

Original article

Autonomic control of cardiorespiratory coupling in healthy subjects under moderate physical exercises

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Received 12 August 2019, Revised 30 September 2019, Accepted 15 October 2019

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Abstract: *Aim* of the study is to apply the analysis of phase dynamics to investigation of the coupling patterns between heart rate variability, respiration and peripheral circulation in healthy subjects at rest and after moderate physical exercises.

Material and Methods — 30-minutes electrocardiogram (ECG), photoplethysmogram (PPG) and respiration records were obtained from healthy subjects aged 22±2 (mean ± standard deviation) before and after active Martine Kushelevsky test (20 squats in 30 seconds). The coherence function was estimated between all the signals from each subject, and the phase dynamics modeling was used to detect the directional coupling in high-frequency (HF; 0.14-0.40 Hz) and low-frequency ranges (LF; 0.04-0.14 Hz).

Results — At rest (before the physical activity) no statistically significant couplings were detected between the HF rhythms of respiration and heart rate (RR intervals). For the HF rhythms in respiration and PPG, the influence respiration→PPG was detected. No couplings were detected between the LF rhythms. After the moderate physical exercise, predominant direction of the influence was RR intervals → PPG in the LF range and both RR intervals → respiration and PPG → respiration in the HF range. The influence PPG → respiration was delayed by several seconds.

Conclusion — adaptation to the moderate physical activity has led to the decrease in the overall coherence and changes in patterns of directional coupling between the LF and HF rhythms in respiration, heart rate and peripheral circulation. The obtained results confirm nonlinearity and complexity of the coupling patterns in cardiovascular system.

Keywords: baroreflex, autonomic control, respiration, photoplethysmogram, heart rate variability, active test, directional coupling detection, phase dynamics modeling, coherence function.

Cite as Kiselev AR, Borovkova EI, Simonyan MA, Ishbulatov YM, Ispiryan AYU, Karavaev AS, Navrotskaya EV, Shvartz VA. Autonomic control of cardiorespiratory coupling in healthy subjects under moderate physical exercises. *Russian Open Medical Journal* 2019; 8: e0403.

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Introduction

Many studies confirmed the interaction between the respiration and oscillations in cardiovascular system [1, 2]. Both the characteristics of cardiorespiratory couplings [3] and interactions of various cardiovascular oscillatory processes have potential importance for medical diagnostics [4]. For fundamental and practical purposes, it is important to investigate the characteristics of coupling between the oscillatory rhythms in heart rate variability (HRV), photoplethysmographic waveform variability (PPGV) and respiration. The coupling can be characterized by its direction, strengths and time delay.

Cardiovascular and respiratory systems are interrelated, nonlinear and space-distributed objects. Dynamical noises and measurement artifacts introduce further complexity. To study the coupling patterns in such systems it is necessary to choose the most precise and robust methods of data analysis and to carefully

set their parameters. The approaches based on analysis of instantaneous phases are perspective for such applications. External factors will cause significant changes in the instantaneous phase of the signal before any changes in the amplitudes become evident [5, 6]. For the analysis of biological data, many researchers use the methods based on phase dynamics modeling, which were first introduced in [7] and later modified and updated in [8, 9]. Phase dynamics modeling is particularly predominant in the analysis of complex brain potentials [10].

In the earlier studies, we estimated the strength and delay in coupling between the low-frequency oscillations (with main rhythm around 0.1 Hz) in heart rate variability and peripheral circulation. The dynamics of peripheral circulation was estimated from the photoplethysmogram (PPG) signals [11]. The study showed that in healthy subjects the influence heart rate → PPG is faster than the influence in the opposite direction. However, the

reliability of this result requires further investigations. The previous studies, including the studies on mathematical models, showed that the heart rate and PPG variability in the LF range are associated with the sympathetic control of heart rate [12-14] and vessels tone [15-17], respectively. Considering the coupling patterns between the LF variability rhythms in healthy subjects [4] it is reasonable to assume that one oscillatory process has a leading, dominant role.

