Determinants of heart rate variability in obstructive sleep apnea syndrome during wakefulness and sleep

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Obstructive sleep apnea syndrome (OSAS) is characterized by repetitive episodes of upper airway collapse during sleep. The obstruction of inspiratory airflow results in asphyxia and subsequent vigorous efforts to breathe, leading eventually to transient arousal and restoration of upper airway patency. These dramatic events exert a profound acute influence on the cardiovascular system. It is believed that long-term exposure to episodic apnea and arousal constitutes an independent risk factor for systemic hypertension, heart failure, myocardial infarction, and stroke (10, 33, 34). In particular, the evidence suggests that abnormal autonomic control is likely a key factor in the causal pathway linking OSAS to these cardiovascular diseases (36, 42).

Time-domain and spectral analyses of heart rate variability (HRV) have been employed as noninvasive methods of assessing cardiovascular autonomic function in patients with OSAS (40, 28). However, these analyses yield information that reflects only the net effect of all the factors that contribute to heart rate control, thus providing little insight into which of the underlying physiological mechanisms are most affected by the abnormality. For instance, fluctuations in blood pressure can affect heart rate through stimulation of the arterial and cardiac baroreceptors (24). This mechanism is believed to be a key determinant of the 0.1-Hz periodicity frequently observed in the heart rate power spectrum in humans. Fluctuations in blood pressure entrained by respiration contribute to some extent to the oscillation more commonly known as “respiratory sinus arrhythmia” (RSA). However, in this case, at least two other mechanisms have also been shown to play an important role (14). Animal studies have pointed to a direct central link between the respiratory and cardio regulatory rhythm-generating neurons in the medulla, so that vagal cardioinhibitory motoneurons of the nucleus ambiguus, which influence heart rate, receive a powerful inhibitory synaptic input throughout inspiration and are excited in the early expiratory or postinspiratory phase (35). A third suggested mechanism for RSA is vagal feedback from PSR, which modulates efferent outflow to the sinus node (2). In their classic studies on anesthetized dogs, Anrep and colleagues (2, 3) used specific denervation of the lung to determine the importance of vagal feedback from lung stretch receptors and tested the importance of the central link of respiratory to cardiovascular regulation by using a cross-cerebral circulation technique to cause hypocapnia-induced apnea (in the recipient dog) without the need for augmented lung stretch. Either of these mechanisms acting alone was shown to be capable of producing a significant RSA in the anesthetized dog. In a study on normal humans and patients who had undergone lung or heart transplantation, Taha et al. (38) showed an obligatory role for lung vagal feedback in mediating RSA but concluded that, in the intact human, central and baroreflex modulation of RSA are also important. Thus different relative contributions from these mechanisms could yield the same magnitude of RSA.

Recently, we introduced a new noninvasive approach that provides a more precise assessment of autonomic function in patients with OSAS (7, 17). This approach combines an experimental protocol, in which a randomly modulated breathing sequence is used to induce fluctuations in heart rate and blood pressure with a computational model that allows estimation of the temporal relations between pairs of these signals. The experimental protocol presents the advantage of allowing the study of OSAS patients during wakefulness and sleep. The computational model enables HRV to be partitioned into a component directly correlated with respiration and a component representing arterial baroreflex (ABR)-mediated effects, thus allowing a less biased estimation of the ABR dynamics. In

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the present study, we have extended this model to include a more complete characterization of the mechanisms involved in RSA. More specifically, we have divided the respiratory-correlated contribution into two components: one representing the respiratory-cardiac neural coupling and the other representing the effect of vagal feedback from the PSR. As in our previous work (7, 17), it was assumed that the various interactions among the cardiorespiratory variables involved in the modulation of heart rate could be described by a linear, time-invariant, stationary parametric model.

Through the application of a novel experimental protocol in conjunction with a mathematical model of HRV, we sought to estimate and quantify the dynamics of the three main physiological mechanisms involved in the control of HRV, i.e., vagal lung stretch reflex, central cardiorespiratory coupling, and ABR; and to determine how these underlying mechanisms of autonomic control of HRV are altered by OSAS as well as by changes in wake-sleep state.

