Heart Rate Variability; An Index for Monitoring and Analyzing Human Autonomic Activities

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Abstract. Heart rate variability (HRV), a non-invasive index of autonomic controls of the heart, has extensively been employed in various fields of study. The application range of this index persists to expand over the research horizon, and the need continues to demand technical complements for hitherto known deficiencies. Concurrent technical advances in recent years have yielded methodological techniques capable of performing quantitative analysis of HRV. However, as various approaches designed for HRV analysis exist, quantitative comparisons of the investigations have become difficult. Current physiological interpretations of the components of HRV are also inconsistent. Moreover, the issue of respiratory control has been much disputed when determining the HRV. In an attempt to clarify the current controversial state in HRV measurement, the present paper, based on the physiological background of HRV, pursued various procedures of HRV measurement and furnished global explanations for special features characterizing the various analytical approaches adopted thereof.


Keywords: heart rate variability, autonomic activity, respiration

Introduction

Since heart rate is the simplest physiological index, it is extensively employed for studying work-load on the living system. Heart rate on a per min basis is generally represented. In a more detailed perspective, heart rate is analyzed on a beat by beat basis (Berntson et al., 1993). In this case, the time interval of beat by beat may not be consistent and manifests a variation range of 10–30% even if the heart rate per min remains constant (Fig. 1). This phenomenon is known as heart rate variability (HRV). In addition, as heart rate is counted to practically the number of R-wave which is the most prominent waveform of the electrocardiogram, this parameter is sometimes known as the R-R interval variation. Compared to analysis of mean heart rate, information on the state of autonomic nervous system can be retrieved in greater detail.

HRV measurement is currently employed extensively in life sciences, and the application range has recently expanded to include research studies involving physiological anthropology, home economics, architecture and the other scientific fields. Recent years have seen humans getting bored with quantitative increases in objects or environment surrounding us, and the demand for quality instead of quantity has surged. In response to this demanding trend of humans, industrial sectors have recently tuned to “comfort”, and conduct product research and development with the human need as the focus.

The present paper briefly discusses the physiological background and measurement procedures of HRV for application in physiological anthropology.

HRV Measurement

Measurement of this index includes:
1. Determination of electrocardiograph (ECG)
2. Digitization of ECG
3. Detection of the R-wave peak followed by derivation of the R-R interval sequence
4. Data interpolation for securing equal interval
5. Conversion to heart rate from the reciprocal of R-R interval
Items 4 and 5 are not compulsory.

The basic devices required for HRV measurement include a bio-amplifier for ECG measurement, an analog to digital (AD) converter and a computer for R-wave analysis and data interpretation.

Electrocardiogram (ECG)

As analyses of ECG waveforms were not the object in this HRV evaluation, practically no repercussions from the results of leading approach of ECG were encountered. For procurement of an enhanced R-wave, standard lead II
Heart Rate Variability

was most frequently connected on the whole. Smaller time constant was purposely appropriate to minimize artifacts from fluctuation of the baseline.

A-D conversion

It was sufficient to set the sampling frequency as 1 kHz in AD conversion of ECG waveforms. Through experience, we could detect the R wave even at 250 Hz. However, when the sampling frequency was further reduced to a level of 100 Hz, detection of the R-wave would then be difficult.

Detection of R-wave

To discriminate the R-wave from other components (especially the T-wave), due consideration on its special features, such as highly amplified amplitude within the shortest possible time, was taken. Supposing a certain value has been designated for the ECG waveform, and \( t_1 \) and \( t_2 \) are defined as the times when any ECG waveform crosses the fixed value from lower and upper, the apex of R-wave is then expressed as \( \frac{t_1 + t_2}{2} \) (Fig. 2). This algorithm assumed the peak of waveform to lie on the center of the R-wave with left-right bilateral symmetry prevailing. As the \( t_1 \)-\( t_2 \) interval was short, and even if artifacts resulting from waveform symmetrical displacements were produced, errors might be within the permitted range. If this is bothersome, one could adjust within the \( t_1 \)-\( t_2 \) range until a point where the change from positive to negative gradient is observed. This detection algorithm is the most simple one, and if other restrictions are added, the consistency of detection will be further improved. Although real time processing can be done with this algorithm, construction of a noise-tolerant R-wave detection system can be secured, if the real time processing are abandoned. As the R-wave amplitude is subjected to changes with the respiratory movement (Rompleman, 1987), in the R-wave amplitude-dependent detection algorithm, the source of errors is most likely to be due to changes in this R-wave amplitude rather than EMG interference. QRS detection algorithm of higher accuracy established in recent years was also introduced in Sugawara et al. (1998).