The later study was aimed to adjust parameters of the methods to achieve better sensitivity and specificity in estimation of characteristics of coupling between the HRV and PPGV LF rhythms [18]. In that study, we refined the algorithms of pre-processing, instantaneous phases calculation, estimation of time scales over which the phase variation is modeled, estimation of needed length of the time series, the methods of statistical analysis. It was also revealed that investigation of time series longer than 20 minutes leads to significantly more precise estimations of coupling characteristics and lower dispersion of the estimations [19]. These results gave us ground to repeat the experimental study of coupling characteristics between the LF oscillations in HRV and PPGV to reevaluate the data, previously obtained from shorter time series [11]. We also decided to include the investigation of the cardiorespiratory interactions and the collective dynamics of autonomic control during moderate physical exercises.

Aim of the study is to apply the phase dynamics modeling to the investigation of the characteristics of coupling between the low-frequency rhythms in heart rate variability, photoplethysmogram waveform variability and respiration. The study involved healthy subjects at rest and after moderate physical exercise.

Material and methods

Subjects

The study involved ten healthy subjects aged 22 ± 2 (mean \pm standard deviation). All subjects gave written consent to be involved in the study.

Design of the experiment

The ECG, PPG and respiration signals were recorded twice for each subject, before and right after the Martine Kushelevsky test (20 squats in 30 seconds). The signals were recorded for 30 minutes during the both phases. During the recording sessions, the subjects were sitting comfortably with their eyes closed in a temperature controlled room. Respiration was not controlled. The experiments took place at least two hours after the previous meal.

The ECG signals were recorded in the I standard lead, the PPG signals were recorded from distal phalange of the left ring finger with 960 nm infrared sensor in reflected light. The respiratory signals were recorded with standard abdominal resistive strain sensor. All signals were recorded with the standard certified registrar – Encephalan–EEGR–19/26 (Medicom MTD, Russia) at 16 bit resolution and 250 Hz frequency for each channel. Band pass filter was set to 0.003–80 Hz for the ECG and PPG signals and to 0.003–80 Hz for the respiratory signal.

Data analysis

For the analysis, only the ECG and PPG signals without artifacts or extra systoles were selected. The sequences of RR-intervals

were extracted from the ECG signals. Equidistant cardiointervalogram (CIG) was then resampled from the RR-intervals with the cubic β -splines at 5 Hz. CIG was used as estimation of HRV.

For HRV, PPGV and respiratory signals we investigated the coupling characteristics using the coherence function and the phase dynamics modeling approach. The couplings between HRV and PPGV were investigated in 0.04–0.14 Hz range or so-called low-frequency (LF) range [20]. Couplings between the “HRV-respiration” and “PPGV-respiration” pairs of signals were investigated in 0.14–0.4 Hz range or so-called high-frequency (HF) range [20].

Calculation of the coherence function $C(f)$ is based on the estimation of the cross-spectra and provides the characteristics of coupling between the spectral components of the signals [21]. The coherence function $C(f)$ was group-averaged. Additionally, the sets of amplitude adjusted Fourier transform (AAFT) surrogate signals [22] were generated to control the statistical significance of the $C(f)$. The results were considered significant if the $C(f)$ was above the critical significance level at $p=0.05$. Only the statistically significant estimates of the $\hat{C}(f)$ were used for further analysis.

The coupling patterns and the characteristics were also investigated using the phase dynamics modeling approach [7]. The strengths of coupling in both directions was investigated ($G(\Delta_{2 \rightarrow 1})$ and $G(\Delta_{1 \rightarrow 2})$) for different time lags Δ between the signals. Analytical approach from [23] was used to estimate the 95%-significance interval. The detected couplings were considered statistically significant if estimations of coupling “strengths” $G(\Delta_{2 \rightarrow 1})$ or $G(\Delta_{1 \rightarrow 2})$ with the 95%-confidence interval were not crossing the zero level. The phase dynamics models predicted the phase variation over $\tau=10$ seconds, and Δ was ranging from 0 to 4 seconds.

For each time lag Δ , group-averaged $G(\Delta_{2 \rightarrow 1})$ and $G(\Delta_{1 \rightarrow 2})$ were calculated with the corresponding standard deviations $\sigma_{G(\Delta_{1 \rightarrow 2})}$ and $\sigma_{G(\Delta_{2 \rightarrow 1})}$. Statistically insignificant results were excluded from the study.

