Higher Autonomic Activation Predicts Better Performance in Iowa Gambling Task

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Objective: To evaluate the relationship between the autonomic nervous system basal state and performance in decision-making tasks.

Background: The link between performance in decision-making tasks and acute changes in autonomic parameters during their execution has been extensively investigated. However, there is lacking evidence regarding the relationship between decision making and basal autonomic state.

Methods: Resting autonomic nervous system activity in 18 healthy individuals was assessed by means of heart rate variability (HRV) analysis before conducting 3 different decision-making tasks: an ambiguous one, the Iowa Gambling Task; a test that assesses risk-taking behavior, the Game of Dice Task; and a test that assesses reversal learning behavior, the Reversal Learning Task. The tasks were administered in a random manner.

Results: There was a direct correlation between the Iowa Gambling Task net score and the resting low frequency HRV (r = 0.73; P < 0.001), which is strongly influenced by sympathetic activity. No correlations were found between HRV and the Game of Dice Task net score or the Reversal Learning Task last error trial.

Conclusions: The results are compatible with the idea that a higher basal activation of autonomic nervous system is beneficial for subsequent decision-making process.

Key Words: autonomic nervous system, heart rate variability, decision making, Iowa Gambling Task

(Cogn Behav Neurol 2011;24:93–98)

Decision making is a cognitive process crucial for daily life. Whether it is an important decision or an irrelevant one, we are unavoidably required to make choices every day. Some of these choices are made on the basis of clear and complete information and are rationally taken. However, in other situations, unconscious processes play an important role in identifying the best option to follow. The processes taking place in these situations are poorly known.

Aiming to bring some light to this issue, Bechara et al developed the Iowa Gambling Task (IGT), which was designed to recreate uncertain decisions made in real life. On the basis of prefrontal-damaged patients’ deficient IGT results, they developed the Somatic Marker Hypothesis, which holds that normal individuals guide their decisions toward advantageous choices, even before consciously knowing they are convenient. For example, it was observed that normal individuals develop a sympathetic activation on the verge of taking disadvantageous choices. Such sympathetic activation would be a marker for avoiding these decisions. Skin conductance responses (sympathetic activation) were not generated in patients with ventral medial prefrontal cortex damage, who perform poorly on IGT. In line with this, a low-IGT-performing group of normal individuals lacked anticipatory skin conductance responses. In addition, the high-performing group showed anticipatory deceleration of heart rate before choices associated with frequent losses. This cardiac response was also absent in the low scorers.

Despite these provocative observations, a causal relationship between autonomic activation and IGT performance is far from being established. For instance, patients suffering from pure autonomic failure, a pathology that makes them unable to generate autonomic arousal due to peripheral denervation of the autonomic nervous system (ANS), did perform at least as well as healthy controls at IGT. This would indicate that peripheral autonomic modifications are not necessary for guiding decision-making processes. Thus, it has been
argued that a deficient IGT performance can be related to alteration of specific decision-making mechanisms other than the one explained by the somatic marker hypothesis.\textsuperscript{4} The preference of the individual for safer or riskier choices is the most studied of these possible alternative mechanisms.\textsuperscript{4,5} Reversal learning, that is, the ability to suppress successfully learnt behavior when it is no longer correct, has also been implicated.\textsuperscript{4}

Owing to the subtle trial-to-trial regulation inherent in somatic marker hypothesis, it does not seem possible to test causality by modifying ANS response at this time scale. Thus, the existence of a more general relationship between decision-making scores and ANS needs to be assessed.

To our knowledge, there is no published study linking decision-making performances with basal ANS activation. To test whether the baseline status of ANS activity relates to complex decision-making processes as assessed by IGT, Heart Rate Variability (HRV) analysis was used before exposure to the task while controlling for other mechanisms, such as risk-taking behavior and reversal learning, which could be involved in this kind of decision-making processes.

**MATERIALS AND METHODS**

This is an observational study in healthy volunteers on the association between cardiac autonomic activity at rest and performance in the IGT as a decision-making process. The study was approved by the local ethics committee.

**Participants**

Healthy volunteers (n = 18), with the mean age of 47.7 years (SD: 15.4; range, 22 to 70 years; 72.2% women), gave full informed consent and were paid for participating, irrespective of their performance in the tasks.

