ORIGINAL ARTICLE

Lifetime shift work exposure: association with anthropometry, body composition, blood pressure, glucose and heart rate variability

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ABSTRACT

Objective To evaluate the association between lifetime exposure to shift work and blood pressure, fasting glucose (FG), anthropometric variables, body composition and heart rate variability (HRV).

Methods Male shift workers (N=438) were evaluated using principal component (PC) analysis. The variables used were: weight, body mass index (BMI), waist circumference (WC), neck circumference (NC), hip circumference (HC), waist-to-hip ratio (WHR), waist-to-height ratio (WHtR), body fat mass (BFKg), body fat percentage (BF%), visceral fat area (VFA), FG, systolic (SBP) and diastolic blood pressure (DBP), and HRV variables. ECG was performed, extracting heart rate (HR), root mean square of the successive differences (RMSSD), high frequency (HF), low frequency (LF) and the LF/HF ratio. Using linear regression models, the lifetime shift work exposure was associated with each PC.

Results Five PCs were obtained, which accounted for 79.6% of the total variation of the data. PC1 (weight, BMI, WC, NC, HC, WHR, WHtR, BFKg, BF% and VFA) was designated as body obesity; PC2 (HF, RMSSD and LF) as good cardiac regulation; PC3 (SBP and DBP) as blood pressure; PC4 (LF/HF ratio and HR) as bad cardiac regulation and PC5 (WHR and FG) as insulin resistance. After age adjustment, the regression analysis showed that lifetime shift work was negatively associated with PC2 and positively associated with PC3.

Conclusions The association of lifetime shift work exposure with PC2 and PC3 suggests that shift work promotes unfavourable changes in autonomic cardiac control related to a decrease in parasympathetic modulation and an increase in blood pressure.

INTRODUCTION

Over the past two decades, occupational conditions have emerged as likely risk factors for cardiovascular diseases (CVD).1 Shift work has been highlighted among these conditions.1,2 Although some studies do not support the association between shift work and cardiovascular morbidity and mortality,3,4 several studies convincingly indicate the existence of this causal relationship and even suggest the potential physiopathological mechanisms involved.1,5

What this paper adds

- Shift work has been inconsistently associated with cardiovascular risk factors.
- Using principal component analysis, we obtained five principal components, which accounted for 79.6% of the total variation of the data, and grouped the measured variables into components of important biological value.
- Our results showed associations between lifetime shift work and cardiovascular risk factors, suggesting that shift work promotes unfavourable changes to autonomic cardiac control that are related to a decrease in parasympathetic modulation and an increase in blood pressure.
- It is necessary to determine effective strategies for health screening and promotion among shift workers, which represent an important operating mode of modern society.

Working in shifts, according to International Labour Office (ILO), is defined as a “method of organisation of working time in which workers succeed one another at the workplace so that the establishment can operate longer than the hours of work of individual workers” and can be classified into fixed shift (working time can be organised in two or three shifts: the early, late and night shifts) or rotating shift (workers may be assigned to work shifts that vary regularly over time). All of the different pathways explaining the connection between working in shifts and CVD share a common primary element: desynchronisation of circadian rhythms.5-7 This desynchronisation would impair the daily intracellular and extracellular physiological variations of the cardiovascular system, leading to a reduced ability of this system to adapt to external stimuli, with a consequent increase in tissue damage and an increased risk of CVD.8 In addition to physiological changes, some authors believe that circadian stress may lead to behavioural (eg, irregular eating, physical inactivity) and psychosocial (eg, working time incompatible with normal social life) changes. These different factors
may be intercorrelated and determine, directly or indirectly, a worse cardiovascular condition in shift workers. Hypertension, obesity, diabetes mellitus type 2 and cardiac autonomic dysfunction have been suggested as important members of this not fully elucidated physiopathological model that culminates in CVD due to circadian stress. The estimation of heart rate variability (HRV) using ECG is a non-invasive method for assessing cardiac autonomic regulation. HRV has important sympathetic and parasympathetic components, and the reduction of the latter has been associated with a worse general health status and greater cardiovascular morbidity and mortality. Furthermore, a decrease in vagal cardiac function accompanies and precedes the development of several risk factors, and high vagal activity is therefore a predictor of good health and good autonomic regulation. Several studies have associated shift work with a reduced global HRV, indicating the existence of a sympathetic preponderance that is detrimental to parasympathetic activity in this population. These unfavourable autonomic changes may be explained by a close association between the circadian clock and cardiovascular regulation because there are projections from the suprachiasmatic nucleus to the autonomic subdivision of the paraventricular nucleus of the hypothalamus. Moreover, drowsiness and inadequate sleep, common experiences among shift workers, are known to activate the sympathetic nervous system, leading to an increase in blood pressure and heart rate (HR), which may be a physiopathological explanation for the association between hypertension and shift work reported in recent studies.

