squares each of area δ^2 created in the area $\cong t * \Delta Y$. Incomplete squares 6 are ignored in the analysis. The quantity of such squares which it is desired to occupy with at least one data point from an input time series distribution is determined using the following relationship wherein N is the maximum number of data points in the time series distribution, $\cong t$ and ΔY are the Cartesian axis lengths, and the side lengths of each of the squares is δ:

$$k_I = \mathrm{int}(\Delta t/\delta_I) * \mathrm{int}(\Delta Y/\delta_I) \qquad (2)$$

where int is the integer operator,

$\delta_I = \sqrt{}[(\Delta t * \Delta Y)/k_o]$, and

$k_o = k_1$ if $|N - k_1| \leq |N - k_2|$ or

$= k_2$ otherwise

where

$k_1 = [\mathrm{int}(N^{1/2})]^2$

$k_2 = [\mathrm{int}(N^{1/2}) + 1]^2$

$k_{II} = \mathrm{int}(\Delta t/\delta_{II}) * \mathrm{int}(\Delta Y/\delta_{II})$

where

$\delta_{II} = \sqrt{}[(\Delta t * \Delta Y)/N]$

$\therefore$ $k = k_I$ if $K_I > K_{II}$

$k = k_{II}$ if $K_I < K_{II}$

$k = \max(k_I, k_{II})$ if $K_I = K_{II}$

where

$K_I = \delta_I^2 k_I /(\Delta t \cdot \Delta Y) \leq 1$ and

$K_{II} = \delta_{II}^2 k_{II} /(\Delta t \cdot \Delta Y) \leq 1$

In cases with very small amplitudes, it may occur that $\mathrm{int}(\Delta Y/\delta_I) \leq 1$ or $\mathrm{int}(\Delta Y/\delta_{II}) \leq 1$. In such cases, the solution is to round off either quantity to the next highest value (i.e., ≥ 2). This weakens the theoretical approach, but it allows for practical measurements to be made.

Thus, for example, if Δt (or N)=30, and ΔY=20, then k=24 and δ=5.0. Accordingly, $k * \delta^2 = 24 * 25 = 600 = \cong t * \Delta Y$. In essence, therefore, the above relation defining the value k selects the number of squares of length δ and area δ^2 which fill up the total space $\cong t * \Delta Y$ to the greatest extent possible (i.e., ideally $k * \delta^2 \cong t * \Delta Y$).

From the selected partitioning parameter k, the region (area) $\cong t * \Delta Y$ is carved up into k squares with the length of each square being δ as defined above. In other words, the horizontal (or time) axis is marked off into intervals, exactly $\mathrm{int}(\Delta t/\delta)$ of them, so that the time axis has the following arithmetic sequence of cuts (assuming that the time clock starts at $\Delta t = 0$):

$$0, \delta, 2\delta, \ldots, \mathrm{int}(\Delta t/\delta) * \delta \qquad (3)$$

Likewise, the vertical (or measurement or amplitude) axis is cut up into intervals, exactly int($\Delta Y/\delta$) of them, so that the vertical axis has the following arithmetic sequence of cuts:

$$\min(Y),\ \min(Y)+\delta,\ \ldots,\min(Y)+\mathrm{int}(\Delta Y/\delta) * \delta = \max(Y), \qquad (4)$$

where min is the minimum operator and δ is defined as above.

Based on the Poisson point process theory for a measurement set of data in a time interval Δt of measurement magnitude ΔY, that data set is considered to be purely random (or "white noise") if the number of partitions k are nonempty (i.e., contain at least one data point of the time series distribution thereof under analysis) to a specified degree. The expected number of nonempty partitions in a random distribution is given by the relationship:

$$k * \Theta = k * (1-e^{-N/k}) \qquad (5)$$

where the quantity Θ is the expected proportion of nonempty partitions in a random distribution and N/k is " the parameter of the spatial Poisson process" corresponding to the average number of points observed across all subspace partitions.

The boundary, above and below $k * \Theta$, attributable to random variation and controlled by a false alarm rate is the so-called "critical region" of the test. The quantity Θ not only represents (a) the expected proportion of nonempty partitions in a random distribution, but also (b) the probability that one or more of the k partitions is occupied by pure chance, as is well known to those in the art. The boundaries of the random process are determined in the following way.

Let M be a random variable representing the integer number of occupied cells (partitions) as illustratively shown in FIG. 2. Let m be an integer (sample) representation of M. Let m_1 be the quantity forming the lower random boundary of the statistic $k * \Theta$ given by the binomial criterion:

$$P(M \leq m) \leq \alpha_0/2, \min(\alpha/2 - \alpha_0/2) \quad (6)$$

where,

$P(M \leq m) = \sum B(m;k,\Theta)$ from $m=0$ to $m=m_1$, and

k and Θ are defined as above.