**Decision-Making Tasks**

**IGT**

The IGT was used to assess decision making. The IGT has a number of design issues that could make its interpretation difficult, such as the involvement of other specific decision-making mechanisms like risk-taking behavior or reversal learning.\textsuperscript{4} However, it has frequently been used in the assessment of decision making under ambiguity.\textsuperscript{1,2,5}

In this task, participants are given simulated money and presented with 4 decks of cards. Each deck is either advantageous or disadvantageous in terms of total monetary outcome. Selection of disadvantageous decks leads to high rewards but even higher losses, whereas advantageous decks bring relatively low rewards but even lower losses. At the beginning of the test, the participants totally ignore what the best choices are. During its execution, they must learn the right strategy to win as much simulated money as possible and/or avoid losing it.

We used a computerized version of the IGT, previously described.\textsuperscript{2} This version was modified by introducing a 6-second delay between each decision. This feature was intended to allow future studies that test heart response during the execution of IGT. The IGT net score was calculated as the total number of advantageous choices minus the total number of disadvantageous choices. Following other studies,\textsuperscript{2,5} the net score was further divided into 5 blocks, each of 20 consecutive card choices.

**Game of Dice Task (GDT)**

The GDT\textsuperscript{5} assesses risk-taking behavior under known probabilities by having explicit rules for gains and losses. Participants are told to maximize 1000 units of fake money within 18 rounds of dice throws. The task is based on guessing what number or combination of numbers will result from rolling a die. There are 4 different alternatives to take a guess: single numbers or combinations of 2, 3, or 4 numbers. Each alternative is associated with specific fictive gains or losses according to the probability of winning (for single numbers the bet is 1000, for combination of 2 numbers the bet is 500, for combination of 3 numbers the bet is 200, and for combination of 4 numbers the bet is 100). Payment in case of winning is always the same amount of money that is betted. Thus, the best option is to choose the combination of 4 numbers. The GDT net score is calculated by subtracting the number of times the participant selected the disadvantageous options (selecting a single number or a combination of 2) from the number of times the participant selected the advantageous options (selecting a combination of 3 or 4 numbers).\textsuperscript{5}

**Reversal Learning Task (RLT)**

In this task, participants learn to touch 1 of two simple patterns that are shown one at a time in a touch screen. They gain a point by touching one of the patterns or by not touching the other. They lose a point by touching the one they should not touch or by not touching the one they should. Once they have learnt to do this, the outcomes are switched. The informed result, RLT last error trial, is the last trial in which an error was made.\textsuperscript{6} Thus, higher result numbers are associated with worse performance at reversing the learnt behavior.

**ANS activity**

ANS activity was assessed by HRV analysis. This analysis is based on the fact that the relatively constant heart rate generated by the sinus node is modulated by several factors that result in a complex heart rate signal. HRV high-frequency component (0.15 to 0.4 Hz) is related to respiratory sinus arrhythmia, and therefore is mediated solely by parasympathetic activity. The low-frequency component (0.04 to 0.15 Hz) is related to baroreflex control and depends mainly on sympathetic and also on parasympathetic influences. A very low-frequency component (\textless 0.04 Hz) of an uncertain origin is also found and has been attributed to thermoregulatory fluctuations in vasomotor tone and to humoral factors.
such as the renin-angiotensin system. In addition, HRV shows fractal scaling and nonlinear properties derived from the complex interaction of the autonomic tone and its central organization and from exteroceptive and interoceptive influences. The nonlinear dynamics of HRV has been characterized in various forms, including short-term fractal exponent α (zs) and Sample Entropy (SampEn). Signal recording: Electrocardiogram signal was recorded using a digital Holter device (Holter HCAA 348/Holtech/Servicios Computados S.A./Buenos Aires/Argentina) and stored in a solid-state memory. Ventricular depolarization (R waves) was detected through the device software. The time elapsed between R waves (RR intervals) was then computed. We visually identified and manually tagged premature and lost beats in the original file of RR intervals. These abnormal beats were replaced by RR intervals resulting from linear interpolation. Only those segments with >85% qualified beats were included in the analysis.

Time domain HRV analysis: Quantitative time series analysis was performed on heart rate by evaluating measures of variation over time. Among these, mean RR interval (RRm) quantifies the mean heart rate, standard deviation of all normal RR intervals (SDNN) represents a coarse quantification of overall variability, and square root of the mean squared differences of successive normal RR intervals (RMSSD) measures high-frequency heart rate variations.