In relation to obesity, elevated body mass index (BMI), elevated waist-to-hip ratio (WHR) and elevated body fat mass in kilograms (BFKg) are common findings among shift workers. These anthropometric and body composition changes are risk factors for changes in glycaemic metabolism and may mediate the relationship between shift work and diabetes mellitus type 2 because greater incidences of this disease are found in these workers than in those who do not work in shifts. However, there are a few studies of which results do not corroborate the association between shift work and alterations in glycaemia, blood pressure, anthropometry and body composition and HRV. In view of this inconsistency in the literature, this study was designed to evaluate the association between lifetime exposure to shift work and blood pressure, glucose levels, anthropometric variables, body composition and HRV. This study has some relevant points: the sample homogeneity (Brazilian, male, rotating shift, mine workers), multiple variables collected on the same sample and a principal component analysis (PCA) approach. PCA is a classical multivariate exploratory tool because it highlights the common variation between variables, allowing inferences about the potential biological meaning of associations between them without pre-establishing hypotheses on the cause–effect relationships.

METHODS

Study design/population

A cross-sectional study was performed with 438 adult Brazilian men, aged between 20 and 55 years, who work as operators of iron ore extraction machinery in four mines owned by the same mining company located in the Inconfidentes region (Minas Gerais, Brazil). The work regimen consisted of rotating shifts (alternating shifts), with 6 h of work followed by 12 h of rest. Each worker alternates between shifts and time off, and for every four shifts of work there is a 36 h rest period, after which a new cycle begins. This cycle consisted of four consecutive shifts: 19:00–1:00; 13:00–19:00; 7:00–13:00; 1:00–7:00. All of the employees had similar environmental and work conditions, that is, they were operators of iron ore extraction machinery working in air-conditioned trucks with protection against noise, water and dust. However, their lifetime shift work (years working in shifts) was different. The procedures were performed in the morning and during the employees’ longer off-duty period; the study was conducted from September 2011 to March 2012.

This study was conducted in accordance with the Declaration of Helsinki and was approved by the Ethics Committee of the local institution. All participants freely signed an informed consent form.

Anthropometric variables

The volunteers’ weight, height and circumferences were measured by trained professionals with volunteers in the orthostatic position and wearing light clothing. Waist circumference (WC) was measured at the mid-point between the last costal arch and the iliac crest. Hip circumference (HC), which was used to calculate the WHR, was measured at the level of the greater trochanter. Neck circumference (NC) was measured immediately below the laryngeal prominence. The WHR and waist-to-height ratio (WHtR) were calculated by dividing WC by HC and by height, respectively. The BMI was calculated by dividing the body weight by the squared height.

Body composition variables

The BFkg, the body fat percentage (BF%) and the visceral fat area (VFA) were calculated via segmental tetrapolar bioelectrical impedance using a body composition analyser, InBody model 720 (Biospace Co Ltd Factory, Korea).