Results

Figure 1 presents the group-averaged coherence functions $\hat{C}(f)$ for HRV, PPGV and respiratory signals before and after the physical exercises. As seen from the figure, the mean value of $\hat{C}(f)$ decreased after the active test. We then calculated the average decrease of $\hat{C}(f)$ and its standard deviation: for the coherence function between the LF rhythms in HRV and PPGV it was 0.017 ± 0.006 (Figure 1b) and for the HF rhythms in PPGV and respiration it was 0.012 ± 0.004 (Figure 1c). Most prominent decrease (0.060) of $\hat{C}(f)$ was observed for the 0.1 Hz-rhythms in HRV and PPGV (Figure 1a). For 0.3 Hz-rhythms in respiration and HRV, the decrease of $\hat{C}(f)$ was 0.042 (Figure 1b), while for the 0.3 Hz-rhythms in respiration and PPGV, it was 0.023 (Figure 1c).

Figure 2 presents the estimated strength of the directional coupling between the LF and HF rhythms of HRV, PPGV and respiratory signals before and after the active tests. From Figure 2(a, d) it is evident that before the test no coupling is detected between the HRV and PPGV (Figure 2a). After the test, the coupling was detected; predominantly it was the influence HRV \rightarrow PPGV (Figure 2d).

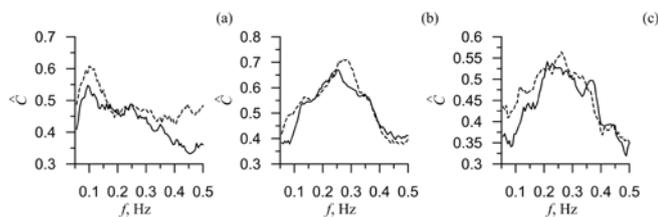


Figure 1. The group-averaged coherence functions $\hat{C}(f)$ between PPG and CIG (a), respiration and CIG (b), respiration and PPG (c). Dashed line is the coherence function before the physical exercises and solid line is the coherence function after the physical exercises. $G(\Delta_{2 \rightarrow 1})$ and $G(\Delta_{1 \rightarrow 2})$

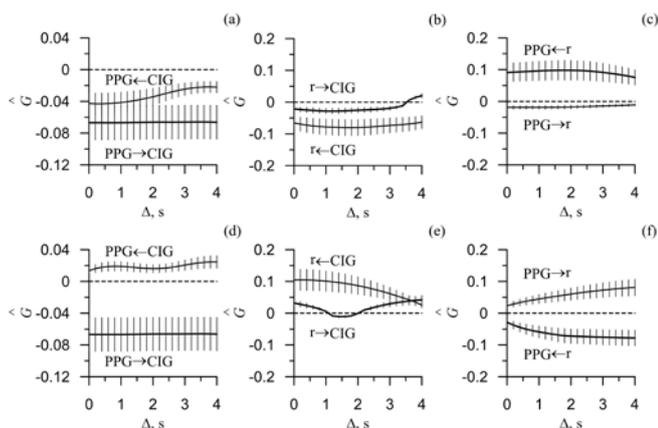


Figure 2. Results of the directional coupling detection using the phase dynamics modeling approach: (a) and (d) coupling between PPG and CIG, (b) and (e) coupling between respiration and CIG, (c) and (f) coupling between respiration and PPG. (a, b, c) coupling calculated before the physical exercise and (d, e, f) coupling calculated after the exercises. The solid line marks the group-averaged estimates of coupling strengths in both directions, $\sigma_{G(\Delta_{1 \rightarrow 2})}$ and $\sigma_{G(\Delta_{2 \rightarrow 1})}$. The vertical lines marks the standard deviation $\sigma_{G(\Delta_{1 \rightarrow 2})}$ and $\sigma_{G(\Delta_{2 \rightarrow 1})}$. The dashed lines mark the zero level. If $\sigma_{G(\Delta_{1 \rightarrow 2})}$ or $\sigma_{G(\Delta_{2 \rightarrow 1})}$ with the respective standard deviation $\sigma_{G(\Delta_{1 \rightarrow 2})}$ or $\sigma_{G(\Delta_{2 \rightarrow 1})}$ was larger than the zero level we assumed the coupling to be statistically significant for the respective time lag Δ .

From Figure 2(b, e) it is evident that before the test no couplings were detected between the HRV and respiration (Figure 2b). After the test, the coupling was detected in both directions; predominantly it was the influence HRV \rightarrow respiration (Figure 2e).