METHODS

Experimental protocol and instrumentation. Nine untreated patients with moderate to severe OSAS (apnea-hypopnea index \( \geq 44.1 \pm 2.8 \text{ h}^{-1} \)) and eight normal controls participated in overnight sleep studies. All subjects were male except for one female in the control group. Age was not significantly different between the two groups: 44.9 ± 2.8 and 50.1 ± 2.0 yr for OSAS patients and controls, respectively. However, body mass index was significantly higher in the OSAS patients (39.7 ± 4.2 vs. 28.4 ± 1.2 kg/m\(^2\), \( P < 0.05 \)). All subjects were normotensive and were free of diabetes, significant cardiac arrhythmia, congestive heart failure, and lung disease. Informed consent was obtained before each study. The protocol for the studies was approved by the University of Southern California Institutional Review Board.

Each subject was connected via nasal mask to a computer-controlled bilevel pressure ventilator (model S/T-D 30, Respironics, Pittsburgh, PA). Measurements of mask pressure (Pmask) and airflow (Q˙) were obtained from a reference pneumotachometer (model 3700, Hans Rudolph, Kansas City, MO). A chin strap was used to keep the mouth closed during wakefulness, the test protocol was applied at least three times in each sleep-wake state. Test sequences in which arousals or outright awakening occurred were terminated and were not repeated until the subject returned to a stable sleep state. Sleep stages were scored according to conventional criteria (31).

Data analysis. R-R intervals (RRI) and systolic and diastolic blood pressure (SBP and DBP) were deduced beat to beat and resampled at 2 Hz using the Berger algorithm (8). Very-low-frequency oscillatory behavior or baseline drift was observed in some of the datasets. These nonlinearities were removed by detrending the datasets before applying spectral analysis and modeling. The detrending procedure consisted of first fitting a fifth-order polynomial to the signals and then removing this curvilinear trend from the time series. The power spectrum of the subtracted trend was visually examined to verify that the detrending process had little effect on frequencies \( >0.04 \text{ Hz} \). After detrending, spectral analysis of RRI, DBP, and SBP was performed using an autoregressive modeling approach described previously (17). The respiratory signals were also resampled at 2 Hz, so that each respiratory value would be synchronized with the corresponding resampled RRI, SBP, and DBP values.

Model. On the basis of our knowledge of the underlying physiology (14, 24), fluctuations in heart rate, respiration, and blood pressure are assumed to be interrelated through the closed-loop control scheme (Fig. 1). For instance, apart from the direct mechanical effects of intrathoracic pressure changes, fluctuations in blood pressure are due in part to changes in cardiac output resulting from variations in heart rate as well as sympathetically driven variations in peripheral resistance. The focus of this study is limited to the other portion of the closed-loop model that accounts for HRV (Fig. 1). Fluctuations in RRI are decomposed into four components. The first three components correspond to the autonomically mediated mechanisms mentioned earlier: 1) vagal feedback from PSR, represented in this model by the coupling between fluctuations in V and RRI, 2) central medullary coupling between respiratory and cardiovagal neurons (RCC), represented in this model by the coupling between fluctuations in respiratory muscle pressure (P\(_{\text{mus}}\)) and RRI, and 3) ABR-mediated fluctuations, represented also in this model by the interaction between blood pressure and RRI. The fourth component encompasses all other influences not explained by the baroreflex or the two respiratory-related components, such as changes in venous return during breathing and changes in chemoreceptor feedback.

A linear autoregressive model with triple exogenous inputs was employed to estimate the impulse responses and transfer functions of the three components of the model schematized in Fig. 1. The change in RRI from the mean at time \( t \) (\( \Delta \text{RRI}(t) \)) is related to weighted combinations of past values of \( \Delta \text{RRI} \), changes in \( P_{\text{mus}} \) (\( \Delta P_{\text{mus}} \)), changes in lung volume (\( \Delta V \)), and changes in SBP (\( \Delta \text{SBP} \)), as given by the following difference equation