Clinical-laboratorial variables

Blood pressure was measured with an automatic digital sphygmomanometer, model HEM705CP (Omron, Japan), in the right arm with the participant in the sitting position after 5 min of rest, three times, with a minimum interval of 1 min between readings. The three measures of systolic blood pressure (SBP) and diastolic blood pressure (DBP) obtained were computed and then averaged to be used for the statistical analysis. A sample of venous blood was collected for the determination of glucose levels after a 12–14 h fast using the glucose oxidase colorimetric enzyme method (Bioclin, Brazil).

Heart rate variability

A computer running WinCardio (Micromed, Brazil, V.5.5.1.3) software controlled the data acquisition of the ECG parameters. ECG recordings were collected at a sampling frequency of 1000 Hz and at lead II. An off-line peak detection algorithm (derivative plus threshold) was used to estimate fiducial R-wave points, after which the series was screened by hand and corrected for artefacts. Successive R-R intervals were estimated in milliseconds. The cardiac parameters were extracted for 3 min in the supine position:

- High frequency (HF), from 0.15 to 0.4 Hz: marker of parasympathetic activity;
- Low frequency (LF), from 0.04 to 0.15 Hz: possibly correlated to sympathetic and parasympathetic activity or to baroreflex sensitivity;
- LF/HF ratio: index of the interaction between sympathetic and parasympathetic activity (autonomic balance).
HR: correlated to increased sympathetic and decreased parasympathetic activity;\(^44\)

Root mean square of successive R-R interval differences (RMSSD): marker of parasympathetic activity.\(^43\)

The data processing followed the recommendations of the Task Force of the European Society of Cardiology and the North American Society of Pacing Electrophysiology.\(^44\) Matlab software (KARDIA) was used to analyse cardiac parameters.\(^46\)

**Lifetime shift work**

The mining company’s human resources department provided a list of dates (month/year) when its workers were hired. On the day of evaluation, the interviewer verified this information with the worker and asked if he had previous experience in shift work. After this, the number of months between the evaluation date and the real start date of shift work was calculated and the result was divided by 12. In this way, a rational number was obtained to represent the total time (in years) that the worker had performed shift work (lifetime shift work).

**Procedure**

After the employees arrived at the research laboratory and signed the informed consent form, a medical interview was performed to collect information on the use and dosage of medications, as well as to calculate the lifetime shift work. Next, body weight, height, WC, HC, NC, BFkg, BF\(^%\) and VFA were measured.

Subsequently, an ECG was performed for 3 min in the supine position. Finally, blood pressure was measured and a blood sample was collected for the biochemical determination of glycaemia. The employees had fasted for 12–14 h.

**Statistical analysis**

The description of variables was presented as mean (±SD) or median (P25–P75) according to normality as tested by the Shapiro Wilk test. The data were analysed using Statistica V.7.0 (StatSoft Inc) software and R Development Core 164 Team (2013). A p<0.05 was considered statistically significant.

PCA was used to group the 18 variables collected: weight, BMI, WC, NC, HC, WHR, WHtr, BFkg, BF\(^%\), VFA, fasting glucose (FG), SBP, DBP, HR, RMSSD, HF, LF and LF/HF ratio. PCA is a statistical multivariate method of which the primary aim is to detect the common variation between the original variables and then condense the data by obtaining other variables, referred to as principal components (PCs). The PCs obtained are linear combinations of the original variables and each PC is, by definition, not correlatable with another. The first PC (PC1) obtained accumulates most of the total variation between the studied variables, while the next components (PC2, PC3, etc) represent, respectively, less variation. So each new PC will progressively include smaller fractions of the variation, which are expressed by their ‘eigenvalues’. Each PC has one eigenvalue, which is obtained by the sum of the squared correlations between the original independent variables and the respective PC, and this sum represents the fraction of variation assigned to each PC. Although several PCs are obtained with PCA, according to the Kaiser criterion\(^47\) only the PCs whose eigenvalue is greater than 1 should be retained. In our study, a varimax rotation was used to improve the interpretability of the data and the cut-off point for the inclusion of a variable in the PC was established as variable loading ≥0.45.\(^48\) Moreover, the Kaiser-Meyer-Olkin test was used to verify the adequacy of the sample for performing PCA and was considered adequate if ≥0.5.