$B(m;k,\Theta)$ is the binomial probability function given as:

$$B(m;k,\Theta) = (k,m)(\Theta)^m(1-\Theta)^{k-m} \quad (7)$$

where (k,m) is the binomial coefficient, (k,m)=k!/m!(k-m)!

and $\sum B(m;k,\Theta)$ from m=0 to m=k equals 1.0.

The quantity α_o is the probability of coming closest to an exact value of the pre-specified false alarm probability α, and m_1 is the largest value of m such that $P(M \leq m) \leq \alpha_0/2$. It is an objective of this method to minimize the difference between α and α_0. The recommended values of α (the probability false alarm rate) for differing values of spatial subsets k are as follows:

$$\text{If } k > 25, \text{ the } \alpha = 0.01;$$
$$\text{If } 5 \leq k \leq 25, \text{ then } \alpha = 0.05; \text{ and} \quad (8)$$
$$\text{If } k < 5, \text{ then } \alpha = 0.10$$

The upper boundary of the random process is called m_2, and is determined in a manner similar to the determination of m_1.

Thus, let m_2 be the upper random boundary of the statistic $k*\Theta$ given by:

$$P(M \geq m) \leq \alpha_o/2, \min(\alpha/2-\alpha_o/2) \quad (9)$$

where

$$P(M \geq m) = \sum_{m=m_2}^{k} B(m;k,\Theta) \leq \alpha_o / 2$$

or

$$P(M \geq m) = 1 - \sum_{m=0}^{m_2} B(m;k,\Theta) \leq \alpha_o / 2$$

α_o is the probability of coming closest to an exact value of the pre-specified false alarm probability α, and m_2 is the largest value of m such that $P(M \geq m) \leq \alpha_o/2$. It is an objective of the invention to minimize the difference between α and α_o.

Hence, the subsystem determines if the signal structure contains m points within the "critical region" warranting a determination of "random".

The subsystem also assesses the random process hypothesis by testing:

$$H_o: \overline{P} = \Theta \text{ (Noise)}$$

$$H_1: \overline{P} \neq \Theta \text{ (Signal + Noise)} \qquad (10)$$

Where $\overline{P} = m/k$ is the sample proportion of signal points contained in the k subregion partitions expected to be occupied by a truly random (stochastic) spatial distribution. As noted above, FIG. 1 shows what a hypothetical white noise (random) distribution looks like in Cartesian time-space.

Thus, if $\Theta \approx \overline{P} = m/k$, the observed distribution conforms to a random distribution corresponding to "white noise".

The estimate for the proportion of k cells occupied by N measurements ($\overline{P}$) is developed in the following manner. Let each of the k cells of length δ be denoted by C_{ij} and the number of objects observed in each C_{ij} cell be denoted card (C_{ij}) where card means "cardinality" or subset count. C_{ij} is labeled from left to right starting at the lower left-hand corner C_{11}, C_{12}, ..., C_{46} (see FIG. 2).

Next to continue the example for k = 24 shown in FIG. 2, define the following count quantity for the 6 x 4 partition comprising whole square subsets:

$$X_{ij} = 1 \text{ if card } (C_{ij}) > 0;\ i = 1 \text{ to } 4,\ j = 1 \text{ to } 6$$

$$X_{ij} = 0 \text{ if card } (C_{ij}) = 0;\ i = 1 \text{ to } 4,\ j = 1 \text{ to } 6 \tag{11}$$

where card is the cardinality or count operator. X_{ij} is a dichotomous variable taking on the individual values of 1 if a cell C_{ij} has one or more objects present, and a value of 0 if the box is empty.

Then calculate the proportion of 24 cells occupied in the partition region:

$$\overline{P} = 1/24\sum\sum X_{ij} \tag{12}$$

where the sums are taken from j = 1 to 6 and i = 1 to 4, respectively.

The generalization of this example to any sized table is obvious, and within the scope of the present invention. For the general case, it will be appreciated that, for the statistics X_{ij} and C_{ij} the index j runs from 1 to int($\Delta t/\delta$) and the index i runs from 1 to int($\Delta Y/\delta$).

In addition, another measure useful in the interpretation of outcomes is the R ratio, defined as the ratio of observed to expected occupancy rates:

$$R = m/(k * \Theta) = \overline{P}/\Theta \tag{13}$$

The range of values for R indicate:

R < 1, clustered distribution

R = 1, random distribution; and

R > 1, uniform distribution.

In actuality, R may never have a precise value of 1.

A rigorous statistical procedure has been developed to determine whether the observed R-value is indicative of "noise" or "signal". The procedure renders quantitatively the interpretations of the R-value whereas the prior art has relied primarily on intuitive interpretation or ad hoc methods, which can be erroneous.

In this formulation, one of two statistical assessment tests is utilized depending on the value of the parameter Θ.