\[
\Delta \text{RRI}(t) = - \sum_{i=1}^{m} \delta_i \Delta \text{RRI}(t - i) + \sum_{j=1}^{m} \beta_j \Delta V(t - j - \Delta \text{DBP})
+ \sum_{k=0}^{m} \gamma_k \Delta P_{\text{mus}}(t - k - \Delta \text{RCC}) + \sum_{l=0}^{m} \phi_l \Delta \text{SBP}(t - l - \Delta \text{ABR})
+ W_{\text{ABR}}(t) \quad (t = 1, \ldots, N)
\]

where \( \Delta \text{DBP} \), \( \Delta \text{RCC} \), and \( \Delta \text{ABR} \) are the delays associated with the corresponding mechanisms, \( W_{\text{ABR}}(t) \) represents the variability of RRI
not explained by the model, \( a_i, b_j, c_k, \) and \( d_l \) represent unknown constant coefficients that need to be estimated from the data, and \( N \) is the total number of data points used in the analysis.

It is imperative to note that the system under study operates in a closed loop (Fig. 1), so that changes in heart rate can subsequently affect SBP through changes in cardiac output. In general, this condition, in which the model input is dependent on its output, can lead to erroneous parameter estimates when conventional analysis techniques are employed. To circumvent this problem, we formulated the model equations in the time domain so that "causality" constraints could be imposed, i.e., the model output was constrained mathematically to be dependent on only past values of the inputs. Previous studies employed similar methodologies that essentially allow the closed loop to be "opened" computationally (4, 7, 12, 17, 27, 41).

A delay of \( \geq 0.5 \) s was assumed for \( D_{\text{ABR}} \), the latency of the baroreflex impulse response. Physiologically based constraints were also imposed on the respiratory-related delays. Several previous studies on humans and animal preparations have noted that, during the average breathing cycle, heart rate begins accelerating in the last segment of the preceding expiratory phase (14, 18). Therefore, the model was allowed to adopt negative values of \( DR_C \). On the other hand, data from subjects who were passively ventilated suggest that, in this case, \( V \) precedes or is roughly in synchrony with \( RRI \) (38). Thus, in this case, \( V \) should precede \( RRI \) (or be almost synchronous with it). As such, the range of \( D_{\text{PSR}} \) was restricted to positive values close to zero. For each combination of delays, least-squares minimization was employed to estimate the unknown coefficients in Eq. 1.

As well, for each combination of delays, a measure of the quality of fit, known as the minimum description length (MDL), was computed as follows (32)

\[
\text{MDL} = \log(J_R) + \frac{\text{total no. of parameters} \times \log(N)}{N}
\]

where \( J_R \) is the variance of the residual errors between the measured data and the predicted RRI. Note that MDL decreases as \( J_R \) decreases but increases with increasing model order. Selection of the “optimal” candidate model was based on a global search for the minimum MDL; in addition, this optimal solution had to satisfy the condition that the cross-correlations between the residual errors and past values of the three inputs \( \Delta V(t), \Delta P_{\text{mus}}(t), \) and \( \Delta \text{SBP}(t) \) were statistically indistinguishable from zero.

With use of this model, it was possible to estimate the impulse response functions, \( h_{\text{ABR}}(t), h_{\text{PSR}}(t), \) and \( h_{\text{RCC}}(t) \), which characterize the gains and temporal properties of the baroreflex, PSR, and central-coupling mechanisms, respectively (Fig. 1): \( h_{\text{ABR}}(t) \) quantifies the time course of the change in RRI resulting from an abrupt 1-mmHg increase in SBP, \( h_{\text{PSR}}(t) \) quantifies the time course of the fluctuation in RRI associated with a very rapid inspiration and expiration of 1 liter of air, and \( h_{\text{RCC}}(t) \) quantifies the time course of the fluctuation in RRI associated with an abrupt 1-cmH2O increase in \( P_{\text{mus}} \). In linear systems theory, the impulse response provides a complete characterization of the dynamic properties of the system in question, because the response of this system to any arbitrary input can be predicted by mathematically convolving the input with the impulse response (20). Each of the impulse response functions, \( h_{\text{ABR}}(t), h_{\text{PSR}}(t), \) and \( h_{\text{RCC}}(t) \), was calculated by setting the associated input to unity at \( t = 0 \) and zero at all
other times, setting all other inputs and \( W_{RR(t)} \) to zero, and using \( I \) to compute the resulting RRI response. Once the impulse response functions were estimated, it was also possible to compute the relative contributions of each of the three mechanisms to the overall predicted \( \Delta \text{RRI} \) by convolving each impulse response with the corresponding input sequence. The ratio of the variance of the resulting convolved signal (representing the portion of the predicted \( \Delta \text{RRI} \) explained by the corresponding input) divided by the variance of the overall predicted \( \Delta \text{RRI} \) (obtained when the 3 inputs were included together) was taken as an index to quantify the relative contribution of each mechanism.

