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(54) **TECHNIQUE FOR DIAGNOSING ATTENTION DEFICIT HYPERACTIVITY DISORDER**

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(57) **ABSTRACT**

A method of determining whether an individual has Attention Deficit Hyperactivity Disorder, comprising: sampling the peripheral skin temperature of an individual during a pre-determined time interval when the subject is in an inactive state to provide sampled peripheral skin temperature data; dividing the data into a series of windows of equal time intervals across the pre-determined time interval; for each window applying a fast Fourier Transform algorithm (FFT) to produce a data set of FFT magnitudes and determining a range of FFT magnitude variation, which is the difference between the FFT magnitude at a frequency f_{max} and the FFT magnitude at the minimum frequency; aggregating the magnitude ranges for one of or less than all of the windows to produce an aggregate magnitude range value which is indicative of whether or not the individual has ADHD; wherein prior to sampling, if one or more of the following noises are present it is eliminated through manual or automatic means: self diversion, external stimulation, technical problems, sleep problems.

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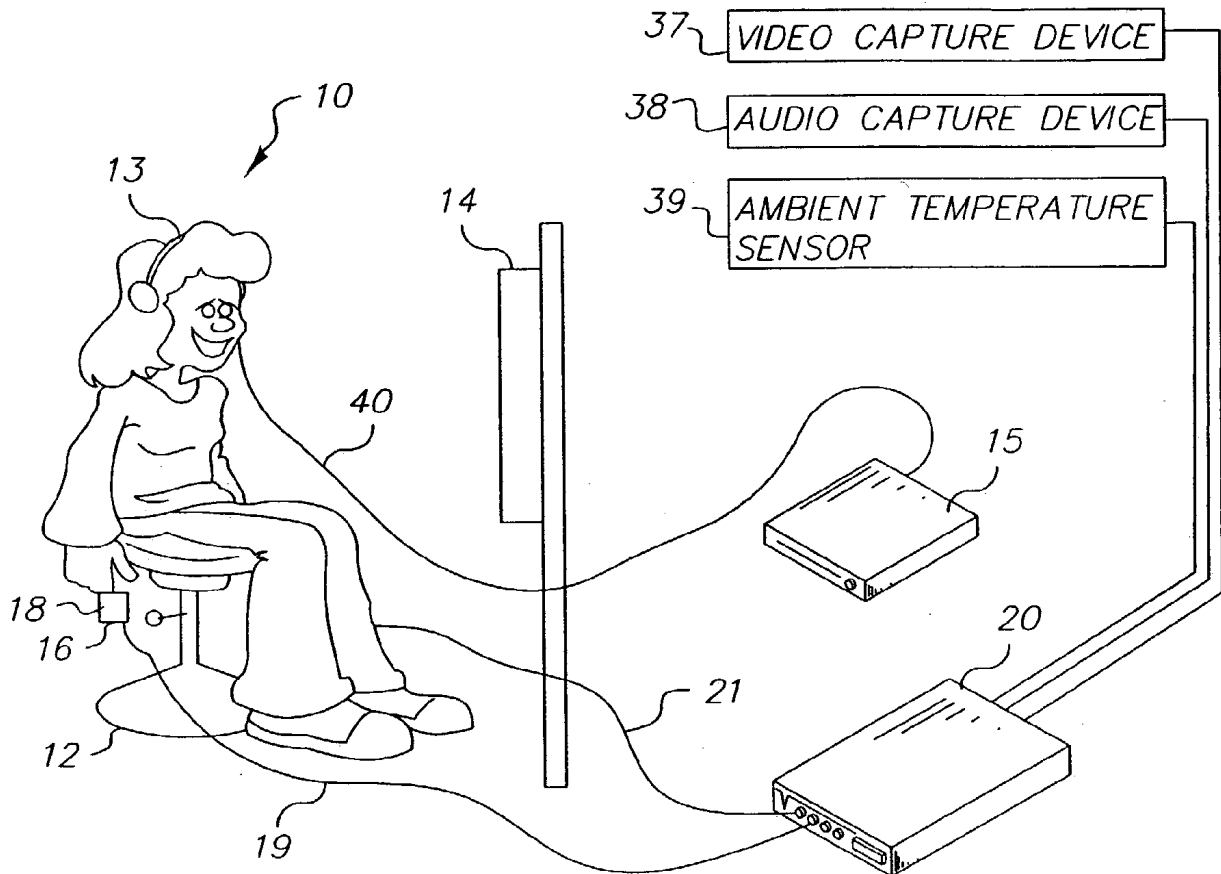
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Related U.S. Application Data

(63) **Continuation-in-part of application No. 09/865,902, filed on May 25, 2001.**



