

Detrended Fluctuation Analysis of Acupuncture based on Neural Electrical Signals

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Abstract—Manual acupuncture (MA) is a traditional Chinese Medicine clinical therapy. Although its effectiveness has been proved by clinical trials over thousands of years, the effect and mechanism of acupuncture on the nervous system is not yet clear. Brain is senior central nervous system, and acupuncture as an external stimulation, can induce various electrical activities in the brain cortex. In this paper, we design experiment that acupuncture at acupoint Zusanli with four different frequencies, and collect neural signals on the brain. Detrended fluctuation analysis method is used to extract features from EEG signals evoked by different MA manipulations, which may provide theoretical basis for quantifying acupuncture and revealing the mechanism of acupuncture from the aspect of neural electrical signals.

Keywords-Acupuncture; EEG; Detrended fluctuation analysis

I. INTRODUCTION

Acupuncture is a unique non-drug treatment method in medical field in China. Two thousand years of clinical practice has verified the effectiveness of acupuncture [1-3]. Organism is taken into consideration as a whole, and acupuncture not only has a significant effect in the treatment of disease, but also can improve people's physical conditions. However, regulation mechanism of acupuncture is still not clear, which has seriously restricted the development of acupuncture in medical field.

When human is exposed on external stimuli, the relevant regions of the brain will appear corresponding discharge phenomenon. Recent clinical and experimental data show that acupuncture can induce specific patterns of brain activity, but the specific effects of acupuncture on the brain is still unknown. Therefore, acquire related EEG signals from the corresponding brain regions, extract features from EEG, so as to obtain the effect of acupuncture on the brain neural electrical activities, which provides certain theoretical basis for reveal the mechanism of acupuncture, and accelerates the process of modernization of traditional acupuncture in the medical field.

Zusanli is a widely used acupoint, which is located under the knee along the stomach meridian. A large number of ancient and modern clinical practices have proved that it is able to regulate and control a variety of diseases and physical

fitness. According to the human anatomy structure of nervous system, when acupuncture at acupoint Zusanli, spinal dorsal root and dorsal horn will generate corresponding neural discharge, and neural information will be uploaded to the brain. Therefore, acupuncture has a certain impact on the brain, while brain will send control instructions based on the acupuncture and focus.

It has been shown that EEG signals emerge long range correlation features of shock dynamics [4]. Detrended fluctuation analysis (DFA) is proposed based on deoxyribonucleic acid (DNA) mechanism, which can describe the long range power-law correlation of non-stationary time series with the scaling exponent [5]. It can identify underlying self-similarity from non-stationary time series, and avoid detecting pseudo correlation caused by noise and external trends. In recent years, it has been applied in life science, geology, meteorology and economic fields, which becomes an effective tool for non-stationary time series. DFA also provides a quantitative description for feature extraction and pattern recognition of anesthesia, sleep and seizure by analyzing EEG signals [6-9].

In this paper, DFA method is used to analyze the long-range exponential correlation of EEG signals evoked by acupuncture from the aspect of different brain rhythms. Through statistical analysis, extract the corresponding feature parameters, analysis the effect of acupuncture on the nervous system.

The paper is organized as follows. In section 2, the experimental process is interpreted. In section 3, the analysis method is introduced. In section 4, data analysis results are presented. Finally, conclusions are given.

II. EXPERIMENTS

Subjects are 9 healthy volunteers (6 males and 3 females, aged between 23 and 27). During the experiments, environment should be quiet and dark. Subjects keep eyes closed in the whole experimental process, but stay awake. When there is no apparent EMG disturbance, physicians start needling. For each subject, the process lasts about 90 minutes. At first, let the subject relax until EEG signal becomes stable; Next, acupuncture at acupoint Zusanli for 2 minutes; then retain needle without manipulation for 10

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minutes break. Repeat this sub-process four times. All acupuncture manipulations are twirling reinforcing manipulation. Each sub-process uses acupuncture manipulation with different frequency, such as 50 times/min, 100 times/minute, 150 times/minute and 200 times/minute. The order of acupuncture manipulations is random, so as to eliminate the effect of the manipulation order. The acupuncture experimental process is shown in Fig. 1.