Frequency domain (spectral) measurements of HRV were obtained by Fast Fourier Transform. It included total area (TA, total spectral power, 0 to 0.4 Hz, ms²), very low frequency (VLF, < 0.04 Hz, ms²), low frequency (LF, 0.04 to 0.15 Hz, ms²), high frequency (HF, 0.15 to 0.4 Hz, ms²), their percentage values, and the LF/HF ratio. Nonlinear HRV analysis: The zs and the SampEn were used as nonlinear HRV indexes. The scaling exponent zs, based on the “detrended fluctuation analysis,” quantifies the short-term (<11 beats) fractal correlation properties of the interbeat time data. Values of zs close to 0.5 are associated with uncorrelated RR intervals, whereas values close to 1.5 are associated with strong correlation between RR intervals. Values close to 1 are characteristic of fractal-like processes, associated with the dynamic behavior of time series generated by complex systems, such as the autonomic regulation of the sinus rhythm in healthy individuals. SampEn measures the degree of irregularity of the RR interval time series. Regular sequences will result in lower SampEn values, whereas random behavior is associated with larger SampEn values. These methods have been previously described. Uncorrelated and irregular behavior is usually associated with parasympathetic prevalence.

Experimental Procedure
First, individuals were required to sit quietly for 10 minutes while recording their electrocardiogram signal.

After this, they were evaluated for decision making. IGT, GDT, and RLT were conducted in a randomized order.

Statistical Analyses
Values were expressed as mean ± SD. Normality of distributions was evaluated by using a Kolmogorov-Smirnov Test. Association between IGT net score and sex was evaluated by mean of α test for independent variables. Bivariate correlations of IGT net score with age, GDT net score, and RLT last error trial were explored by means of a Pearson correlation test. In addition, correlations between the decision-making tasks and the different HRV components were explored by means of the aforementioned test.

Finally, to establish whether significant correlations between IGT net scores and HRV indexes could be explained by age, sex, or specific decision-making mechanisms underlying IGT, partial correlations were conducted controlling for age, sex, GDT net score (as an index of risk behavior), and RLT last error trial (as an index of reversal learning).

RESULTS
Significance of Kolmogorov-Smirnov test was >0.10 for all variables except for RLT last error trial, which was between 0.05 and 0.10. This variable was log transformed for statistical analyses to allow a better fit to a normal distribution, being the significance of Kolmogorov-Smirnov test for the transformed variable equal to 0.501.

Mean IGT net score was 11.3 ± 29.2, mean GDT net score was 7.9 ± 8.1, and mean RLT last error trial score was 11.2 ± 11.6. A total of 72.2% of the participants completed the IGT with positive (>0) net scores (that is, more advantageous than disadvantageous choices).

Basal HRV values were (mean ± SD): RRm 845.09 ± 104.26 ms, SDNN 41.42 ± 13.97 ms, RMSSD 30.77 ± 11.66 ms, ln TA 7.25 ± 0.67 ms², ln VLF 0.24 ± 0.25 ms², ln LF 0.04 ± 0.15 Hz ms², ln HF 0.15 to 0.4 Hz ms². The correlations between decision-making tasks scores and HRV indexes are shown in Table 1. Significant correlations were found for IGT score with global HRV (SDNN, TA) and low-frequency HRV (LF and LF%) (Table 1 and Fig. 1). No significant correlations were found between the other tasks and HRV measurements (Table 1). HRV indexes that were significantly correlated with IGT net score were further correlated with the scores of the different IGT blocks, as shown in Table 2. Correlations become significant with the last 3 blocks (Table 2).

IGT net score showed no significant association with sex (T = -1.696, P = 0.109), whereas it was significantly correlated with age (r = -0.60, P = 0.009). GDT net score (r = 0.28, P = 0.260) showed no
correlation with IGT net score or with the net scores at any block of the task. Transformed RLT last error trial yielded a significant correlation with IGT net score ($r = -0.62$, $P = 0.006$) and the net scores at blocks 3 ($r = -0.63$, $P = 0.005$) and 5 ($r = -0.65$, $P = 0.003$). As a higher RLT last error trial indicates worse performance, the correlation between performance in IGT and RLT is in fact positive.