To quantify the strength of each mechanism (PSR, ABR, and RCC) on the modulation of HRV, two indexes representing the gain of each interaction were computed. 1) The impulse response magnitudes (IRM) were calculated as the difference between the maximum and minimum values of the estimated impulse responses. 2) The dynamic gains (DG) were estimated by first taking the Fourier transform of the impulse responses to obtain transfer functions and then calculating the average transfer function gains between 0.04 and 0.45 Hz. This range covers the span of frequencies pertinent to HRV. The dynamic characteristics (e.g., response time) of each mechanism were also quantified from the impulse response by computing the delay or latency (\( D_{\text{PSR}}, D_{\text{RCC}}, \) or \( D_{\text{ABR}} \)) and the duration between start of the impulse response and its first major peak or valley (\( T_{\text{peak}} \)).

Estimation of \( P_{\text{mus}} \). The respiratory rhythm-generating neurons in the medulla control phrenic activity, which, in turn, drives the respiratory muscles. Our model assumes, as previous physiological studies have suggested, a direct central link between these respiratory neurons and cardioregulatory neurons in the medulla, which, in turn, modulate heart rate. Because respiratory neural activity is not directly measurable in humans, \( P_{\text{mus}} \) was chosen to represent central respiratory drive. \( P_{\text{mus}} \) was deduced from \( Q \) and mask pressure (\( P_{\text{mask}} \)) signals using the following equation (26).

\[
P_{\text{mus}} = RQ + EV - P_{\text{mask}}
\]

The first term on the right-hand side of Eq. 3 represents the pressure used to overcome the resistance (\( R \)) of the respiratory system, whereas the second term represents the pressure associated with respiratory elastance (\( E \)). Previous studies have demonstrated that postinspiratory \( P_{\text{mus}} \) can persist for up to 70% of the time of expiration in anesthetized normal subjects (6) and in anesthetized patients with kyphoscoliosis, in whom respiratory elastance is markedly increased (5). Because \( E \) is also increased in OSAS patients, it is expected that postinspiratory \( P_{\text{mus}} \) also persists in a manner similar to that in normal subjects. So, on the basis of the previous findings, it is expected that postinspiratory braking is probably present to a similar degree in the OSAS and control groups and would reduce the variability in both groups to a roughly equal degree and, thus, would not contribute to major differences in the ABR, RCC, or PSR functions. Under this assumption, in the present study, expiration was taken to be entirely passive to make the computations and interpretations as simple as possible.

To calculate \( P_{\text{mus}} \), it was necessary to estimate the respiratory mechanics parameters, \( R \) and \( E \). \( R \) was estimated using a pulse technique (26). Brief pressure pulses, with amplitude of 5 cmH\(_2\)O and duration of 500 ms, were applied at randomly selected breaths through computer control of the inspiratory pressure level of the ventilator. Each pulse was delivered at the beginning of inspiration, when the elastic recoil of the respiratory system is least. To calculate \( R \), the increase in pressure above continuous positive airway pressure (CPAP) during the pulse was divided by the peak increase in flow elicited by the pulse. This procedure was applied before the delivery of each test protocol. Estimates of \( R \) were based on the average to at least five responses.