Because the PCs are variables derived from other original variables, each individual included in the study has a score related to each PC. Based on these scores, the relationship between the PCs and lifetime shift work was evaluated using univariate linear regression. Subsequently, to reduce the influence of age on these associations, an adjustment by age was performed to obtain the estimated lifetime shift work for each fixed age. In more detail, first, an adjustment was made to obtain estimated, typical lifetime shift work associated to each fixed age group. We achieved this using a linear regression model. The idea was that because (as it is known) there is a strong correlation between age and lifetime shift work, it would be appropriate to obtain such estimates in a manner that also accounts for this correlation. Therefore, by denoting age and lifetime shift work by Z and Y, respectively, the model adjustment that we considered was

\[
\hat{Y} = \beta_0 + \beta_1 Z,
\]

and estimating the parameters above directly by regression gave us the estimated values \(\hat{Y}_t, t=23; 35; 45; 55\), of Y for each of our age groups. In the sequence, these values were used in linear regression adjustments of each of the PCs PC\(_j\), \(j=2; 3; 4; 5\), in terms of Y. Because the PC\(_j\)s can be considered independent, this is adequate, and it is not necessary to consider a multivariate regression model to satisfactorily estimate the PC\(_j\)s in terms of lifetime shift work. Notice then that these are estimates within each age group. If the regression equations for the PC\(_j\) in terms of Y are

\[
PC_j = \gamma_{0j} + \gamma_{1j} Y, \quad j = 2, 3, 4, 5,
\]

we see that our PC estimates are given (with an ad hoc modification in the case of PC3, for which a quadratic model was found to be slightly better) by

\[
PC_j = \gamma_{0j} + \gamma_{1j} Y + \gamma_{2j} Y^2, \quad j = 2, 3, 4, 5.
\]

**Figure 1** Estimated values of each principal component (PC) for different lifetime shift work (estimated for fixed age). PC2 (circle) represents the cardiac component of good regulation. PC3 (triangle) represents the blood pressure component. PC4 (plus sign) represents the cardiac component of bad regulation. PC5 (X mark) represents the insulin resistance component.
Generally, these were the values we used in the present work to obtain the results referred to in figure 1 and to proceed with our analysis.

In order to contrast the PCA approach with the ‘standard’ modelling approach, linear regression analyses were applied to examine the relationship between the lifetime shift work and each variable of the PCs previously associated with the respective lifetime.

Finally, our analysis considered the wide range of medications reported to be modifiers of HRV. Two different physicians classified the medications used by each worker, and in the event of disagreement between them, a third physician was consulted. Different groups of medicines are known to differentially affect HRV; therefore, to identify the possibility of drug bias, all the analyses were conducted again after excluding workers who were taking medications that could influence the evaluated data.

RESULTS
The median age (P25–P75) of the 438 alternating shift workers participating in this study was 34 (31–39) years and the median of lifetime shift work exposure was 7 (5–10) years. The anthropometric, body composition, clinical-laboratorial and HRV variables are shown in table 1.