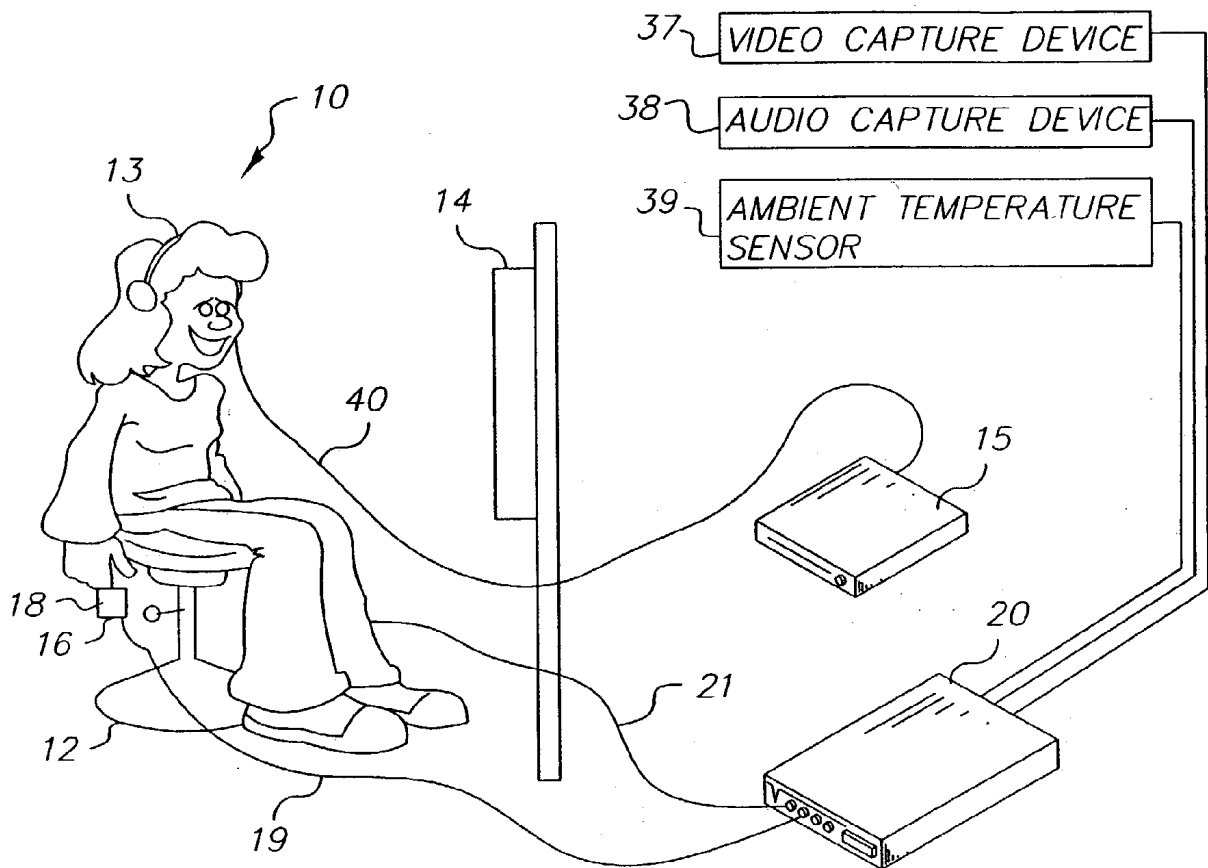


FIG. 1

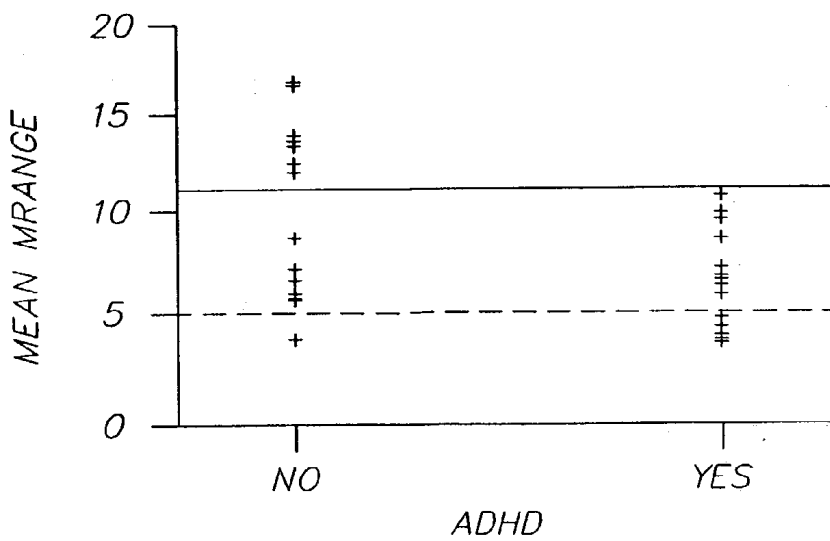


FIG. 5

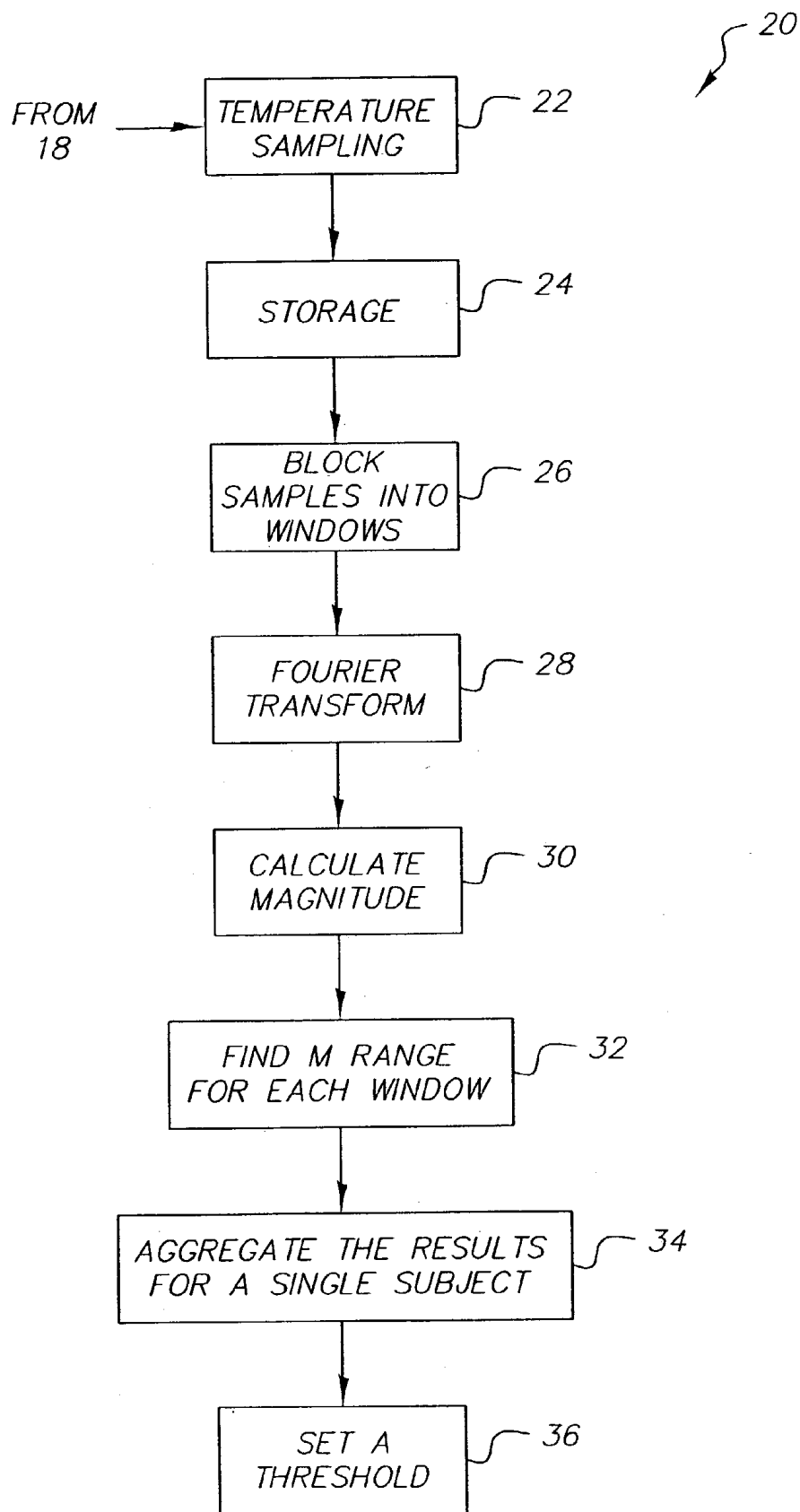


FIG. 2

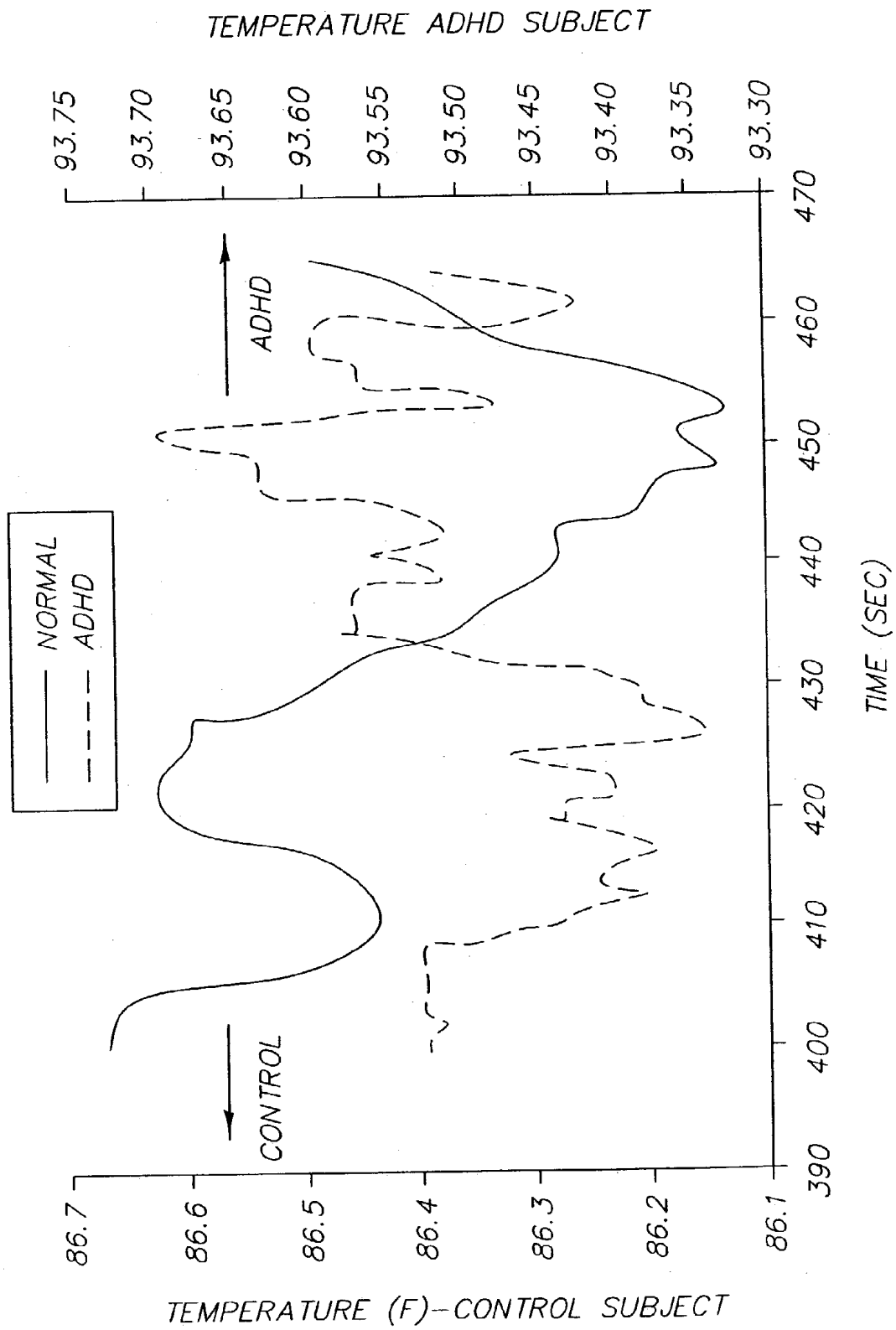


FIG. 3

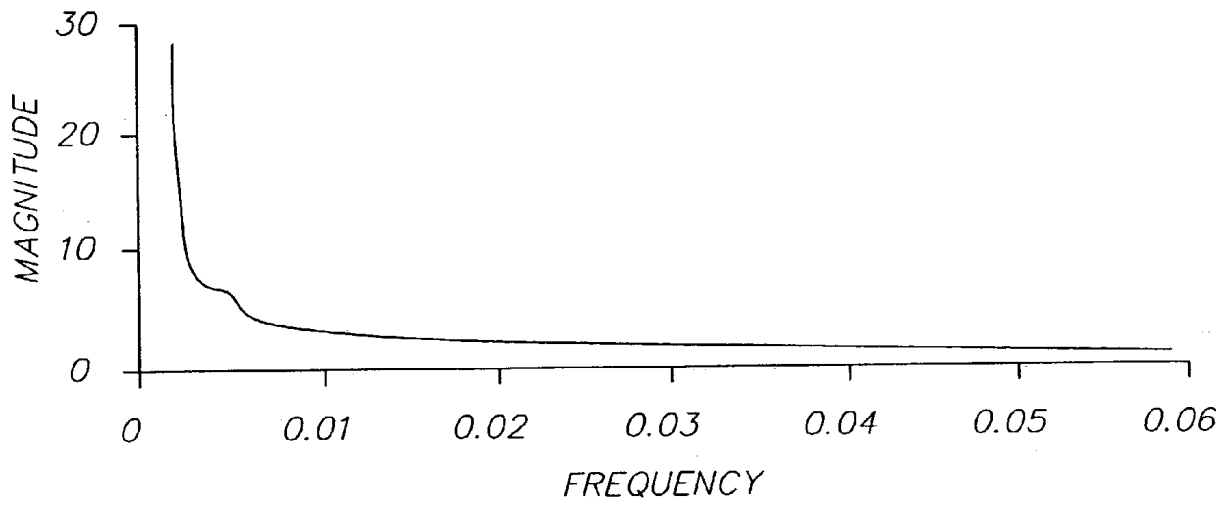


FIG. 4A

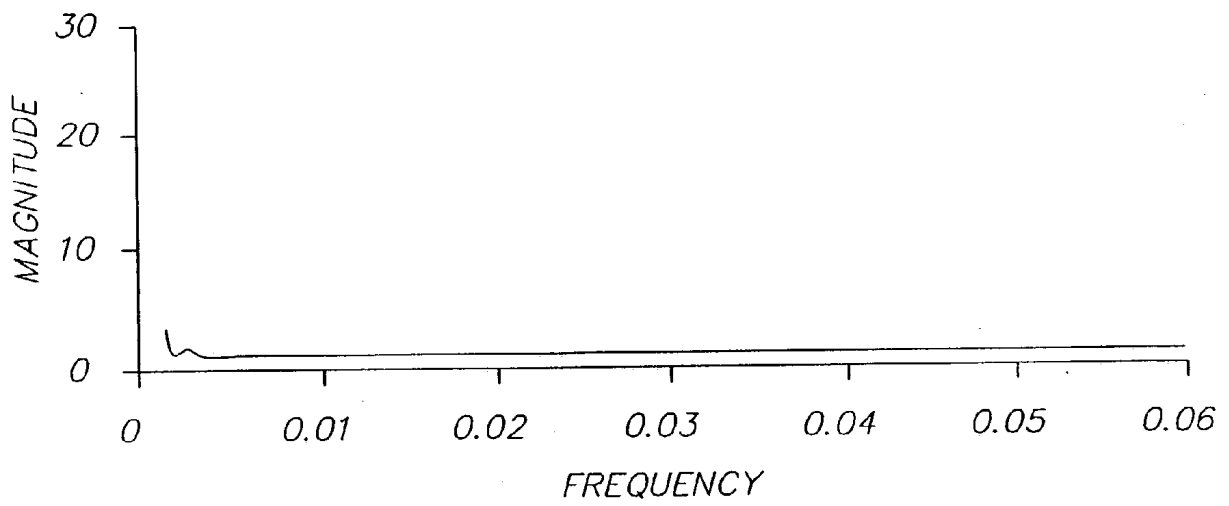


FIG. 4B

TECHNIQUE FOR DIAGNOSING ATTENTION DEFICIT HYPERACTIVITY DISORDER

CROSS REFERENCE TO RELATED APPLICATIONS

[0001] This application is a Continuation-in Part Application of U.S. patent application Ser. No. 09/865,902 filed May 25, 2001.

FIELD OF THE INVENTION

[0002] This invention relates in general to methods for improving the reliability of a technique for diagnosing Attention Deficit Hyperactivity Disorder (ADHD) and more particularly to a technique for measuring an individual's peripheral temperature to determine values indicative of ADHD by recording environmental conditions such as noise during the test procedure.

BACKGROUND OF THE INVENTION

[0003] ADHD is the most common neurobehavioral disorder of childhood as well as among the most prevalent health conditions affecting school-aged children. Between 4% and 12% of school age children (several millions) are affected. \$3 billion is spent annually on behalf of students with ADHD. Moreover, in the general population, 9.2% of males and 2.9% of females are found to have behavior consistent with ADHD. Upwards of 10 million adults may be affected.

[0004] ADHD is a difficult disorder to diagnose. The core symptoms of ADHD in children include inattention, hyperactivity, and impulsivity. ADHD children may experience significant functional problems, such as school difficulties, academic underachievement, poor relationships with family and peers, and low self-esteem. Adults with ADHD often have a history of losing jobs, impulsive actions, substance abuse, and broken marriages. ADHD often goes undiagnosed if not caught at an early age and affects many adults who may not be aware of the condition. ADHD has many look-alike causes (family situations, motivations) and comorbid conditions (depression, anxiety, and learning disabilities) are common.

[0005] Diagnosis of ADHD involves a process of elimination using written and verbal assessment instruments. However, there is no one objective, independently validated test for ADHD. Various objective techniques have been proposed but have not yet attained widespread acceptance. These include:

[0006] 1. The eye problem called convergence insufficiency was found to be three times more common in children with ADHD than in other children by University of California, San Diego researchers.

[0007] 2. Infrared tracking to measure difficult-to-detect movements of children during attention tests combined with functional MRI imaging of the brain were used by psychiatrists at McLean Hospital in Belmont, Massachusetts to diagnose ADHD in a small group of children (*Nature Medicine*. Vol. 6, No. 4, April 2000, Pages 470-473).

[0008] 3. Techniques based on EEG biofeedback for the diagnoses and treatment of ADHD are described by Lubar (*Biofeedback and Self-Regulation*, Vol. 16, No. 3, 1991, Pages 201-225).

[0009] 4. U.S. Pat. No. 6,097,980, issued Aug. 1, 2000, inventor Monastra et al, discloses a quantitative electroencephalographic process assessing ADHD.

[0010] 5. U.S. Pat. No. 5,913,310, issued Jun. 22, 1999, inventor Brown, discloses a video game for the diagnosis and treatment of ADHD.

[0011] 6. U.S. Pat. No. 5,918,603, issued Jul. 6, 1999, inventor Brown, discloses a video game for the diagnosis and treatment of ADHD.

[0012] 7. U.S. Pat. No. 5,940,801, issued Aug. 17, 1999, inventor Brown, discloses a microprocessor such as a video game for the diagnosis and treatment of ADHD.

[0013] 8. U.S. Pat. No. 5,377,100, issued Dec. 27, 1994, inventors Pope et al., discloses a method of using a video game coupled with brain wave detection to treat patients with ADHD.

[0014] 9. Dr. Albert Rizzo of the Integrated Media Systems Center of the University of Southern California has used Virtual Reality techniques for the detection and treatment of ADHD.

[0015] 10. U.S. Pat. No. 6,053,739, inventors Stewart et al., discloses a method of using a visual display, colored visual word targets and colored visual response targets to administer an attention performance test. U.S. Pat. No. 5,377,100, issued Dec. 27, 1994, inventors Patton et al., discloses a system of managing the psychological state of an individual using images. U.S. Pat. No. 6,117,075 Barnea discloses a method of measuring the depth of anesthesia by detecting the suppression of peripheral temperature variability.