\( E \) was estimated as follows. At end expiration on a randomly selected breath, an abrupt change in pressure of 2–4 cmH\(_2\)O and duration of five breaths was imposed on the subject by controlling the inspiratory pressure level of the ventilator. To estimate \( E \), the change in inspiratory pressure was divided by the difference between the end-expiratory volume levels before and immediately after the pressure step. At least three estimations of \( E \) were performed before the application of each test protocol.

In the normal subjects during stage 2 sleep, \( R \) was found to be \( 6.21 \pm 0.34 \text{ cmH}_2\text{O}\cdot\text{s}^{-1} \cdot \text{l}^{-1} \), and \( E \) was \( 9.38 \pm 0.69 \text{ cmH}_2\text{O}\cdot\text{l}^{-1} \). These values are consistent with normal respiratory mechanics (1, 25). However, \( R \) is higher and \( E \) slightly lower than the corresponding values reported by Meza et al. (26), who employed similar estimation methods in normal subjects during nonrandom eye movement (NREM) sleep. The discrepancy between our estimates and theirs is likely due to application of a smaller CPAP applied (2–3 cmH\(_2\)O) in our subjects. In contrast, in the study of Meza et al., the applied CPAP was titrated to a level (~5 cmH\(_2\)O) sufficient to minimize \( R \) during NREM sleep.

**Statistical analysis.** To facilitate statistical comparison, compact descriptors were derived from each model impulse response and its corresponding transfer function. These included the two measures of gain (IRM and DG) and the two measures of temporal behavior of the impulse response (delay and \( T_{\text{peak}} \)) described above (see Model). Each estimated parameter was subjected to two-way repeated-measures analysis of variance (with subject group as nonrepeated factor and sleep-wake state as repeated factor). Post hoc multiple pairwise comparisons (Student-Newman-Keuls test) were carried out if statistical significance (\( P < 0.05 \)) was indicated.

**RESULTS**

**Sample time series.** A representative segment of data obtained from one of the OSAS patients is displayed in Fig. 2.
CPAP of ~7 cmH₂O was applied throughout the duration of sleep. Figure 2 shows the randomly timed increases in inspiratory pressure delivered during the test procedure. The consequent variability imposed on the breathing pattern is quite evident in amplitude and duration. These respiratory variations, in turn, led to corresponding fluctuations in SBP and RRI. However, spontaneous fluctuations in SBP that were largely independent of the respiratory changes also occurred. Application of the random sequence of inspiratory pressure changes helped reduce the correlation between V and Pmus. For instance, at ~50 s, the applied inspiratory pressure occurred in synchrony with a large inspiratory effort (Pmus), whereas at ~130 s, there was no intrinsic respiratory drive during the ventilator-imposed breath, and from ~170–200 s, spontaneous breaths were generated by the subject in the absence of any applied inspiratory pressure.

Because the subjects were heavily instrumented, total sleep time was spent primarily in stage 2 and rapid eye movement (REM) sleep. Thus comparisons of the results were made across only three states: wakefulness, REM sleep, and NREM stage 2 sleep.

Measures of cardiovascular variability. Mean RRI was significantly lower in OSAS patients than in controls (799 ± 39 vs. 980 ± 44 ms in wakefulness, P < 0.005); in both groups, mean RRI increased from wakefulness to sleep (P < 0.02). Average minute ventilation during the test protocol in the control subjects was 9.0 ± 0.02 l/min, respectively. The corresponding values for the OSAS group were 9.9 ± 0.5 l/min, in wakefulness, REM sleep, and stage 2 sleep, respectively. The initial negative dip is consistent with the well-accepted assumption a biphasic shape, in which there was an initial large negative undershoot followed by a smaller positive overshoot. The subsequent positive overshoot in the impulse response reflects the vagal feedback from PSR, was a fast initial positive overshoot and a subsequent smaller negative undershoot. The estimated RCC impulse responses from the same control subject and OSAS patient are shown in Fig. 3. In both individuals, the primary component of hPSR(t), which represents the vagal feedback from PSR, tended to be similar in magnitude in the OSAS patient and the control subject.