If $\Theta \leq 0.10$, then a Poisson distribution is employed. To apply the Poisson test, the distribution of the N sample points is observed in the partitioned space. It will be appreciated that a data sweep across all cells within the space will detect some of the squares being empty, some containing $k = 1$ points, $k = 2$ points, $k = 3$ points, and so on. The number of points in each k category is tabulated in a table such as follows:

Frequency Table of Cell Counts

k (number of cells with points)	N_k (number of points in k cells)
0	N_0
1	N_1
2	N_2
3	N_3
$\vdots$	$\vdots$
K	N_k

From this frequency table, two statistics are of interests for the Central Limit Theorem approximation:

The "total", $Y=\sum_{k=0}^{K} kN_k$, and (14)

the sample mean, $\mu_0 = \dfrac{\sum_{k=0}^{K} kN_k}{\sum_{k=0}^{K} N_k}$.

Then, if $\Theta \le 0.10$, the following binary hypothesis is of interest:

$$\begin{aligned} &H_0 : \mu = \mu_0 (NOISE) \\ &H_1 : \mu \neq \mu_0 (SIGNAL) \end{aligned} \tag{15}$$

The Poisson test statistic, derived from the Central Limit Theorem, Eq. (3) is as follows:

$$Z_P = \frac{Y - N\mu_0}{\sqrt{N\mu_0}}, \qquad (k>25) \tag{16}$$

where

$$Y = \sum_{k=0}^{K} kN_k ,$$

and N is the sample size. Then

$$\mu_0 = \frac{\sum_{k=0}^{K} kN_k}{\sum_{k=0}^{K} N_k} \tag{17}$$

is the sample mean and sample variance. (It is well known that $\mu = \sigma^2$ in a Poisson distribution).

The operator compares the value of Z_p against a probability of False Alarm α. α is the probability that the null hypothesis (NOISE) is rejected when the alternative (SIGNAL) is the truth.

The probability of the observed value Z_p is calculated as:

$$p = P\left(|z_p| \le Z\right) = 1 - \frac{1}{\sqrt{2\pi}} \int_{-|z_p|}^{+|z_p|} \exp(-.5x^2)dx \qquad (18)$$

where $|x|$ means "absolute value" as commonly used in mathematics.

The calculation of Eq. 6, as known to those skilled in the art, is performed in a standard finite series expansion.

On the other hand, if $\Theta > .10$, the invention dictates that the following binary hypothesis set prevail:

$$\begin{aligned} &H_0 : \mu = k\theta(NOISE) \\ &H_1 : \mu = k\theta(SIGNAL) \end{aligned} \qquad (19)$$

The following binomial test statistic is employed to test the hypothesis:

$$Z_B = \frac{m \pm c - k\theta}{\sqrt{k\theta(1-\theta)}} \qquad (20)$$

where $c = 0.5$ if $X < \mu$ and $c = -0.5$ if $X > \mu$ (Yates Continuity correction factor used for discrete variables).

The quantities of Z_B have been defined previously.

The probability of the observed value Z_B is calculated as

$$p = P(|z_B| \leq Z) = 1 - \frac{1}{\sqrt{2\pi}} \int_{-|z_B|}^{+|z_B|} \exp(-.5x^2)dx \tag{21}$$

in a standard series expansion.

For either test statistic, Z_p or Z_B, the following decision rule is used to compare the false alarm rate α with the observed probability of the statistic, p:

$$\begin{aligned} &if\ p \geq \alpha \Rightarrow NOISE \\ &If\ p < \alpha \Rightarrow SIGNAL \end{aligned} \tag{22}$$

Thus, if the calculated probability value $p > \alpha$, then the spatial distribution is deemed "noise"; otherwise the data is characterized as "signal" by the Rtest.

The R statistic may be used in conjunction with the formulation just described involving the binomial probability distribution and false alarm rate in deciding to accept or reject the "white noise" hypothesis - or it may be used as the sole determinant. In summary, operators may find the role of the R statistic to be more intuitively useful. Intelligent operators will always employ a plot of time series and its amplitude, in order to eliminate the obvious situations of a "perfect" functional or relational form being analyzed which can be easily seen to be nonrandom, but which computationally may be concluded to be random. The enhancement of the R statistic by means of a statistical significance test lessens the likelihood of such a perfect relation going undetected, this being a significant enhancement over the prior art.

EXAMPLE

Having thus explained the theory of the invention, an example thereof will now be presented for purposes of further illustration and understanding (see, FIGS. 3 and 4). A value for N is first selected, here N = 30 (step 100, FIG. 3). A time series distribution of data points is then read into a display/operating subsystem 8 adapted to accommodate a data set of size N from data processing system 10 (step 102). An illustrative example of the data points of the time series distribution from the field of processing naval sonar signals would be the input time series signal distribution derived from processing acoustic input signals in well known ways to indicate, for example, relative angular bearing of the acoustic source to the course of a submarine which is the platform for the acoustic receiver. Another illustrative example would be data points derived from questionnaires in sociological research. Thereafter, the absolute value of the difference between the largest and the smallest data points ΔY is determined by a first comparator device 12 (step 104). In this example, it will be assumed that $N = \Delta t = 30$ measurements with a measured amplitude of $\Delta Y = 20$ units. The N and ΔY values are then used by window creating device 14 to create a virtual window in a naval sonar information display system, or in an operating system of any other naval systems which employ sonar information as an input. The virtual window encloses the input time series distribution. The size of the window so created is $\Delta t * \Delta y = 600$ (step 106). Such information display systems and other systems employing sonar information as an input are sometimes in this specification and in the appendant claims collectively referred to as naval sonar information utilization systems.