[0016] There are several clinical biofeedback and physiologic monitoring systems (e.g. Multi Trace, Bio Integrator). These systems are used by professional clinicians. Although skin temperature spectral characteristics have been shown to indicate stress-related changes of peripheral vasomotor activity in normal subjects, there has been no disclosure of use of variations in skin-temperature response to assist in diagnosing ADHD. (See: *Biofeedback and Self-Regulation*, Vol. 20, No. 4, 1995).

[0017] As discussed above, the primary method for diagnosing ADHD is the use of a bank of written and verbal assessment instruments designed to assess criteria established by the American Medical Association (AMA) as described in the Diagnostic and Statistics manual (DSM-IV) and administered by the school psychologist or other licensed practitioner. In some cases those individuals who meet DSM-IV criteria for ADHD diagnosis are prescribed a drug such as Ritalin. Behavioral observations of the patient while on Ritalin are conducted to assess the impact of prescribed medication.

[0018] There is thus a need for a simple, inexpensive, and reliable technique for assisting in the diagnosis of ADHD.

SUMMARY OF THE INVENTION

[0019] According to the present invention, there is provided a solution to the problems and fulfillment of the needs discussed above.

[0020] According to a feature of the present invention, there is provided a method of determining whether an

individual has Attention Deficit Hyperactivity Disorder, comprising: sampling the peripheral skin temperature of an individual during a pre-determined time interval when the subject is in an inactive state to provide sampled peripheral skin temperature data; dividing the data into a series of windows of equal time intervals across the pre-determined time interval; for each window applying a fast Fourier Transform algorithm (FFT) to produce a data set of FFT magnitudes and determining a range of FFT magnitude variation, which is the difference between the FFT magnitude at a frequency f_{\max} and the FFT magnitude at the minimum frequency f_{\min} ; aggregating the magnitude ranges for one of or less than all of the windows to produce an aggregate magnitude range value which is indicative of whether or not the individual has ADHD; wherein prior to sampling, if one or more of the following noises are present it is eliminated through manual or automatic means: self diversion, external stimulation, technical problems, sleep problems.