In the PCA, the first five PCs were retained because they presented eigenvalues ≥1.0 (Kaiser criterion). Together, these five components account for 79.6% of the total variation of the data. Table 2 shows the loadings related to the variables in each component. For PC1, which accounted for 40.21% of the variance, 10 variables had loadings greater than the predefined cut-off point, indicating a strong positive association between weight, BMI, WC, NC, HC, WHR, WHR, BFKg, BF% and VFA. In PC2, three variables (RMSSD, HF and LF) reached the cut-off point in the same direction of variation (positive loadings), corresponding to 14.95% of the total variation. Accounting for 10.02% of the total variation, only the pressure variables (SBP and DBP) were included in PC3. Contributing to 7.36% of the variation, PC4 was composed of positive loadings for HR and LF/HF. Finally, PC5 included the variables WHR and FG, accounting for 7.08% of the data variation. Because the analysis performed allows for naming the components according to what they appear to express from a biological point of view, the PCs were respectively named as: body obesity component; cardiac component of good regulation; blood pressure component; cardiac component of bad regulation and insulin resistance component. The sampling adequacy was satisfactory for PCA because the Kaiser-Meyer-Olkin test showed an adequacy value of 0.73.

The data from the linear univariate analysis of the association between lifetime shift work and each PC are shown in table 3. Component 2 (p=0.003) was negatively associated with lifetime shift work, while components 3 (p=0.023), 4 (p=0.017) and 5 (p=0.006) were positively associated with it. After age adjustment, the values of each PC were obtained for the different estimated lifetime shift work (5, 12, 18 and 24 years) for fixed ages (25, 35, 45 and 55 years old, respectively), which demonstrated that PC2 remained with a definitive negative association with

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* Loading denotes the contribution of the variable to each PC. A higher absolute value indicates that the variable has a major influence on the PCs. Bold numbers indicate absolute loadings ≥0.45. BMI, body mass index; HF, power in high-frequency range; LF, power in low-frequency range; LF/HF, ratio LF (mm²)/HF (mm²); RMSSD, root mean square of successive R-R interval differences.
lifetime shift work and PC3 with a definitive positive association (figure 1).

Table 4 shows the association between lifetime shift work and each variable of the PCs previously associated with the respective lifetime, allowing a comparison between ‘standard’ and PCA approaches. In general, the separate variables showed the same pattern of association with lifetime shift work as their PCs, however, the results of LF (p=0.077) and HR (p=0.069) were only marginally significant.

Ninety-three of the 438 employees were using some type of medication. Of those 93, 27 were using one of the following groups of medicines, which were not identified as HRV modifiers: proton pump inhibitors, corticosteroids, non-steroidal anti-inflammatory drugs, oral anticoagulants and/or statins. The other 66 were using at least one of the following groups, which can potentially influence HRV:49 antihypertensive drugs (thiazide-type diuretics; calcium channel blockers; ACE inhibitors; angiotensin receptor blockers; β-blockers), anticholinergics, β2-agonists, antidepressants, anxiolytics, sedatives/hypnotics, insulin, oral hypoglycaemic agents and/or levethyroxine. After excluding these 66 employees, we repeated the statistical analyses (data not shown) and noted that the significant results found for the total 438 employees remained. This allowed us to rule out the possibility of drug bias on the findings reported in this study.

DISCUSSION

The findings reported here point to the existence of five PCs that account for most (79.6%) of the total variation between the studied variables. As PCA groups variables that behave in similar fashion, each PC can be associated with a biological phenomenon and thus can be given a fictitious name. PC1, named the body obesity component, included the anthropometric and body composition variables (body weight, BMI, WC, HC, NC, WHR, WHtR, BFkg, BF% and VFA). PC2 included the variables HF, RMSSD and LF, and was named the cardiac component of good regulation. PC3 included the variables SBP and DBP, and was named the blood pressure component. PC4 was designated as the cardiac component of bad regulation, and included LF/HF and HR. Finally, PC5, named the insulin resistance component, included WHR and FG. After controlling for the influence of age, lifetime shift work had a definitive negative association with PC2 and a positive association with PC3.