EEG signals are recorded continuously during the whole experimental process from 20 electrodes according to 10-20 system (Fp1(1), Fp1(2), F7(3), F3(4), Fz(5), F4(6), F8(7), T3(8), C3(9), Cz(10), C4(11), T4(12), T5(13), P3(14), Pz(15), P4(16), T6(17), O1(18), Oz(19) and O2(20)). Sampling frequency is 256Hz. Hardware filter passband is 0.5Hz-100Hz. For each subject, we intercept 6 epochs of data, which corresponds to six states, i.e. before acupuncture (BA), acupuncture with 50 times/min (A50), acupuncture with 100 times/min (A100), acupuncture with 150 times/min (A150), acupuncture with 200 times/min (A200) and after acupuncture (AA). Each epoch last 80s.

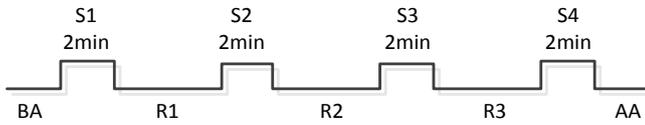


Fig. 1. Acupuncture experimental process. (BA - Before acupuncture; AA - AfterAcupuncture; S1, S2, S3, S4 - different acupuncture manipulations; R1, R2, R3- Retaining needle between different acupuncture manipulations)

III. METHODS

A. Detrended fluctuation analysis method, DFA

Detrended fluctuation analysis method (DFA) is proposed based on stochastic process theory and chaos dynamics, which can describe the long-range correlation of non-stationary time series[5]. DFA algorithm is as follows [10-11].

Step 1: Calculate the cumulative deviation of the original sequence $\{x(t) | t = 1, 2, \dots, N\}$. Then, $x(t)$ is converted to a new sequence $\{y(t) | t = 1, 2, \dots, N\}$

$$y(t) = \sum_{i=1}^t (x(i) - \bar{x}) \quad (1)$$

where \bar{x} is the average value of $x(t)$.

Step 2: $\{y(t) | t = 1, 2, \dots, N\}$ is divided into m disjoint interval. s is the interval length, which presents the time scale. $m = N/s$ is interval number. Fit local trend function $y_s(t)$ of each subinterval with least squares method. According to the the degree of polynomial fitting, it can be written as first-order detrended fluctuation analysis (DFA1), second-order detrended fluctuation analysis (DFA2) etc.

Step 3: Eliminate local trend of sequence $y(t)$ at each interval, and calculate the root mean square of the new sequence, i.e. fluctuation function

$$F(s) = \sqrt{\frac{1}{N} \sum_{i=1}^N (y(t) - y_s(t))^2} \quad (2)$$

Step 4: For different interval length s , repeat Step 3. Then, the relationship between s and $F(s)$ is obtained.

$$F(s) \propto s^\alpha \quad (3)$$

Usually, $F(s)$ will increase with the increase of s . The linear relationship in double logarithmic diagram of $F(s)$ and s represents long range correlation phenomenon. Scaling exponent α is the slope of linear fitting by using least square method, which represents long-range correlation property of time series and is independent of signal amplitude and external trends. If $0 < \alpha < 0.5$, sequence shows inverse correlation in the power-law form. If $\alpha = 0.5$, sequence is completely random and has no long-range correlation phenomenon. If $0.5 < \alpha < 1$, there is continuous correlation in the sequence. If $\alpha = 1$, the sequence is 1/f noise. If $\alpha > 1$, there isn't power-law formed correlation in the sequence. If $\alpha = 1.5$, the sequence fluctuation is Brown motion, which shows long-range correlation. If $\alpha > 1.5$, the sequence shows periodic fluctuation.

B. Normalization method

In order to eliminate individual differences and improve the comparison of the results, we define changing rate as normalization index.

$$R_\alpha = \frac{\alpha}{\alpha_0} \quad (4)$$

where α_0 is the scaling exponent calculated from EEG before acupuncture, which is used as reference value. α is the scaling exponent of EEG under different acupuncture epochs. R_α reflects the relative variance compared with α_0 . If $R_\alpha > 1$, scaling exponent of current epoch increase compared with the epoch before acupuncture, vice versa.