ADVANTAGEOUS EFFECT OF THE INVENTION

[0021] The invention has the following advantages.

[0022] 1. A technique for diagnosing ADHD is provided which is simple, inexpensive and reliable.

BRIEF DESCRIPTION OF THE DRAWINGS

[0023] FIG. 1 is a diagrammatic view illustrating an embodiment of the present invention.

[0024] FIG. 2 is a block diagram of a system incorporating the present invention.

[0025] FIGS. 3, 4a and 4b are graphical views useful in explaining the present invention.

[0026] FIG. 5 is a diagram of an example of finding the proper threshold θ to separate ADHD subjects from non-ADHD subjects.

DETAILED DESCRIPTION OF THE INVENTION

[0027] According to the invention, it has been found that a signature of ADHD is hidden in fluctuation of the temperature of the skin as measured at the extremities such as at a fingertip. It is well known in the art that as a person's stress level increases the blood vessels in the body constrict, (as is evidenced by the fact a person's blood pressure increases as their level of stress increases.) As the blood vessels in the body constrict, blood flow is restricted. This is most evident in the extremities such as the fingers, because the blood vessels in the extremities are small and furthest from the heart. A direct result of decreased blood flow to the blood vessels in the extremities is a decrease in the peripheral temperature of the extremities. Conversely, as a person's stress level decreases and one relaxes the blood vessels also relax and dilate causing blood flow to increase. As the blood flow to the vessels in the extremities increases, the peripheral temperature of the extremities increases. When a subject with ADHD is subjected to sensory deprivation such as being made to look at a blank screen or an obscured image, the lack of stimulation increases and their level of anxiety and their stress level increases. As their stress level increases their blood vessels constrict and the peripheral

temperature of their extremities decreases. Biofeedback practitioners have long used measurement of hand temperature to help subjects manage their physiology by controlling blood flow to the extremities. The literature reports that reduced blood flow to the brain is frequently found in patients with ADHD.

[0028] In addition to peripheral skin temperature and peripheral skin temperature variability as indications of stress and therefore ADHD there are other physiologic measures which are known (or potential) indicators of stress and therefore ADHD such as; bilateral temperature variability, heart rate, heart rate variability, muscle tension (excessive and chronic measured via surface electromyography—SEMG), bilateral muscle tension imbalance, galvanic skin response (i.e., electro dermal response—EDR), eye saccades, blood oxygen (SpO_2), salivary IGA, electroencephalography (EEG), peripheral blood flow (measured via photoplethysmography—PPG), and peripheral blood flow variability (PPG).

[0029] As shown in FIG. 1, a subject 10 is sitting on a chair 12 watching a screen 14. The subject is at rest in an inactive state. The subject 10 is shown wearing a set of earphones 13 connected via a wire 40 to a sound generating device 15. The earphones 13 may be used to reduce or eliminate the audio stimulus from the environment during the test. The method described in this embodiment of the present invention places the subject in sensory deprived surroundings. Other examples of providing sensory deprivation are to have the subject wear a pair of translucent glasses, goggles or eye mask not shown. The glasses or goggles block any visual stimulus from the subject 10. A sensor 18 measures the temperature of a fingertip 16 of subject 10. The temperature readings are supplied to module 20 via a wire 19. The temperature sensor for the opposite hand is not shown but is connected via wire 21 to module 20. The temperature can be taken from one hand or both hands and from one or more locations on each hand. Video (video camera) and audio (microphone) capture devices 37 and 38 respectively record test environment and ambient temperature sensor 39 is used to record test environment temperature. This data is fed to module 20.

[0030] As shown in FIG. 2, module 20 includes temperature sampling circuit 22, data storage 24, window blocking 26, Fourier transform 28, Magnitude calculation 30, Mrange calculation 32, aggregation step 34 and Thresholding step 36.

[0031] In FIG. 1, the fingertip temperature is first recorded during an interval when the subject 10 has been asked to sit quietly for a period of about ten minutes. The temperature data is sampled 22 at a time interval Δt creating a list of n temperature samples, which are stored in storage 24.

[0032] Now referring to FIG. 2, in block 26, the n samples are divided into groups of m samples, each group corresponding to a given time window of width Δt (~32-64 sec) equally spaced in time (~50 sec) across the entire data collection time interval Δt . The data from each window is then passed through a Fast 30 Fourier Transform (FFT) algorithm producing 2^{m-1} data points spaced equally in frequency space. The values are complex numbers having form

$$FFT(f_m) = A(f_m) + B(f_m)i$$

[0033] where i is the $\sqrt{-1}$. The Phase $\Phi(f_m)$ is then found from the equation

$$\Phi_l(f_m) = \text{Tan}^{-1}\left(\frac{B(f_m)}{A(f_m)}\right) \quad (1.0)$$

[0034] and the Magnitude $M(f_m)$ from

$$M_l(f_m) = \sqrt{B(f_m)^2 + A(f_m)^2} \quad (1.1)$$

[0035] In the equations 1.0 and 1.1 the subscript l refers to the fact that a separate signal is extracted for each hand so the subscript is l for data extracted from the left-hand data and r for data from the right hand. **FIG. 3** graphically illustrates the temperature signal during one window for a normal subject and a person diagnosed with ADHD.

[0036] **FIGS. 4a** and **4b** graphically illustrate the magnitude transform for the data corresponding with a subject with ADHD and normal subject. The magnitude spectrum undergoes dramatic changes essentially changing from a hyperbolic curve to a flat response.

[0037] Referring again to **FIG. 2**.

[0038] Raw Data

[0039] The raw data $T_{k,l}(t)$ is the temperature taken from hand l at a fingertip **16** as shown in **FIG. 1**, during the 10-minute session. The sessions were taken over a period of weeks. Some subjects had as few as 2 sessions and some as many as 5 sessions. k is used to represent the session.

[0040] Windows

[0041] The data for each session were divided into a series of windows (step **26**) prior to performing the Fourier Transform operation **28**. Call the window width w . In this analysis, the window width was 64 seconds and there were 10 windows spaced at 50-second intervals (the windows overlap) across the 1600 sec baseline spanning the range of 100-500 sec, other values of w can be used. The window number in a session is referred to with the letter j . For each window a FFT algorithm calculates the Fourier Transform $F(f)$. The Magnitude and Phase of this transform are defined as given above.