Thereafter, as described above, the virtual window is divided by the window creating device 14 into a plurality k of cells C_{ij} (see FIG. 4), each cell having the same geometric shape and enclosing an equal area so as to substantially fill the virtual window containing the input time series distribution set of data points (step 108). The value of k is established by the relationships:

$$k = \mathrm{int}(\Delta t/\delta) * \mathrm{int}(\Delta Y/\delta) = 6 * 4 = 24$$

$$\delta = \sqrt{(\Delta t * \Delta Y)/k} = 5.0 \qquad (23)$$

Thus, the 600 square unit space of the virtual window is partitioned into 24 cells of side 5.0 so that the whole space is filled ($k * \delta^2 = 600$). The time-axis arithmetic sequence of cuts are: 0, 5, ..., $\mathrm{int}(\Delta t/\delta) * \delta = 30$. The amplitude axis cuts are: $\min(Y)$, $\min(Y) + \delta$, ..., $\min(Y) + \mathrm{int}(\Delta Y/\delta) * \delta = \max(Y)$.

Next, the probability false alarm rate is set at step 110 according to the value of k as discussed above. More particularly, in this case $\alpha = 0.01$, and the probability of a false alarm within the critical region is $\alpha/2 = 0.005$.

The randomness count is then calculated by first computing device 16 at step 112 according to the relation $k * \Theta = k * (1-e^{-N/k})$ which in this example equals 0.713. Therefore, the number of cells expected to be nonempty in this example if the input time series distribution is random is about 17.

The binomial distribution discussed above is then calculated by a second computing device 18 according to the relationships discussed above (step 114, FIG. 3). Representative values for this distribution are shown in FIG. 5 for each number of possible occupied cells m.

The upper and lower randomness boundaries then are determined, also by second calculating device 18. Specifically, the lower boundary is calculated using m_1 from FIG. 5 (step 116). Then, computing the binomial probabilities results in $P(M \leq 10) = .0025$. Thus, the lower bound is $m_1 = 10$. FIG. 5 also shows the probabilities for $\Theta = .713$, $k = 24$.

The upper boundary, on the other hand, is the randomness boundary m_2 from the criterion $P(M \geq m) \leq \alpha_o/2$. Computing the binomial probabilities gives $P(M \geq 23) = .0032$; hence $m_2 = 23$ is taken as the upper bound (step 118). The probabilities necessary for this calculation also are shown in FIG. 5.

Therefore, the critical region is defined in this example as $m_1 \leq 10$, and $m_2 \geq 23$ (step 120).

The actual number of cells containing one or more data points of the time series distribution determined by analysis/counter device 20 (step 122, FIG. 3) is then used by divider 22 and a second comparator 24 in the determination of the randomness of the distribution (step 124, FIG. 3). Specifically, using m = 16 as an example, it will be seen that $\overline{P} = m/k = 0.667$, and that $R = \overline{P}/\Theta = 0.667/0.713 = 0.93$. This value is close to the randomness boundary without consideration of the discrete binomial probability calculations discussed above.

Branching to step 123 (FIG. 3) which the sparse data decision logic module performs, the R statistic value of 0.93 is evaluated statistically. A more precise indicator is obtained by applying the significance test in accord with the present invention, as described earlier. For this calculation, we note that $\theta = .713$, which invokes the Binomial probability model to test the hypothesis:

$$\begin{aligned} H_0 &: \mu = k\theta(NOISE) \\ H_1 &: \mu = k\theta(SIGNAL) \end{aligned} \qquad (24)$$

In this case, $k\theta = 17.12$. Thus, applying the Binomial test gives:

$$Z_B = \frac{m \pm c - k\theta}{\sqrt{k\theta(1-\theta)}}$$

$$= \frac{16 - .5 - 17.12}{\sqrt{24(.713)(1-.713)}} \approx -.43 \qquad (25)$$

The p value is computed to be:

$$p = P\left(\left|z_B\right| \le Z\right) = 1 - \frac{1}{\sqrt{2\pi}} \int_{-|-.43|}^{+|-.43|} \exp(-.5x^2)dx = .66 \qquad (26)$$

Since p = .66 and $\alpha = 0.1$, and since $p \ge \alpha$, we conclude that the R test shows the volumetric data to be random (NOISE only, with 99% certainty) with the value of R= .93 computed for this spatial distribution in 32-space.