The estimated RCC impulse responses from the same control subject and OSAS patient are displayed in Fig. 3. Note that hRCC(t) was substantially smaller in magnitude in the OSAS patient than in the control subject during stage 2 and REM sleep. Also, hRCC(t) in the normal subject and OSAS patient assumed nonzero values before t = 0, indicating the previously observed tendency for changes in RRI to precede changes in respiratory effort (12, 27, 41). Both hPSR(t) and hRCC(t) assumed a biphasic shape, in which there was an initial large negative undershoot followed by a smaller positive overshoot. The initial negative dip is consistent with the well-accepted notion that the RSA consists of an acceleration of heart rate (or, equivalently, a reduction in RRI) during inspiration. The subsequent positive overshoot in the impulse response reflects the decrease in heart rate during expiration (25).

Sample traces of the estimated baroreflex impulse responses from the same control subject and OSAS patient are also displayed in Fig. 3. In this case, there was an initial large positive overshoot and a subsequent smaller negative undershoot, consistent with the notion that the net baroreflex response to an increase in blood pressure is a slowing of heart rate (or, equivalently, an increase in RRI). In all three states, hABR(t) was substantially smaller in magnitude in the OSAS patient than in the control subject. In the OSAS patient and the control subject, there was a clear increase in the magnitude of hABR(t) from wakefulness to stage 2 sleep.

Statistical comparison of model parameters. Table 2 displays the group-averaged results for all compact descriptors derived from the PSR, RCC, and ABR impulse responses; Fig.
changes in RRI precede changes in mechanical inspiration. Negative values, reflecting previous findings by others that patients than in the controls (greater in the controls. RCC gain was lower in the OSAS higher in OSAS in all states (time to peak of the RCC impulse response was significantly increased from wakefulness to sleep (P < 0.02 for IRM_RCC and P < 0.03 for DRCC); however, this increase was clearly greater in the controls. RCC gain was lower in the OSAS patients than in the controls (P < 0.04 for IRM_RCC and P < 0.02 for DRCC), but the difference was largest in sleep. The time to peak of the RCC impulse response was significantly higher in OSAS in all states (P < 0.03). DRCC assumed negative values, reflecting previous findings by others that changes in RRI precede changes in mechanical inspiration.

Table 2. Model parameter estimates

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Control</th>
<th>OSAS</th>
<th>P</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Wakefulness</td>
<td>REM</td>
<td>Stage 2</td>
</tr>
<tr>
<td>PSR</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>IR magnitude, ms/l</td>
<td>24.4±5.3</td>
<td>37.1±8.5</td>
<td>35.2±5.9</td>
</tr>
<tr>
<td>Dynamic gain, ms/l</td>
<td>31.9±6.0</td>
<td>46.1±6.7</td>
<td>44.3±7.5</td>
</tr>
<tr>
<td>Latency, s</td>
<td>0.22±0.88</td>
<td>0.37±0.13</td>
<td>0.37±0.15</td>
</tr>
<tr>
<td>Time to peak, s</td>
<td>1.87±0.21</td>
<td>2.96±0.41</td>
<td>2.34±0.37</td>
</tr>
<tr>
<td>RCC</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>IR magnitude, ms/cmH2O</td>
<td>3.5±5.5</td>
<td>5.8±1.1</td>
<td>5.4±0.8</td>
</tr>
<tr>
<td>Dynamic gain, ms/cmH2O</td>
<td>6.5±0.9</td>
<td>10.4±1.9</td>
<td>9.5±1.4</td>
</tr>
<tr>
<td>Latency, s</td>
<td>−1.40±0.13</td>
<td>−1.34±0.23</td>
<td>−1.18±0.15</td>
</tr>
<tr>
<td>Time to peak, s</td>
<td>2.50±0.18</td>
<td>2.31±0.23</td>
<td>2.25±0.18</td>
</tr>
<tr>
<td>ABR</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>IR magnitude, ms/mmHg</td>
<td>2.5±0.3</td>
<td>5.6±1.5</td>
<td>6.2±1.0</td>
</tr>
<tr>
<td>Dynamic gain, ms/mmHg</td>
<td>2.7±0.5</td>
<td>4.6±1.5</td>
<td>4.9±0.9</td>
</tr>
<tr>
<td>Latency, s</td>
<td>0.84±0.10</td>
<td>1.25±0.18</td>
<td>1.18±0.21</td>
</tr>
<tr>
<td>Time to peak, s</td>
<td>2.06±0.15</td>
<td>2.31±0.44</td>
<td>2.37±0.39</td>
</tr>
</tbody>
</table>

Values are means ± SE. PSR, pulmonary stretch reflex; IR, impulse response; RCC, respiratory-cardiac coupling; ABR, arterial baroreflex. *P < 0.05.
DISCUSSION

Delineation of major components of HRV. Many previous studies have sought to determine the origin of RSA (2, 3, 35).