[0042] In step **32** the range of magnitude variation during a window is calculated using equation (1.2) below where f_{\max} and f_{\min} are the frequencies where the Magnitude is the greatest and the least respectively (note the dc component at frequency zero is excluded).

$$M_{\text{range}} = [M(f_{\max}) - M(f_{\min})] \quad (1.2)$$

[0043] In a further embodiment of this method, other statistics from a Fourier Transform, calculated from the quantities denoted above as $A(f_m)$, $B(f_m)$, $\Phi(f_m)$, and $M(f_m)$ can be used. In addition to using Fourier Transforms, this further embodiment can use statistics derived from a Wavelet transform of the data or other filtering of the data (as in Strang, G. and Nguyen, T. (1996), *Wavelets and Filter Banks*, Wellesley-Cambridge Press, Wellesley, Mass.).

[0044] Aggregation of Samples

[0045] Samples are aggregated in step **34**. There are 10 samples from each hand from each session. The first step is to choose an aggregation statistic, which can be the mean,

median, variance, or other statistic, which is an aggregate of the computed M_{Range} values in each window for each session and each hand. Other statistics that can be used for aggregation include the standard deviation, range, interquartile distance, skewness, kurtosis, Winsorized mean and variance, and robust estimates of mean and variance. Equations below are given for aggregating the mean and the variance. The mean magnitude range for the left hand of session i during session k is found from equation 2.0. where z is the number of windows in the session.

$$\langle M_{k,l} \rangle = \frac{\sum_{j=1}^z [M(f_{\max})_j - M(f_{\min})_j]}{z} \quad (2.0)$$

[0046] And the corresponding variance is:

$$\langle \text{Var}_{k,l} \rangle = \frac{\sum_{j=1}^z \{[M(f_{\max})_{j,l} - M(f_{\min})_{j,l}] - \langle M_{k,l} \rangle\}^2}{z-1} \quad (2.1)$$

[0047] Combining these session means and variances over both hands and all the sessions s that a subject attended gives an aggregated mean μ and aggregated variance var_i .

$$\mu = \frac{\sum_{k=1}^s \sum_{l=1}^2 \langle M_{k,l} \rangle}{2s} \quad (2.2)$$

$$\langle \text{var} \rangle = \frac{\sum_{k=1}^s \sum_{l=1}^2 \text{Var}_{k,l}}{2s} \quad (2.3)$$

[0048] Other embodiments of this aggregation step include using the data from only one hand—either the left hand, the right hand, or the dominant hand (and if the subject is ambidextrous, the dominant hand would be defined as the average of both hands.) In addition, these embodiments may not require averaging of several sessions, but selecting only one session for use or using a weighted combination of each session's results.

[0049] Thus, the totality of these embodiments include methods that involve any and all combinations of: statistics derived from Fourier or Wavelet transformations or other filtering, plus any one of many possible aggregation statistics, plus using data from only one hand or the dominant hand or the average of both hands, plus using either all sessions or a subset of the sessions or a weighted combination of each session's results.

[0050] Diagnostic Indicator A Diagnostic indicator is determined by setting a threshold level θ for the aggregation statistic in step **36**. When the subject's measured aggregate statistic is less than the threshold θ , the test indicates the subject has ADHD. When the subject's measured aggregate statistic is greater than the threshold θ the test indicates the subject does not have ADHD. A single threshold may be

used for all subjects or the threshold may be set differently for different groups such gender or age.

[0051] The method of obtaining the threshold θ is now described. It is similar to a method in the statistical literature called "discriminant analysis". In fact, one could use discriminant analysis c for this data; however this method was devised because it can be enhanced and used for purposes discriminant analysis cannot handle. This enhancement will be described later.

[0052] To find the value of θ that gives the highest percentage of correct diagnoses, a simple example must first be illustrated. In this example, there are 32 aggregation statistics, one for each subject in the study. Next thresholds $\theta=11.5$ and $\theta=5$ were considered. The 32 aggregation statistics are shown in FIG. 5, along with threshold $\theta=11.5$ as the solid line and $\theta=5$ as the dashed line. A different percent of correct diagnosis results when $\theta=11.5$ is used compared to $\theta=5$. Naturally, there are an infinite number of potential values for θ , and a procedure to pick the one that gives the highest percent of correct diagnoses are needed.

[0053] Thus, the following procedure was used: Twenty-five equally spaced values, spanning the range of the 32 aggregation statistics, were found. At each of these 25 values, the percent p of correctly diagnosed subjects was computed. A spline is fitted through this data, so that p is now estimated as a smooth function of θ . Then, the maximum value of this smooth function is found, and θ is set to be where the percent of correct diagnoses is maximized. Since this is often an interpolation, the value of the spine function at θ is not used, but instead is recomputed to percent of correct diagnoses at θ .

[0054] An enhanced method that works in situations where discriminant analysis does not work calls for replacing the percent of correct diagnoses in the above procedure with a weighted percent of false positive and false negative diagnosis, and then to minimize this weighted percent. This method allows the flexibility to choose the relative importance of false positive and false negatives, and to have this used in determining θ . One way to set the relative importance is to use the cost of a false negative diagnosis.

[0055] Virtually every analysis method tried produced correct diagnoses at a rate that is statistically above chance results at the $\alpha=0.05$ level, and many methods produced statistically significant results at the $\alpha=0.01$ level (see Table 1 through Table 8). This indicates that the diagnosis method proposed, using windowed Fourier transforms of hand temperatures, has found a real effect. The diagnoses obtained are significantly better than one would obtain using random chance.

[0056] For example, comparing the case where the variance was used on all data with one threshold for everyone, we see the method produces 68.8% correct diagnoses. If the variance is used with gender thresholds, the percent correct increases to 84.4%. Using different thresholds by gender improves the diagnoses when using all data, see Table 1. This is consistent with statements by Raymond, K. B. (1997). *Dissertation Abstracts International*: Section A: Humanities and Social Sciences, 57 (12-A) 5052, and also Katz, L., Goldstein, G., Geckle, M. (1998). *Journal of Attention Disorders*. 2(4), 239-47, who state that females with ADHD are under-diagnosed. This suggests that a

different standard of diagnosis is necessary for females. Age based thresholds improve the percent correct by 3% (see Table 1). Any of the methods of separating thresholds by gender or age or neither, produce diagnoses that are statistically better than chance.

[0057] Another result shown in tables reveals that removing noises (as described below) produce the highest percent correct diagnosis. This is consistent with the fact, that the data removed was contaminated and less likely to demonstrate the effect of interest. Further, note that without using gender or age thresholds after removing noises, the variance produces correct diagnoses 84.6% of the time. Using gender or age thresholds after removing noises, or using the mean or median, did not improve the results.

[0058] Listed below are the types of noise:

[0059] Self Diversion

[0060] Children divert themselves by moving, using mental exercises or external tools such as gum or suckers

[0061] External Stimulation

[0062] Noises, Room Temperature, Parents in Room etc.