Time domain analysis

An approach in the 1960's, basically a conventional analytical approach that preceded the use of methods involving frequency analysis in this field of study, has been employed in the time domain analysis of HRV. However, the extensive use of this conventional approach, especially in the field of clinical applications, is still validated through persistent improvements. Numerous indexes have been proposed to serve as a quantitative determinant in the time domain of HRV. Eight disparate parameters were introduced by Luczak and Laurig (1973) and 26 indexes were proposed by Opmeer (1973). In 1996, a task force set up by the European Society of Cardiology and North American Society of Pacing and Electrocardiology recommended six statistical and four geometrical indexes. Strong correlation between these indexes were the issue then. In their report, the following 4 parameters are particularly selected on a basis that the above issues are made irrelevant:

1. SDNN (Standard deviation of all NN intervals)
2. HRV triangular index (Total number of all NN intervals divided by the height of the histogram of all NN intervals measured on a discrete scale with bins of 7.8125 ms (1/128 seconds))

Fig. 1 Tachogram of heart rate variability. Periodic variations of heart rate are shown.

Fig. 2 Illustration of R-wave detection algorithm. \( t_1 \) and \( t_2 \) are the time when ECG waveform crosses a threshold value (T) from lower to upper and from upper to lower, respectively. Then an apex of the R-wave is calculated as \( \frac{t_1 + t_2}{2} \).
(3) SDANN (Standard deviation of the averages of NN intervals in all 5-minute segments of the entire recording)
(4) RMSSD (The square root of the mean of the sum of the squares of differences between adjacent NN intervals)

Note that SDNN and HRV triangular index determine the global HRV; while RMSSD and SDANN represent the short and long cyclical variability, respectively.

Other than these, coefficient of variance of R-R interval (CV-RR) has been proposed as an index for the time domain analysis. This index, originally introduced by Kageyama in 1980s, represents the standard deviations (SD) of heart beat intervals divided by the mean heart beat interval value and multiplying by a factor, 100. CV-RR basically reflects an activity of the parasympathetic nervous system, although weak responses of the sympathetic nervous system are incorporated as well.

To date, CV-RR is still extensively applied in studies related to clinical medicine and ergonomics (e.g. Hayano, 1988; Sakakibara, 1992; Hasebe 1995; Hashimoto et al., 1996).

However, frequent may be the application of this index in Japan, CV-RR is seldom employed in studies conducted by scientists outside Japan. Suzuki (1992) criticized the experimental findings based on the use of CV-RR by Kageyama et al. (1983) and raised doubts on the meaning of dividing the SD by the mean value.

Furthermore, Luczak (1979) disapproved of the trend stirred up by the proposal of indexes one after another, as quantitative comparisons of various findings yielded according to different indexes were practically unproductive then.

Frequency domain analysis

In a similar manner, numerous methods in analysis of HRV have been proposed. However, HRV is composed of complex frequency-related components, which have been known to elicit different functional aspects of the ANS. As such, HRV analyses involving procedures appropriate for interpreting frequency-related data are definitely convenient and effective.

In frequency analysis, several methods are available. The two procedures most commonly employed for HRV analysis are the fast Fourier transform (FFT) and maximum entropy method (MEM). As many publications furnish detailed explanations and interpretations on the theory and calculation approach of these methods (Akaike, 1970; Burg, 1972; Cerutti et al., 1984), our present review explores and exploits the viewpoints related to special features of the analytical methods and their applications in HRV analysis.

MEM displays characteristic features that effectively facilitate HRV analysis, and its application in biosignals has been documented in many reports related to HRV and electroencephalographic studies in recent years.

The advantages of MEM over FFT may be summarised as follows:

1. higher spectral resolution,
2. consistent spectra can be established with limited data,
3. number of data is not restricted to \( n^2 \),
4. continuous spectral values can be established.

The two problems conceivable at the same time are:

1. the model order has to be determined by a certain method, and
2. linearity values on spectra appropriated for steep peaks can not be guaranteed.

On the aspect of deciding the order described as the first problem, use of either the final prediction error (FPE) or Akaike information criterion (AIC) would solve the issue (Note 1). AIC or FPE can be derived by the following expression:

\[
\text{AIC (m)} = N \log 2_p + N \log \sigma^2 + N + 2 (m + 1)
\]

\[
\text{FPE (m)} = \left( 1 + \frac{m + 1}{N} \right) \left( 1 - \frac{m + 1}{N} \right)^{-1} \text{Em}^2
\]

where,

\[
\text{Em}^2 = \frac{1}{N} \sum_{k=m+1}^{N} \left( y_k - \sum_{i=1}^{m} a_i y_{k-i} \right)^2
\]

and \( N \) represents the quantity of data.