When controlling IGT net score for sex, age, GDT net score, and RLT last error trial, significance was still observed for LF ($r = 0.73$, $P = 0.003$) and TA ($r = 0.56$, $P = 0.037$) but not for LF% ($r = 0.50$, $P = 0.066$) and SDNN ($r = 0.52$, $P = 0.056$). Similar results were observed for the correlation between IGT net scores of the last blocks and HRV indexes, where significance after controlling for other variables was maintained for LF (Block 3: $r = 0.667$, $P = 0.009$; Block 4: $r = 0.716$, $P = 0.004$; Block 5: $r = 0.744$, $P = 0.002$) and TA (Block 4: $r = 0.642$, $P = 0.013$; Block 5: $r = 0.671$, $P = 0.009$).

**DISCUSSION**

The main finding of this study is that performance in the IGT was related to baseline resting LF HRV, an association that is maintained even after controlling for age, sex, risk-taking behavior, and reversal learning processes.

There is growing evidence showing that resting autonomic patterns can influence cognitive performance, although none of these studies specifically assessed decision making. For example, grouping healthy individuals according to their resting HRV, the high HRV group showed better results and faster reaction times in a continuous performance test that assesses both executive and nonexecutive functions. Our results seem to contradict these findings as they report that individuals with higher high-frequency HRV (a measure of the parasympathetic activity) perform better at executive function tasks. Unfortunately, no mention was made in those studies of the low-frequency HRV component.

On the other hand, our results are in line with a study that stated that lower resting baroreflex sensitivity, usually associated with a reciprocal reduction of parasympathetic activity and an increase of sympathetic activity, predicted higher values in all parameters of attentional capacity. The results of the study suggest that the intuitive idea of an association between improved attentional capacity and higher resting levels of parasympathetic cardiac tone (probably as a reflection of lower anxiety) does not universally hold. According to the investigators, the sympathetic prevalence is interpreted as necessary to establish a pattern of cardiovascular adjustment suitable for coping with increased mental demand.

Another potential explanation for our results is that at least some of the brain regions involved in complex decision making are the same regions responsible for autonomic arousal. In line with this, Critchley et al...
observed that activity in several areas, mainly anterior cingulate cortex (ACC), was strongly associated with increased LF power during cognitive effort. Furthermore, patients with ACC damage failed to generate contextually appropriate autonomic activation, especially sympathetic. Although related to different aspects of the task, activation in ACC has been repeatedly demonstrated during execution of IGT and some of its variants.21–25

In addition, one must take into account the possibility that the correlation between IGT net scores and low frequency HRV could be attributed to other specific decision-making mechanisms underlying IGT.4 Previous publications demonstrated that the GDT could be used as an index of risk-taking behavior during the last blocks of IGT.5 However, our results did not replicate the correlations between those tasks, probably due to the small number of participants. Reversal learning is also known to be a process occurring during IGT execution.4 Although there is no established test related to the IGT to assess reversal learning, the reported correlation between these tasks makes it reasonable to use the RLT as a control of the reversal learning process during the IGT. The correlation pattern obtained between RLT last error trial and the scores of IGT blocks is different from the published pattern for GDT net score correlation with trial and the scores of IGT blocks. However, our results did not replicate the possibility that the correlation between IGT net scores and low frequency HRV could be attributed to other specific decision-making mechanisms underlying IGT.4

In this study, the possibility that correlations between IGT net scores and HRV indexes could be explained by the mechanisms discussed above was ruled out by conducting a partial correlation controlling for age, sex, GDT net score, and RLT last error trial. As the results remained significant for LF HRV, basal sympathetic activation may play an important role in decision making. In addition, the correlation is significant only in the last 3 blocks of the IGT, supporting the hypothesis that basal ANS activity is critical in the evolution of decision-making processes.1

The small sample size should be noted as a limitation of this study. Correlations reported herein are larger than those previously reported between physiological measurements and decision-making tasks.31 A possible explanation is that Pearson correlation coefficients may be biased toward higher values due to the small sample size of the study. Further research with more participants is necessary to confirm these results.

To sum up, autonomic activity could be an important factor for the decision-making processes occurring during the IGT, although it is not established whether previous autonomic arousal is crucial for improving decision making or whether it is just an index of neural cortical processes. Future investigations should focus on establishing whether there is a causal relationship between higher LF-HRV and decision-making tasks, possibly by interfering ANS basal state, either pharmacologically or by means of physical training.

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