It is also worth noting in this regard that the total probability is 0.0023 + .0032 = .0055, which is the probability of being wrong in deciding "random". This value is less than the probability of a false alarm. Thus, the actual protection against an incorrect decision is much higher (by about 45%) than the a priori sampling plan specified.

Since m = 16 falls inside of the critical region, i.e., $m_1 \le 16 \le m_2$, the decision is that the data represent an essentially white noise distribution (step 126). Accordingly, the distribution is labeled at step 128 by the labeling device 26 as a noise distribution, and transferred back to the data processing system 10 for further processing. In the naval sonar situation, a signal distribution labeled as white noise would be discarded by the processing system, but in some situations a further analysis of the white noise nature of the distribution would be possible. Similarly, the invention is contemplated to be useful as an improvement on systems which look for patterns and correlations among data points. For example, overlapping time series distributions might be analyzed in order to determine where a meaningful signal begins and ends.

It will be understood that many additional changes in the details, materials, steps and arrangement of parts, which have been herein described and illustrated in order to explain the nature of the invention, may be made by those skilled in the art within the principles and scope of the invention as expressed in the appended claims.

Attorney Docket No. 83995

ENHANCED SYSTEM FOR DETECTION OF RANDOMNESS IN SPARSE TIME SERIES DISTRIBUTIONS

ABSTRACT OF THE DISCLOSURE

A two-step method and apparatus are provided for automatically characterizing the spatial arrangement among the data points of a time series distribution in a data processing system wherein the classification of said time series distribution is required. In a first stage, the method and apparatus utilize a grid in Cartesian coordinates to determine (1) the number of cells in the grid containing at least one input data point of the time series distribution; (2) the expected number of cells which would contain at least one data point in a random distribution in said grid; and (3) an upper and lower probability of false alarm above and below said expected value utilizing a discrete binomial probability relationship in order to analyze the randomness characteristic of the input time series distribution. In a second stage, a statistical test of significance of the sparse data is utilized to determine the existence of noise and/or signal whereby a comparison of the results from the first stage and the second stage increase the probability of distinguishing noise from signal.