[0063] Technical Problems

[0064] Loose sensors, missing sensors, pauses, computer failures

[0065] Sleep problems

[0066] Child falls asleep during the session

[0067] Medication Problems

[0068] Child's medication is still active during session or child is on long acting drug

[0069] Automatic or manual elimination of temperature data during a noise event can be accomplished by recording ambient temperature, video and audio data during the test. These data are synchronized to the recording of subject temperature data through a time stamp. Automatic or manual analysis techniques can also be used to identify noise event occurrences and either alert the test evaluator of the occurrence for manual elimination of subject temperature data through the time stamp or automatically eliminate suspect temperature data through more advanced techniques. Specifically, video and audio recording can provide historical reference of the test, which can be used by the test evaluator in conjunction with the subject temperature data to eliminate suspect subject temperature data. More sophisticated techniques such as scene change algorithms, and peak noise detection can be used to alert the test evaluator to a noise event to manually eliminate subject temperature data during the noise event. These same techniques in conjunction with the time stamp could also be used to automatically eliminate suspect subject temperature data. Video and audio recording addresses self-diversion, external stimulation and sleep problems. Ambient temperature recording in conjunction with comparison data from the subject temperature data addresses the problems of missing or loose sensors which might cause a loss of temperature data from the subject.

[0070] Other analysis methods were tried and found to be less successful, though these methods were significantly

better than chance. For example, applying a Butterworth filter to the temperature data as suggested by Shusterman, V. and Barnea, O. (1995). *Biofeedback and Self-Regulation*, 20(4), 357-365 did not produce improved results. Nor did separating the data by session (Table 7) or by hand (Table 8). The highest accuracy is obtained by averaging sessions and averaging two hands for tests. The benefit of using both sessions and both hands is that reduction of variability occurs, enabling more reliable diagnoses. A well-known statistical principle is that the variability of the average of multiple sessions or two hands is less than the variability of one session or one hand. Nor did removing the first two time periods (Tables 3, 4 and 6) improve the percent of correct diagnoses.

[0071] The percent of false positive and false negative diagnoses was examined. Using the mean statistic and one threshold for all subjects, a result of 25% false positive diagnoses and 0% false negative diagnoses was achieved. Using separate thresholds by gender and the variance statistic produced a result of 9.4% false positive diagnoses, and 6.3% false negative diagnoses.

[0072] The test method was applied to 50% ADHD subjects and 50% non-ADHD subjects; however, if it was applied only to symptomatics (a subset of the population in

which most have ADHD), it is shown below that the method test actually will produce higher accuracy. The actual rate of false diagnoses depends on the assumed percent of true ADHD subjects in the population of symptomatics to be tested.

[0073] Let p be the proportion of subjects in the study who actually have ADHD. Let f_+ be the proportion of false positive diagnoses of those subjects who do not have ADHD. Let f_- be the proportion of false negative diagnoses of those subjects who do have ADHD. Then the proportion c of correct diagnoses is:

$$c=1-(f_-p-f_+(1-p))$$

[0074] The derivative of c is:

$$\partial c/\partial p=f_+-f_-$$

[0075] The derivative is positive whenever f_+ is greater than f_- . Thus, increasing the value of p will increase the proportion c of correct diagnoses.

[0076] The invention has been described in detail with particular reference to certain preferred embodiments thereof, but it will be understood that variations and modifications can be effected within the spirit and scope of the invention.

APPENDIX

5 The following is a list of tables in the Appendix that show the percent of subjects correctly diagnosed by different analysis methods, or by using different portions of the data, or by a combination of analysis methods and different portions of the data (note that in all the following cases, sessions whose subjects had medication problems have been removed):

0 Table 1: All remaining data used.

 Table 2: Windows with technical problems (sensor falling off or pause button pushed) eliminated.

5 Table 3: First two time windows removed.

 Table 4: Same as Table 2, but first two time windows are removed.

 Table 5: Sessions where there were serious self-diversion problems
10 were removed.

 Table 6: Same as 5, but first two time periods were also removed.

 Table 7: Same as 1, one threshold for all subjects, but data from
15 only session 1, or only session 2 or both sessions were used.

 Table 8: Same as 1, one threshold for all subjects, but data from left hand; or right hand; or dominant hand used.

Table 1 -- Percent of Correct Diagnoses

Subjects with medication problems removed (2822 & 2813 Sesson 1)

Both Hands/Both Sessions, N=32

95% Significance is >65.6% correct, 99% Significance is >71.9% correct

5

Data used: All Data

		Statistic Used:		
		Mean	Median	Variance
		% Correct Diagnoses	% Correct Diagnoses	% Correct Diagnoses
10	Thresholds used			
	One Threshold for Everyone	75.00	68.75	68.75
15	Age Thresholds	78.13	71.88	71.88
20	Gender Thresholds	81.25	68.75	84.38

Table 2 -- Percent of Correct Diagnoses

Subjects with medication problems removed (2822 & 2813 Sesson 1)

25 Both Hands/Both Sessions, N=32

95% Significance is >65.6% correct, 99% Significance is >71.9% correct

Data used: Remove technical problems

		Statistic Used:		
		Mean	Median	Variance
		% Correct Diagnoses	% Correct Diagnoses	% Correct Diagnoses
30	Thresholds used			
	One Threshold for Everyone	68.75	68.75	68.75
35	Age Thresholds	75.00	75.00	75.00
40	Gender Thresholds	78.13	68.75	81.25

Table 3 -- Percent of Correct Diagnoses

Subjects with medication problems removed

Both Hands/Both Sessions, N=32

95% Significance is >65.6% correct, 99% Significance is >71.9% correct

5

Data used: Remove 1st 2 time periods

		Statistic Used:		
		Mean	Median	Variance
		% Correct	% Correct	% Correct
		Diagnoses	Diagnoses	Diagnoses
10	Thresholds used			
	One Threshold for Everyone	68.75	65.63	65.63
	Age Thresholds	71.88	68.75	65.63
20	Gender Thresholds	71.88	65.63	68.75

Table 4 -- Percent of Correct Diagnoses

Subjects with medication problems removed

25 Both Hands/Both Sessions, N=32

95% Significance is >65.6% correct, 99% Significance is >71.9% correct

Data used: Remove technical problems and 1st 2 time periods

		Statistic Used:		
		Mean	Median	Variance
		% Correct	% Correct	% Correct
		Diagnoses	Diagnoses	Diagnoses
30	Thresholds used			
	One Threshold for Everyone	65.63	65.63	68.75
	Age Thresholds	71.88	68.75	68.75
40	Gender Thresholds	68.75	65.63	71.88

Table 5 -- Percent of Correct Diagnoses

Subjects with medication problems removed (2822 & 2813 Session 1)