The most appropriate model order then is the minimum value of the values derived thus. On an elaborative basis, AIC or FPE is derived in an ascending order, starting from the small order, and the minimum value is then derived. In doing this, the maximum value required to be computed based on these standards then becomes the issue. In general, while the order of HRV shows 3-10th under normal resting state, computations up to dimensions of 20–25th will suffice. In the case where respiration is controlled at a constant rhythm or where high heart rate is induced by physical work-load, eventuate a order of 20th or over (Note 2).

The beat to beat interval sequence is non-equivalent interval data. By resorting to frequency analysis, an equivalent interval data can most likely be established by interpolation of the issue of data points when FFT is used. In the case of using MEM, as consistent spectra can be established even with limited data points, data can thus be interpreted without any additional interpolate procedures. This is, in fact, one of the advantages of using MEM in HRV analyses.

As for the issue of linear structural properties of the peak value, MEM spectra reflect well the alterations of frequency components, eliciting therefore no practical problems when this approach is appropriated for HRV analysis.

Although this MEM has promptly been suggested by
Sayers (1973) for exploiting possible application in analysis of R-R interval variability, its use in the analysis of R-R interval variability was not realized until after 1980. Since then, numerous findings have been documented with application of the said method (e.g., Bartolli et al., 1985; Baselli et al., 1985; Pagani et al., 1986; Kitney, 1987).

Note 1: The most appropriate orders derived by either AIC or FPE converges in most cases.
Note 2: The order derived are cases where raw data of heart rate interval series devoid of any interpolations are employed. Data with interpolations indicate higher orders.

Estimation of Autonomic Activities

Two components in HRV

Two components with different frequencies prevail in HRV: the respiratory arrhythmia (RSA) that shares the respiratory frequency and the variability related to blood pressure fluctuation characterized by a frequency approximating 0.1 Hz (Fig. 3). The blood-pressure-related component is known as the Mayer-wave-related sinus arrhythmia (MWSA) or 0.1 Hz component attributed to the relevant frequency displayed. The former component reflects the parasympathetic activity while the latter displays the summated sympathetic and parasympathetic activities (Pomerantz et al., 1985; Pagani et al., 1986). The attenuations of MWSA by mental work load have been reported, and numerous reports on application research have documented the substitution of this index (e.g., Hyndman and Gregory, 1975; Egeland, 1982; Vincent et al., 1987; Mulder and Mulder, 1988; Waard et al., 1995; Noguchi, 1997).

Furthermore, a component related to thermoregulatory systems (VLF) can be established at much lower frequency bands (<0.03 Hz) in addition to these two components (Sayers, 1973). This component appears relatively unstable and rarely poses an issue in analyses of HRV.

Pagani et al. (1986) abbreviated the low-frequency MWSA and high-frequency RSA as LF and HF, respectively (e.g., Weise et al., 1989; Ori et al., 1992; Bootsma et al., 1994; Tuininga et al., 1995; Noguchi, 1997 etc). The HRV components have since been commonly abbreviated as LF and HF. However, note that some researchers designate MWSA and RSA as MF and HF (or RF), and describe thermoregulatory component around 0.03Hz as LF (e.g., Weise and Heydenreich 1989, Langewitz et al., 1991; Yeragani et al. 1994).

LF and HF components

From the power spectra of HRV, the integral power of frequency band for the respective components can be extracted and used for evaluation purposes. In such a case, the integrating frequency band becomes the focal issue. For the band widths of integration of LF and HF, a standard does not exist, and it may not be exaggerating to say that the situation is confusing with the procedure being controversial, complicated by the different standards adopted by the many researchers (Table 1).

With varying integration band, totally different values are produced, and direct comparisons among LF and HF values published by many scientific documents can no longer be validated.

Table 1  Frequency bands of integrating LF and HF

<table>
<thead>
<tr>
<th></th>
<th>LF band (Hz)</th>
<th>HF band (Hz)</th>
<th>Respiratory control (Hz)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Hayashi et al. (1997)</td>
<td>0.05–0.15</td>
<td>0.15–1.0</td>
<td>-</td>
</tr>
<tr>
<td>Malfatto et al. (1996)</td>
<td>0.03–0.15</td>
<td>0.15–0.35</td>
<td>-</td>
</tr>
<tr>
<td>Pagani et al. (1997)</td>
<td>0.09–0.11</td>
<td>0.22–0.24</td>
<td>-</td>
</tr>
<tr>
<td>Pomerantz et al. (1985)</td>
<td>0.04–0.12</td>
<td>0.224–0.280</td>
<td>0.25</td>
</tr>
<tr>
<td>Sayers (1973)</td>
<td>0.07–0.10</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>Task force (1996)</td>
<td>0.04–0.15</td>
<td>0.15–0.40</td>
<td>-</td>
</tr>
<tr>
<td>Veltman and Gaillard (1998)</td>
<td>0.07–0.14</td>
<td>0.15–0.5</td>
<td>-</td>
</tr>
<tr>
<td>Vincente et al. (1987)</td>
<td>0.06–0.14</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>Waard et al. (1995)</td>
<td>0.07–0.14</td>
<td>-</td>
<td>-</td>
</tr>
</tbody>
</table>