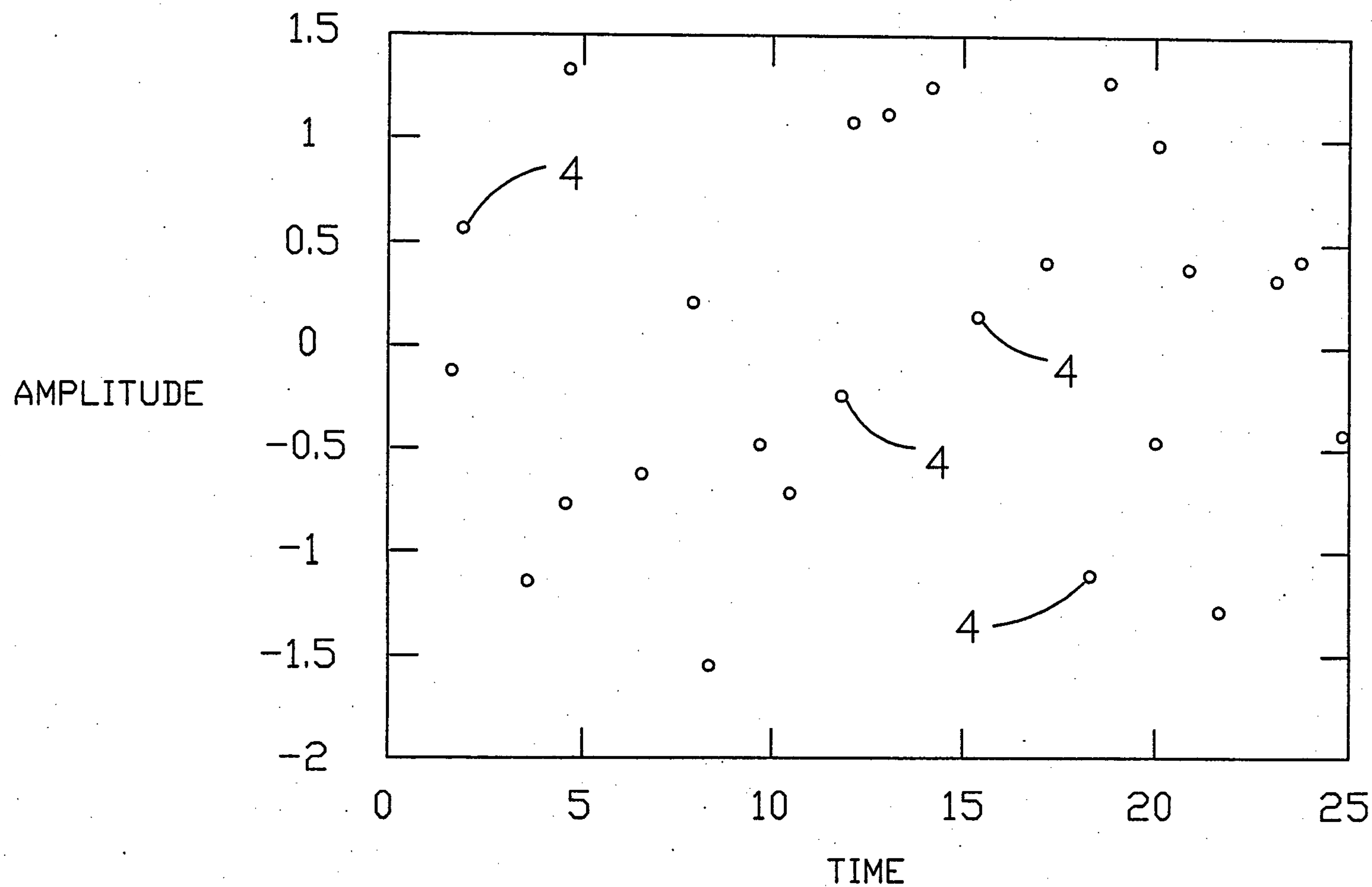

FIG. 1

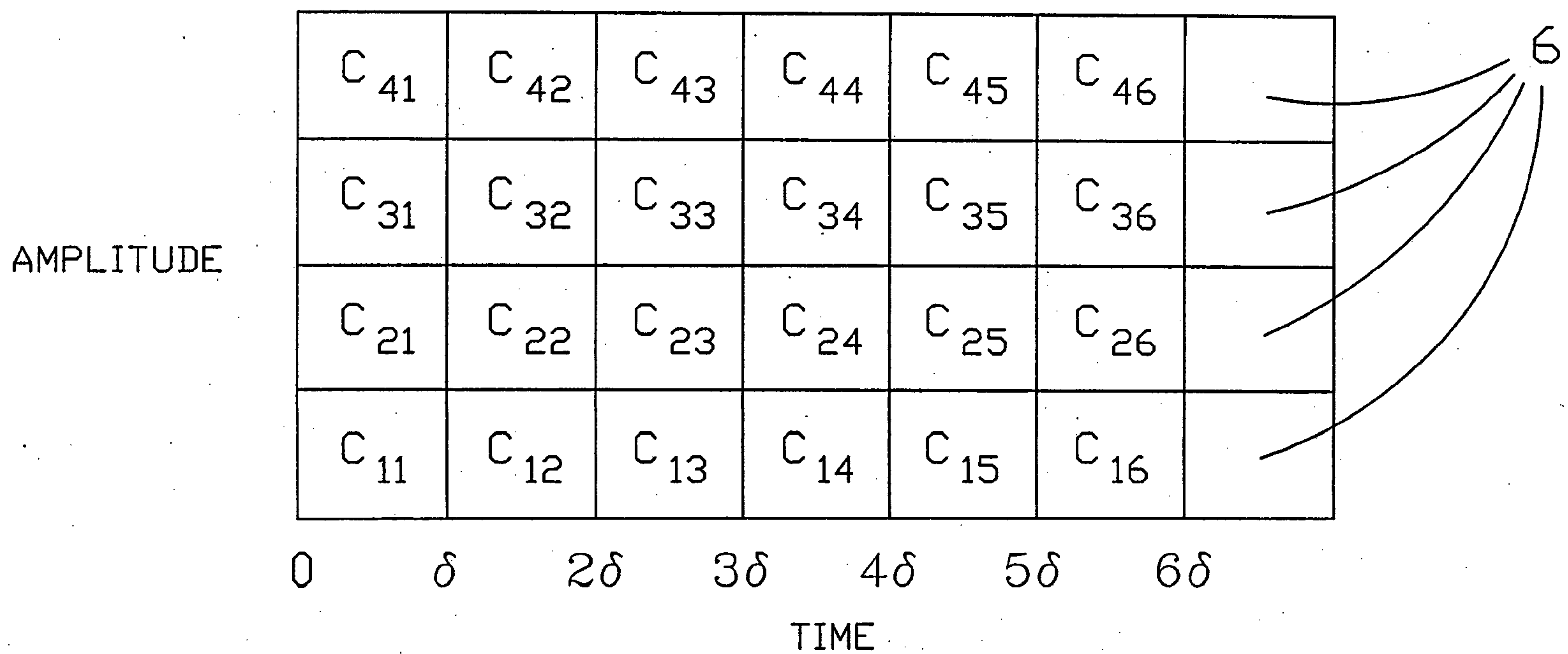
6
AMPLITUDE
C 41
C 42
C 43
C 44
C 45
C 46
C 31
C 32
C 33
C 34
C 35
C 36
C 21
C 22
C 23
C 24
C 25
C 26
C 11
C 12
C 13
C 14
C 15
C 16
0
δ
2δ
3δ
4δ
5δ
6δ
TIME