Both Hands/Both Sessions, N=26

95% Significance is >65.4% correct, 99% Significance is >73.1% correct

5

Data used: Remove tech/external/self-diverted problems

		Statistic Used:		
		Mean	Median	Variance
		% Correct Diagnoses	% Correct Diagnoses	% Correct Diagnoses
10	Thresholds used			
	One Threshold for Everyone	76.92	73.08	84.62
15	Age Thresholds	84.62	76.92	84.62
20	Gender Thresholds	76.92	76.92	84.62

Table 6 -- Percent of Correct Diagnoses

Subjects with medication problems removed (2822 & 2813 Session 1)

25 Both Hands/Both Sessions, N=32

95% Significance is >65.4% correct, 99% Significance is >73.1% correct

Data used: Remove tech/external/self-diverted problems and
1st 2 time pds

		Statistic Used:		
		Mean	Median	Variance
		% Correct Diagnoses	% Correct Diagnoses	% Correct Diagnoses
30	Thresholds used			
	One Threshold for Everyone	73.08	65.38	73.08
35	Age Thresholds	80.77	76.92	76.92
40	Gender Thresholds	69.23	73.08	76.92

Table 7 -- Percent of Correct Diagnoses by Session

Subjects with medication problems removed (2822 & 2813 Session 1)

5 Data used: All Data

		Statistic Used:		
		Mean	Median	Variance
		% Correct	% Correct	% Correct
		Diagnoses	Diagnoses	Diagnoses
10	Session Used			
15	Session 1	68.75	68.75	71.88
	Session 2	71.88	65.63	68.75
	Both Sessions	75.00	68.75	68.75

20

Table 8 -- Percent of Correct Diagnoses by Hand

Subjects with medication problems removed (2822 & 2813 Session 1)

25 Data used: All Data

		Statistic Used:		
		Mean	Median	Variance
		% Correct	% Correct	% Correct
		Diagnoses	Diagnoses	Diagnoses
30	Hand Used			
35	Both Hands	75.00	68.75	68.75
	Dominant Hand	75.00	65.63	65.63
40	Left Hand	65.63	62.50	71.88
	Right Hand	65.63	68.75	68.75

PARTS LIST

- 10 subject
- 12 chair
- 13 earphones
- 14 screen
- 15 sound generating device
- 16 fingertip
- 18 sensor
- 19 wire
- 20 module
- 21 wire
- 22 temperature sampling circuit
- 24 data storage
- 26 window blocking
- 28 Fourier transform
- 30 Magnitude calculation
- 32 Mrange calculation
- 36 thresholding step
- 37 video capture device
- 38 audio capture device
- 39 ambient temperature sensor
- 40 wire

What is claimed is:

1. A method of determining whether an individual has Attention Deficit Hyperactivity Disorder, comprising:

sampling the peripheral skin temperature of an individual during a pre-determined time interval when the subject is in an inactive state to provide sampled peripheral skin temperature data;

dividing the data into a series of windows of equal time intervals across the pre-determined time interval;

for each window applying a fast Fourier Transform algorithm (FFT) to produce a data set of FFT magnitudes and determining a range of FFT magnitude variation, which is the difference between the FFT magnitude at a frequency f_{max} and the FFT magnitude at the minimum frequency;

aggregating the magnitude ranges for one of or less than all of the windows to produce an aggregate magnitude range value which is indicative of whether or not the individual has ADHD;

wherein prior to sampling, if one or more of the following noises are present it is eliminated through manual or automatic means:

self diversion, external stimulation, technical problems, sleep problems.

2. The method of claim 1 wherein said aggregate magnitude range value is the mean of said magnitude range values of all of said windows.

3. The method of claim 1 wherein said aggregate magnitude range value is the median of said magnitude range values of all of said windows.

4. The method of claim 1 wherein said aggregate magnitude range value is the median of said magnitude range values of all of said windows.

5. The method of claim 1 wherein said aggregate magnitude range value is determined by an aggregation statistical algorithm of said magnitude range values of all of said windows.

6. The method of claim 1 wherein said sampling includes sampling the peripheral skin temperature of said individual

during two or more pre-determined time intervals, wherein said aggregate magnitude range value is aggregated for one or more of said pre-determined time intervals.

7. The method of claim 1 wherein said sampling includes sampling the peripheral skin temperature of one or both hands of said individual, and wherein said aggregate for said one or said both hands.

8. The method of claim 1 wherein said the magnitude range of one or more windows at the beginning of said pre-determined time interval is not used to produce said aggregate range value.

9. The method of claim 1 where environmental data including subject motion, audible environment and ambient temperature data are recorded.

10. The method in claim 9 where said environmental data is synchronized to the subject temperature data.

11. The method of claim 9 wherein environmental data are manually analyzed by a test evaluator for the occurrence of one or more noise event occurrences.

12. The method of claim 11 wherein suspect subject temperature data is eliminated from the sampled data during a noise event occurrence.

13. The method of claim 9 wherein said environment data is automatically evaluated using scene change algorithms and peak noise detection to determine if a noise event occurred during the sampling.

14. The method of claim 13 wherein automatic analysis techniques are used to alert the test evaluator of a noise event.

15. The method of claim 14 where the test evaluator eliminates suspect subject temperature data if he/she is alerted to a noise event.

16. The method in claim 13 wherein noise occurrence environment data is used to automatically remove specific temperature data during a noise occurrence during the sampling.

17. The method of claim 9 wherein said environment data is automatically evaluated using a loss of temperature signal or abrupt change in temperature signal from the subject.

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专利名称(译)	诊断注意缺陷多动障碍的技术		
公开(公告)号	US20030120172A1	公开(公告)日	2003-06-26
申请号	US10/301072	申请日	2002-11-21
[标]申请(专利权)人(译)	伊斯曼柯达公司		
申请(专利权)人(译)	伊士曼柯达公司		
当前申请(专利权)人(译)	MCLEAN HOSPITAL CORPORATION, THE		
[标]发明人	FOUST GREGORY B PATTON DAVID L BLAZEY RICHARD N MILLER PAIGE		
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IPC分类号	A61B5/01 A61B5/00 A61B5/16 A61B10/00		
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其他公开文献	US6843774		
外部链接	Espacenet USPTO		

摘要(译)

一种确定个体是否具有注意力缺陷多动障碍的方法，包括：在受试者处于非活动状态的预定时间间隔内对个体的外周皮肤温度进行采样，以提供采样的外周皮肤温度数据；在预定的时间间隔内将数据分成一系列相等时间间隔的窗口；对于每个窗口应用快速傅里叶变换算法（FFT）以产生FFT幅度的数据集并确定FFT幅度变化的范围，其是频率 f_{max} 处的FFT幅度与最小频率处的FFT幅度之间的差值；聚集一个或少于所有窗口的幅度范围，以产生总量值范围值，该值表示个体是否患有ADHD；其中，在采样之前，如果存在以下一种或多种噪声，则通过手动或自动方式消除：自我转移，外部刺激，技术问题，睡眠问题。