C. Statistical analysis

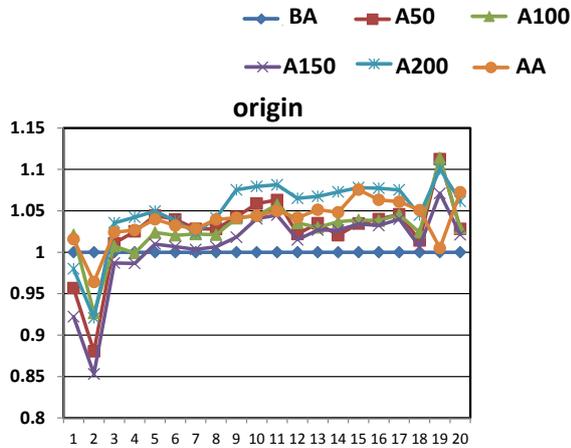
One-way analysis of variance (ANOVA) is introduced to identify the difference between epoch before acupuncture and other epochs, respectively. In this paper, $P < 0.05$ indicates that there is difference between different epochs, while there is significant difference when $P < 0.01$.

IV. RESULTS

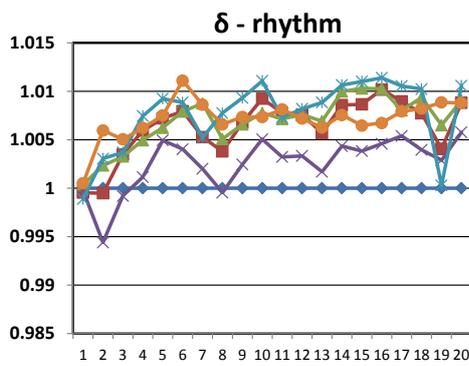
DFA method is used to study the long-range correlation from EEG signals under different acupuncture epochs. For a certain epoch, we adopt sliding window technique with 4s window length and 3s window overlap to calculate α of each channel. For each window, we get a α . Hence, as the window sliding, there is a sequence of α for each channel,

then the average of sequence α is scaling exponent of corresponding channel.

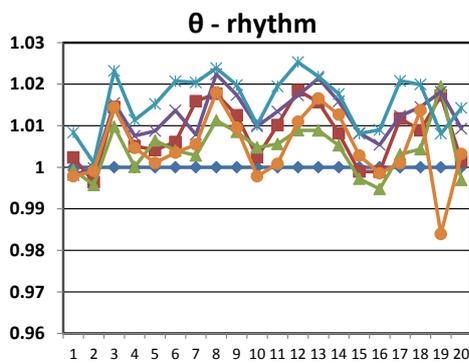
Fig. 2 shows relative scaling exponent of each channel under different acupuncture epochs at different EEG rhythms. The abscissa represents EEG channel number, and the vertical represents R_α . Different line type means different epochs. Fig. 2(a) is scaling exponent of EEG signals at 0.5-47Hz, and Fig. 2(b)-(e) are scaling exponent at different EEG rhythm. It is found that compared with BA epoch, R_α increases at delta and theta rhythms for most brain areas under other five epochs, while decreases at alpha and beta rhythms for most brain areas.



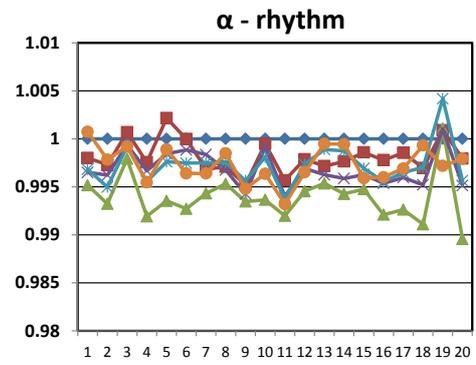
(a)



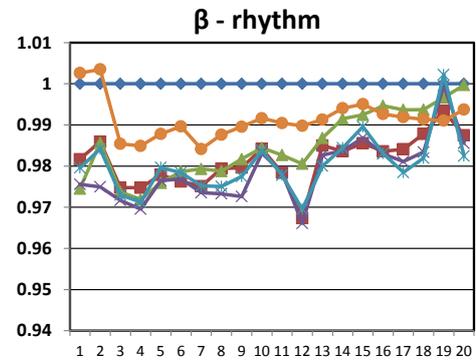
(b)



(c)



