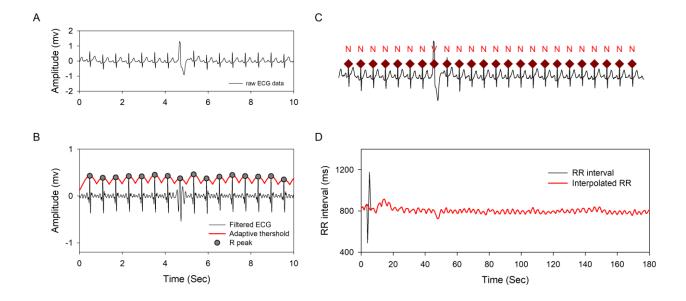
Supplemental Text 1

Section 1: Pre-processing of ECG data

In this study, the R waves were first detected using the automatic algorithm with a detection accuracy of ~99.98% based on the online MIT-BIH Arrhythmia Database (<u>https://physionet.org/</u>). A bandpass finite-impulse response filter was applied on the ECG signal to remove baseline wandering and increase the accuracy of the QRS detector even in the presence of a noisy signal. Furthermore, the algorithm automatically identifies abnormal beats, such as premature atrial and ventricular contractions. Next, the QRS waves were detected using a temporally adaptive threshold to find the locally prominent activation wave (R wave). Supplemental Figure S1 illustrates the QRS detection procedure for the recorded ECG data. The raw data (A) were bandpass-filtered, and an adaptive QRS detector was then applied to the data (B). The automated annotated results generated by the QRS detector were visually inspected and corrected (C). Finally, normal-to-normal sinus intervals, extracted from the corrected R-R interval time series (D; black line) by removing the abnormal beats, were resampled and interpolated by cubic spline interpolation at an even interval of 0.125 of a second (i.e., 8 Hz; D, red line). These steps ensure highquality detection of normal heart beats.



Section 2: Continuous wavelet transform (CWT)

The adopted continuous Morlet wavelet function is shown as

$$\Psi_0(t/s) = \pi^{-1/4} e^{j\omega_0 t/s} e^{-1/2(t/s)^2}$$

where t and s are time and scale, respectively. W_0 is the frequency of the wavelet. The convolution of the analyzed function g(t) with a scaled wavelet function shifted by τ is then defined as CWT.

$$W(s,\tau) = \int g(t)\Psi_s(t-\tau)dt$$

The relationship between the temporal and spectral aspects of the Morlet wavelet is shown below:

$$\lambda = \frac{1}{f} = \frac{4\pi s}{\omega_0 + \sqrt{2 + \omega_0^2}}$$

where λ is the Fourier wavelength and g(t) is derived from the evenly resampled normal-to-normal time series of heart beats (RR intervals). The time function of the instantaneous high-frequency (HFi) and low-frequency (LFi) components is derived as the integral of CWT over the frequency band of HF (0.15–0.4 Hz) and LF (0.04–0.15 Hz), respectively. That is:

$$HFi(\tau) = \int_{0.15Hz}^{0.4Hz} W(f,\tau)df$$

$$LFi(\tau) = \int_{0.04Hz}^{0.15Hz} W(f,\tau)df$$

The temporal LFi/HFi ratio is the ratio of the LFi component to the HFi component. The HFi of the heart beat series was used as a marker of vagal modulation², while the ratio of the LFi to the HFi (LFi/HFi) was generally considered to be an indicator of the balance between sympathetic and vagal modulation.^{3,4}

Data normalization

To eliminate inter-subject differences in baseline autonomic function, the derived parameters were normalized individually to a common scale of 0–100 by their maximum and minimum values (formula 1):

$$X' = \frac{X - X_{min}}{X_{max} - X_{min}} \times 100(1)$$

where x is the value in the series, x_{min} is the lowest value, and x_{max} is the highest value. The transformed variables had the same normalized value range and distribution in each patient with the same normalized unit. The parameters were

calculated and normalized offline using MATLAB (2016b, MathWorks Inc., Natick, MA, USA).

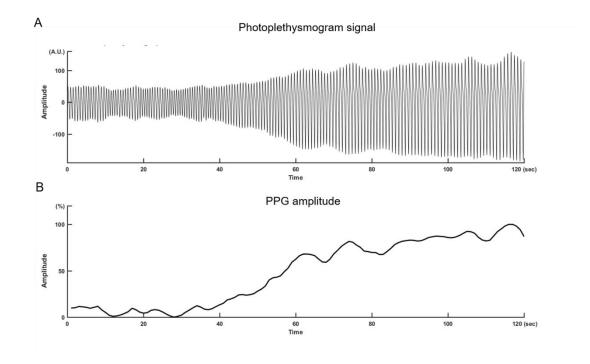
Section 3: Fourier analysis

A Fourier spectrum power over the HF and LF frequency bands was calculated based on the published guidelines.⁵ A Gaussian window function centered at a certain time period was multiplied by the whole R-R signal before the Fourier transform to maintain the same frequency resolution over the spectrum.

Section 4: Pulse photoplethysmography analysis and amplitude extraction

To extract the pulse photoplethysmography amplitude (PPGA) from the PPG, we applied a bandpass filter (e.g., 1–4 Hz) to remove noise, baseline wandering, and amplitude fluctuations related to respiration. The critical points of each PPG waveform were derived from the points of the zero crossing of their first derivatives.

Supplemental Figure S1 shows (A) the changes in finger PPG after administration of propofol and (B) the corresponding normalized PPGA (B).



Supplemental Figure S1. (A) Changes in finger PPG after administration of propofol. (B) Corresponding normalized PPGA. The amplitude between the trough and peak of each pulse wave in PPGA was then calculated to assess changes in the state of the autonomic system, especially for peripheral vasoconstriction related to sympathetic activation. The PPGA sequences were also resampled at 1 Hz to match the temporal changes in autonomic function assessed by CWT.

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