From Figure 2(c, f) it is seen that before the test the influence respiration \rightarrow PPGV is detected (Figure 2c). After the active test, the predominant coupling was detected in the opposite direction (Figure 2e). Moreover, Figure 2(e) shows several seconds delay in the influence PPGV \rightarrow respiration. It argues for the complexity of coupling pattern between the investigated signals.

Discussion

Understanding the dynamics of coupling between the rhythms of heart rate variability, peripheral circulation and respiration is important for the fundamental understanding of the circulation physiology. The study aimed to investigate this dynamics in healthy subjects before and after the moderate physical exercise. To estimate the activity of peripheral circulation, we investigated the PPG signals, because of their close relation to the blood filling of peripheral vessels [24].

The LF oscillations are present in a number of cardiovascular signals, including heart rate, arterial pressure and peripheral blood filling [15, 25-29]. The LF rhythms in heart rate and arterial pressure are caused by baroreflexory control [12-14]. The LF rhythms in PPGV are caused by the sympathetic control [15-17, 30] of the vessels' tone, oscillations of vessels' wall that have local myogenic origin [30], and variability in the blood flow through the distal arteries [31]. In [32] it is stated that the LF oscillations in HRV and PPG are functionally independent, but in healthy subjects they are synchronized for a majority of time [4]. In the patients with cardiovascular conditions, the level of synchronization is decreased that is potentially important for clinical applications [4].

The study revealed that in healthy subjects, before physical activity there are no couplings between the LF rhythms in HRV and PPGV. After the physical activity, the statistically significant influence HRV \rightarrow PPG appeared which may suggest that autonomic control of heart rate is dominant in the regulation of vessels' tone during the adaptation period. The study also revealed the significant influence respiration \rightarrow PPGV in the HF range. The revealed coupling can be explained by the fact that the HF rhythms in PPGV are passively affected by respiration [33].

After the moderate physical exercise, the coupling between the PPGV and respiration changes its direction. Since respiratory wave is mainly related to the venular flow, rise of its amplitude is a sign of decreased venous return [34]. We assume that decreased influence of respiration on PPGV is related to the increase in microcirculation after the physical activity. Analysis of the influence PPGV \rightarrow respiration revealed a time delay of several seconds. This result argues for the complexity of the coupling patterns in cardiovascular system.

For the HF rhythms in respiration and heart rate, we detected no significant coupling before the physical activity. However, after the active test, we detected the coupling in both directions with dominant influence HRV \rightarrow respiration. The coupling pattern of such complexity may be caused by the mechanical stimulation from the rib cage, since physical activity leads to deeper and faster respiration, and, therefore, to strengthened suction of a rib cage and increased venous return [35]. Another explanation is increased parasympathetic activity. Many researchers linked the HF rhythms in HRV with respiration and parasympathetic activity [20].

The decrease of the coherence function after the moderate physical exercise may be explained by the increased role of local control mechanisms.

We assume that obtained results can help in understanding of the adaptation dynamics that occurs in cardiovascular system in response to the moderate physical exercises. The results also confirm the nonlinearity and complexity of the coupling patterns, which can show a time delay of up to several seconds.

Conclusion

Adaptation of the cardiovascular system to the moderate physical exercises has led to the decrease in the average level of coherence and to changes in the directional coupling patterns between the HF and LF rhythms in respiration, heart rate and peripheral circulation. In particular, moderate physical activity has led to manifestation of the coupling between the LF rhythms in HRV and PPGV signals. Dominant direction of coupling was from HRV to PPGV. After the physical activity, the HF rhythms in both

HRV and PPGV started affecting the HF rhythms in respiration. The coupling between the PPGV and respiration was delayed by several seconds that confirms the complexity of coupling patterns between the respiration and peripheral circulation.

Limitations

Detection of weak coupling between the noisy nonlinear oscillations is a non-trivial task. We applied the most precise methods of weak coupling detection. However, we sacrificed a part of precision in order to achieve statistically significant results. Statistical insignificance of the detected couplings doesn't necessarily mean their absence. It only means that the data is insufficient to prove the presence of couplings. The insignificance of the results may be caused by high level of noises, weakening of coupling or absence of oscillations in the investigated ranges.

Such conditions require cautious interpretation of the absolute values of delays in couplings. Therefore we didn't give much attention to the exact values of this parameter.

Funding

The reported study was funded by the grant of the President of the Russian Federation, project No. MD-418.2019.7.

Conflict of interest

The authors declare that they have no conflict of interest.

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