PC1 included all of the anthropometric and body composition variables assessed. These variables are known to be highly intercorrelated,40 indicating that although some are obtained from specific body parts, they can also reflect the individual’s general obesity and vice-versa. That is why component 1 was named the ‘body obesity component’. Component 5, in turn, the ‘insulin resistance component’, included only FG and WHR. This finding corroborates recent studies50–51 demonstrating that WHR is the best anthropometric measure to predict diabetes type 2. Curiously, PC5 did not include VFA, suggesting that FG in shift workers is more associated with WHR than with VFA itself. This finding appears contradictory because WHR is considered a risk factor exactly because it estimates the visceral adipose tissue.26,52 However, it should be highlighted that the quantity of visceral fat for a given WHR can be substantially altered53,54 by individual differences in the quantity of total body fat and therefore it is possible that, in a given population (eg, shift workers), WHR reflects an ‘adipose balance’ that is more associated with glucose alterations than VFA. There is not sufficient evidence in the literature to prove this hypothesis, but there are studies26–55 relating occupational and psychosocial factors to a more specific anthropometric change on the WHR.

PC2 included three components of HRV: HF, RMSSD and LF. Different from HF and RMSSD, which are already well established as parasympathetic measures as well as markers of good health and lower cardiovascular morbidity and mortality,15,16,43,44 the origins and clinical significance of LF are still controversial. This measure has already been indicated to be an index of cardiac sympathetic discharge26 and a double marker: a greater influence of sympathetic and smaller influence of parasympathetic discharge.2,43,44 Recently, some studies have provided strong evidence that LF is associated with baroreflex sensitivity; therefore, it is a marker for the proper functioning of this reflex.45–47 The present study corroborates the baroreflex thesis because the PCA showed that the LF variable has a strong positive association with HRV measures that are predictors of good health and good regulation.45,46 This strong association also corroborates a recent review58 where LF was argued to be mostly determined by the parasympathetic system. Therefore, PC2 was named the ‘cardiac component of good regulation’.

Conversely, PC4 was considered to be the ‘cardiac component of bad regulation’ because it was comprised of the variables HR and LF/HF ratio, of which high levels have been related to
several risk factors for the development of CVD. Although there are controversies, which even include a discussion on the meaning of LF, it is relatively accepted that the LF/HF ratio reflects the cardiac sympathovagal balance, and an increase in this ratio indicates sympathetic dominance and a decrease indicates parasympathetic dominance. Because the sympathovagal balance is directly related to HR, a positive association between the members of component 4 was expected. Component 3 was named the ‘blood pressure component’ because it presented the variables SBP and DBP.

Using linear regression models, we noted a positive association between lifetime shift work and the components ‘bad cardiac regulation’, ‘blood pressure’ and ‘insulin resistance’. There was a negative association with the ‘cardiac component of good regulation’ and no association with the ‘body obesity component’. After a re-analysis considering the influence of age on the associations found, there remained a definitive positive association of lifetime shift work with the ‘component blood pressure’ and a negative association with the ‘cardiac component of good regulation’. Therefore, the longer the lifetime exposure to shift work, the higher the blood pressure and the lower the autonomic cardiac regulation. Collectively, these findings suggest the occurrence of a gradual worsening of health, particularly cardiovascular health, in long-term shift work.

There is much controversy about the association between shift work and cardiovascular risk factors, but numerous recent studies reported consistent associations with those reported here. In a sample of 360 workers in alternating shifts, Ha and Park observed that the longer the lifetime shift work (years), the higher the blood pressure, but they found no association between years of shift work and FG levels and BMI. Additionally, van Amelsvoort et al found a positive association between lifetime shift work (years), and BMI and WHR, with a more pronounced effect for WHR. The authors argue that these anthropometric changes are likely because shift work induces eating changes, physical inactivity and alterations in the circadian rhythm of metabolic processes such as digestion and insulin and cortisol secretion.