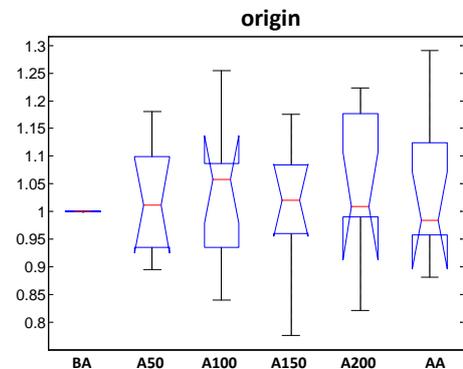
(d)



(e)

Fig.2. Normalized scaling exponent of 20 EEG channels under different acupuncture epochs at different EEG rhythm for nine subjects.

Fig.3 is statistical analysis results of normalized scaling exponent of global brain domain under different acupuncture epochs at different EEG rhythm for nine subjects. It can be seen that (1) R_α during A50, A200 and AA epochs is much higher than that during BA epoch at the delta rhythm; (2) R_α during A100 and A200 epochs increase compared with BA epoch at the theta rhythm; (3) R_α during A100 is much lower than that during BA epoch at the alpha rhythm; (4) R_α during A50, A150 and A200 epochs decrease compared with BA epoch at the beta rhythm.



(a)

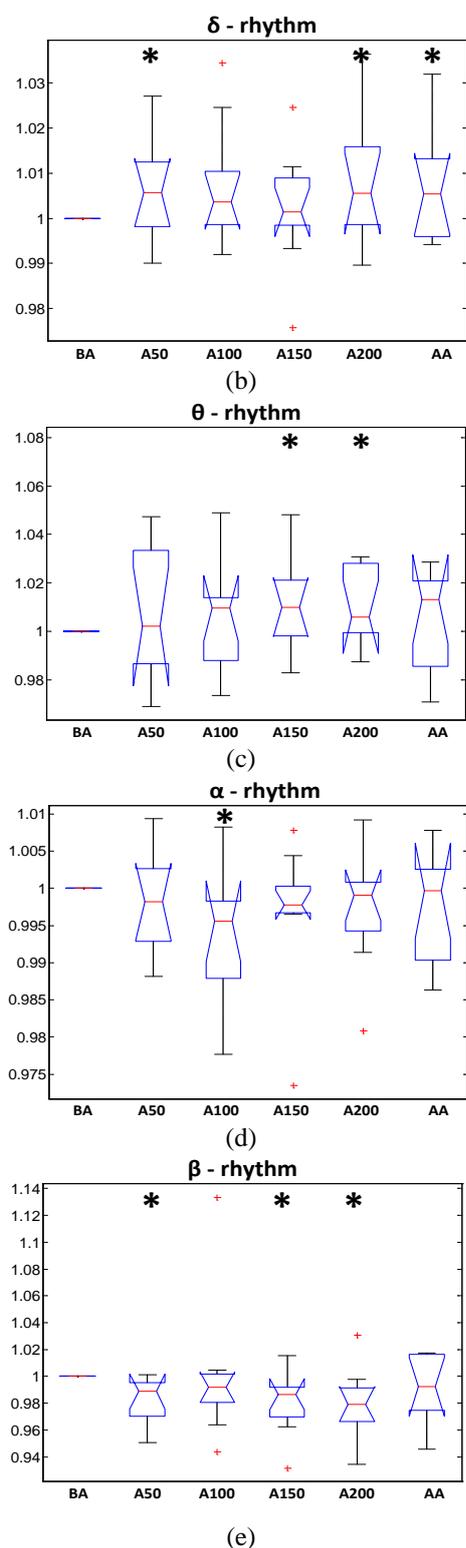


Fig. 3.Boxplot for normalized scaling exponent of averaged global brain domain under different acupuncture epochs at different EEG rhythm for nine subjects. (The star presents that there is different between current epoch and BA epoch at the level of 0.05.)

V. CONCLUSIONS

We use DFA method to study the EEG signals evoked by different acupuncture manipulations at acupoint Zusanli at different EEG rhythms from 9 subjects. It is found that a certain acupuncture manipulation has different effects on different EEG rhythms, and the response of a certain EEG rhythm to different acupuncture manipulation is different. Moreover, we found that acupuncture induce the increase of scaling exponent at the delta and theta rhythm, while decrease at alpha and beta rhythm.

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