FIG. 2

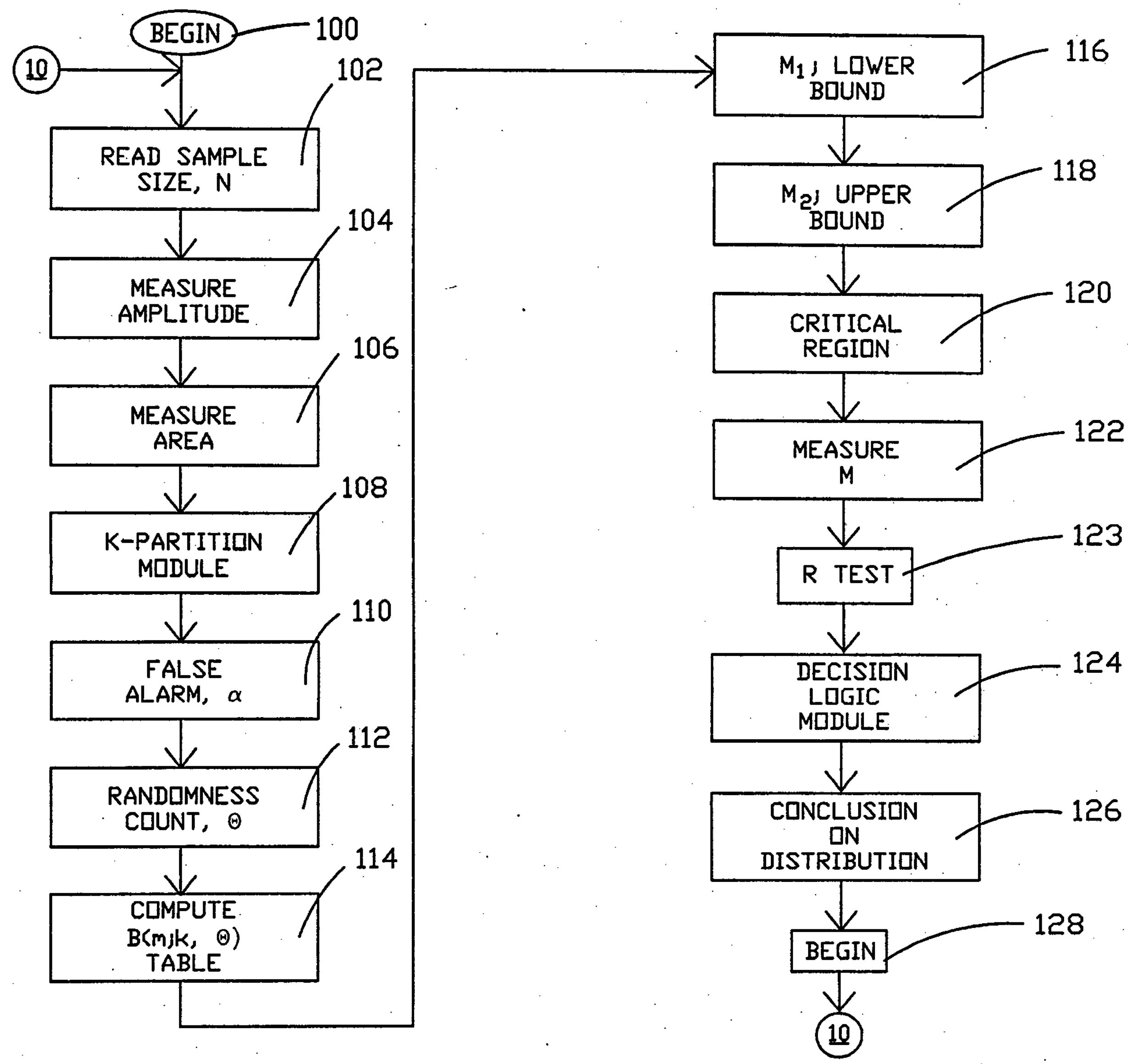
BEGIN
100
10
READ SAMPLE SIZE, N
102
MEASURE AMPLITUDE
104
MEASURE AREA
106
K-PARTITION MODULE
108
FALSE ALARM, α
110
RANDOMNESS COUNT, Θ
112
COMPUTE B(m,k, Θ) TABLE
114
M1, LOWER BOUND
116
M2, UPPER BOUND
118
CRITICAL REGION
120
MEASURE M
122
R TEST
123
DECISION LOGIC MODULE
124
CONCLUSION ON DISTRIBUTION
126
BEGIN
128
10