Regarding the HRV parameters, our group did not find any studies examining the associations between HRV and lifetime shift work in the literature. However, several other studies indicate that these workers have less favourable autonomic cardiovascular regulation (greater sympathetic modulation and lower parasympathetic modulation) compared with the control group (day shift workers). Compared with daytime workers paired by age, BMI and cholesterol levels, and found lower RMSSD levels and higher LF/HF ratios among shift workers. The authors argue that this finding indicates lower parasympathetic modulation and greater sympathetic cardiac modulation in shift workers, changes that are both associated with a higher risk of CVD. However, there are studies that did not find differences in the autonomic cardiovascular regulation between shift workers and daytime workers.

Comparing the PCA approach with the analysis shown in table 4 (standard ‘approach), we can note that although the relationship between lifetime shift work and the individual variables and their PCs were similar, PCA revealed some advantages. First, PCA is a statistical multivariate method of which the primary aim is to detect the common variation between the original variables (without a priori suppositions of the evaluator) and then condense the data by obtaining other variables, referred to as PCs. Besides detecting the main source of inherent variation among the investigated variables, the PCA approach allowed us to perform an age adjustment, which would not have been possible in the conventional method because there is a strong association between lifetime shift work and age.

In general, our results corroborate the association between shift work and cardiovascular risk, but some limitations of the study should be noted. First, as the sample is composed of male shifts workers with the same work arrangement of alternating shifts, comparisons of gender and shift work specificities (permanent or alternating; clockwise or counterclockwise, etc) could not be performed. Second, although the employees had similar environmental and work conditions (they were all operators of iron ore extraction machinery working in air-conditioned trucks with insulation against sound, water and dust), data on behavioural factors (eg, eating habits, physical activity, sleep-wake pattern, smoking) were not collected. This information could be useful in determining potential mediators of the associations found. It should be emphasised that, considering the current pathophysiological hypotheses to explain the influence of shift work on health status, adjustment for psychosocial and behavioural factors is not appropriate in studies such as ours because they should not be considered confounders, but rather steps in the causal chain. Third, a cross-sectional design is not the best design for establishing the risks associated with lifetime shift work exposure, but without some preliminary identification of transversal associations, longitudinal studies of the development of cardiovascular risk factors during a professional life in shifts are unlikely to be undertaken. Fourth, when one investigates the influence of lifetime shift work, age is an important confounder because it is inherently and positively associated with time (years) working in shifts. In our study, an age adjustment was performed to reduce the influence of this confounding variable. However, as discussed by van Amelsvoort et al, there is no way of eliminating this influence completely. Fifth, we did not measure breathing during the ECG recordings. Although it is known that breathing potentially influences the HF variable, Pentila et al demonstrated that RMSSD values are not affected by the respiratory pattern. In the present study, both parasympathetic variables (HF and RMSSD) appeared in the same PC, and even contributed with almost identical variable loadings; therefore, we can infer that the lack of breathing control most likely did not bias the study results.

CONCLUSION

When the influence of age is diminished, longer lifetime shift work exposure clearly promoted a decrease in the cardiac component of good regulation (HF, RMSSD and LF) and an increase in the blood pressure component (SBP and DBP). These changes represent a gradual worsening of general health status during a lifetime of shift work, which is a possible explanation for the increased risk of CVD in this population. Because shift work constitutes an important operational mode, accounting for 15% of labour arrangements in the USA and approximately 15–20% in modern society, the findings of this study indicate the need for determining effective strategies for the evaluation and promotion of health among shift workers. These measures may anticipate the diagnosis of health deterioration and ensure early intervention to prevent the development of CVD in this population, which is a population that is essential for the current economic and social challenges of nations worldwide.

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Contributors RMNN, MLS and SNF contributed in the conception and design of the study. BBS, NMM, JMO and GGGLS compiled the data for analysis and made the analysis plan. JMO monitored the data acquisition of the heart rate variable parameters. FLPO performed the statistical analyses. BBS, NMM, FLPO and GGGLS participated in data interpretation. BBS and GGGLS drafted the manuscript. All the authors critically reviewed and approved the final version of the manuscript.

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