Fig. 3  HRV spectrum derived from MEM (10th order). Two components are revealed approximately in 0.1 and 0.25 Hz.
The band-width of integration of HF relates to the experimental conditions where regulation of controlled respiration has indeed been imposed, or the frequency used when a controlled respiratory condition has actually been imposed. As the peak of using MWSA is most likely to occur at frequencies below 0.1 Hz (around 0.07–0.09 Hz), it would be more precise to focus the band width of LF integration in sites slightly below 0.1 Hz to achieve more accurate analyses. As for the frequency band-width of LF, the subjects of analyses may vary according to focusing merely on MWSA or include the VLF component reflecting activities related to the body temperature regulatory mechanisms. Analyses encompassing low-frequency components to 0.03 Hz in the band width of LF integration would contain the latter choice (e.g., Malfatto et al., 1996).

HRV components and autonomic activities

Current interpretations using HF as an index for the parasympathetic nervous system encounters little or no dispute among the various scientists. However, varying interpretations on the analysis are attempted when LF is used. The LF analysis may be interpreted as follows:

1. LF is considered as an index for both sympathetic and parasympathetic activities.
2. LF is considered as an index for sympathetic activity.

Based on the experimental findings by Pomerantz et al. (1985), LF is often considered as an index for both sympathetic and parasympathetic responses. However, if the findings of Pomerantz et al. were to be adopted, the component enveloping the sympathetic components containing the LF would compose 17% of the total, revealing major involvement of the parasympathetic components. In other experimental results, the contribution of sympathetic components are of an ever lesser rate (e.g., Langewitz et al., 1991; Grossman and Kollai, 1993; Cacioppo et al., 1994).

The idea that LF reflects sympathetic activity is based upon the fact that LF may be in relation to the blood pressure fluctuation and the fact that LF shows significant increase during orthostatics (Weise and Heydenreich, 1989; Kobayashi, 1996).

From pharmacological studies using various blockers, LF may not be the only index representing sympathetic activities. The fact that LF increases in orthostatic load and its correlation with mean heart rate implies that it may be difficult to expound phenomena where LF is used as the index reflecting sympathetic activities.

Furthermore, explanations of the relevant LF/HF ratio may be pursued as follows:

1. The LF/HF ratio as an index balancing the sympathetic and parasympathetic responses, and
2. The LF/HF is considered as an index of sympathetic activities.

The interpretation with the LF/HF ratio as an index of sympathetic activities per se, LF is then considered as an index of both sympathetic and parasympathetic responses, where the dividing factor HF will render extraction of only the sympathetic effects possible. However, simply dividing by HF may not be able to completely eradicate the parasympathetic components from LF. As such, the LF/HF ratio may therefore appropriately interpret and serve as an index equilibrating the sympathetic and parasympathetic activities. In strict sense, the implication of this "equilibrium" does not represent the ratio of sympathetic/parasympathetic activity, and it cannot be denied that this approach represents a mere approximation.

Much room remains for discussion on the index-autonomic activity relationship, and future research studies will definitely complement the imperfect and incomplete status on indexing the state of the autonomic nervous system.

Control of respiration

The respiratory frequency approximates to 0.2–0.4 Hz during resting in general. In this case, although it is possible to separate the frequency band between MWSA and RSA, the respiratory frequency may vary greatly because of various factors, leading often to overlapping of band-widths of the LF component (Sayers, 1973; Kobayashi, 1995). As a result, clear separation between MWSA and RSA cannot be realized. As low-frequency respiration in the vicinity of 0.1 Hz displays extremely high amplitudes (Angelone and Coulter, 1964; Hirsch and Bishop, 1981; Kobayashi, 1997), values of the LF component play an influential role. This may be one of the factors contributing to HRV inconsistencies (Ori et al., 1992; Kobayashi et al., 1995).

From the above argument, it is therefore essential to impose control in conditioning a constant respiratory rhythm by using a metronome and other corrective procedures in measuring HRV. In fact, measurements of the HRV are often performed with regulatory control imposed on the subject to secure a constant respiratory frequency in recent times (e.g., Mukae and Sato, 1992; Puig et al., 1993). Respiratory frequency is not the only factor of concern in HRV; the tidal volume has to be considered as well. As such, in a stricter sense, controls of the respiratory frequency and tidal volume have to be regulated (Kobayashi, 1998).

References


Opmeer CHJM (1973) The information content of successive R-R intervals in ECG. Preliminary results using factor analysis and frequency analysis. Ergonomics 16 (1): 105-112


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