FIG. 3

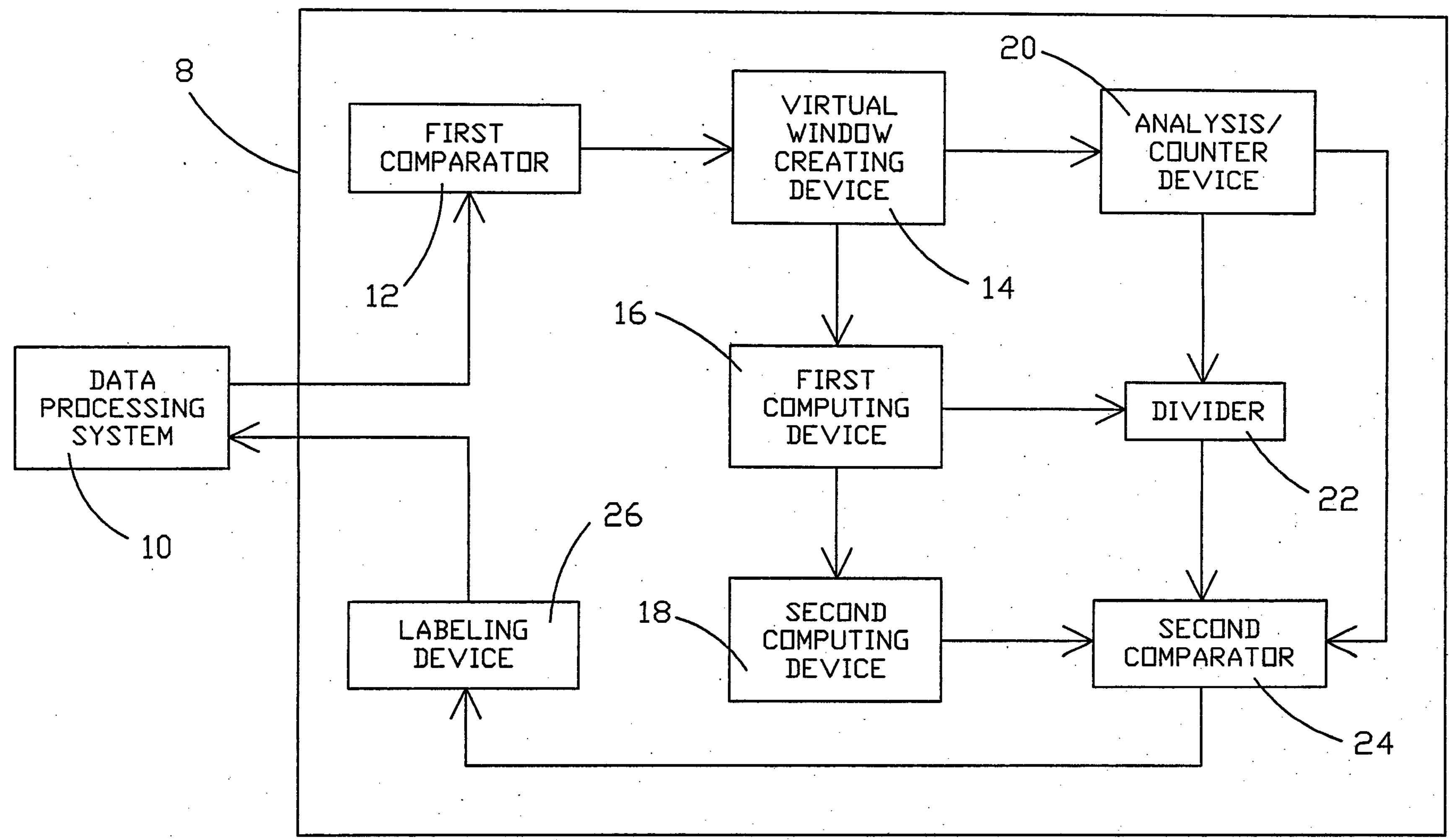
8
FIRST COMPARATOR
12
VIRTUAL WINDOW CREATING DEVICE
14
20
ANALYSIS/ COUNTER DEVICE
DATA PROCESSING SYSTEM
10
16
FIRST COMPUTING DEVICE
DIVIDER
22
26
LABELING DEVICE
18
SECOND COMPUTING DEVICE
SECOND COMPARATOR
24

FIG. 4

BINOMIAL TABLE FOR $k=24, \Theta=.713, \alpha=.01$

m	$P(M=m)=\binom{k}{m}\Theta^m(1-\Theta)^{k-m}$	$P(M\le m)=\sum_0 P(M=m)$ (CUMULATIVE)	$P(M\ge m)$
0	0	0	
1	0	0	
2	0	0	
3	0	0	
4	0	0	
5	0	0	
6	0	0	
7	0	0	
8	.0001	.0001	
9	.0005	.0006	
10	.0017	.0023(m_1) , $P(M\le m)\le\alpha_0$	
11	.0053	.0076	
12	.0144	.0220	
13	.0334	.0551	
14	DATA NOT SHOWN FOR m=14 to 20		
15			
16			
17			
18			
19			
20			
21	.0397	.9833	.0564
22	.0135	.9968	.0167
23	.0029	.9997(m_2) , $P(M\ge m)\le\alpha_0/2$	.0032
m=k=24	.0003	1.0000	.0003

FIG. 5