Table 3. Relative component contributions to heart rate variability

<table>
<thead>
<tr>
<th>Component</th>
<th>Control</th>
<th>OSAS</th>
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<tbody>
<tr>
<td>PSR</td>
<td></td>
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<td>RCC</td>
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<tr>
<td>ABR</td>
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</table>

However, most of these studies involved the use of anesthetized animal preparations and surgical or pharmacological techniques to eliminate one or more of the hypothesized mechanisms. In humans, because these highly invasive methods are clearly not applicable, ingenious experiments have been devised to delineate the different components of RSA. For example, in the study of Taha et al. (38), mechanical positive-pressure ventilation at high tidal volume was used to inhibit the central control of respiration (i.e., elimination of RCC output). In another study (30), phasic neck suction was applied to stimulate the ABRs at a prescribed frequency and, thus, enhance the output of the ABR mechanism. Although these techniques are noninvasive, they are nonetheless rather intrusive and impractical to apply in a clinical setting, particularly during sleep.

In contrast, the experimental protocol that we applied in the present study is relatively noninvasive and generally well tolerated during sleep in normal subjects and patients with OSAS. Combining this experimental procedure with a model-based analysis allows us to computationally delineate the mechanisms of RSA. As well, the test protocol, in which positive airway pressure is introduced at randomly selected inspirations, serves as a means of perturbing the cardiorespiratory system over a broad range of frequencies. This facilitates the determination of the contributions of the three mechanisms over a dynamic range that extends beyond the band of respiratory frequencies (0.15–0.4 Hz) generally associated with RSA. Furthermore, broad-band stimulation has been shown to enhance the accuracy and reliability of parameter estimation (22). The random application of positive airway pressure also is useful for a third purpose. During spontaneous breathing, Pmus is strongly correlated with V, because a breath with stronger respiratory drive would generally produce a larger tidal volume; under such conditions, it would not be possible to distinguish the contributions of PSR from those of RCC. However, the introduction of a randomly modulated ventilatory sequence acts to decorrelate Pmus and V, because a large tidal volume can also result from a breath with weak intrinsic respiratory drive that has been bolstered by the ventilatory assist (Fig. 2).

An important feature of this study is the use of a mathematical model that allows the contributions from the three inputs to be delineated. Although respiration and blood pressure are treated as though they are mutually independent inputs, these variables and heart rate are part of a larger closed-loop control system (Fig. 1). However, delays in this closed-loop system allow us to computationally partition the feedforward and feedback portions, so that each portion of the system can be treated as though it is operating under open-loop conditions (20). For instance, in the baroreflex portion of the closed-loop system, present changes in RRI are constrained in the mathe-
mathematical model to be influenced by only past, but not present or future, fluctuations of blood pressure. This imposed constraint (termed causality in systems engineering terminology) forces the estimation scheme to converge toward a solution that reflects the effect of blood pressure on RRI (i.e., the baroreflex) rather than one that reflects the effects of RRI on blood pressure (i.e., the feedforward component). This approach of "temporal delineation" can only be applied by employing a model that has been formulated in the time domain. In contrast, frequency-domain methods, such as techniques that employ cross-spectral analysis, do not allow the implementation of causality constraints. The application of this type of computational approach to cardiovascular control has appeared in publications by other researchers (12, 27, 41).

Effects of CPAP. During sleep in the OSAS patients, CPAP was applied at individually prescribed levels to ensure upper airway patency throughout the test procedure. We believe that this was an important part of the experimental procedure, because it enabled us to perform a comparative assessment of autonomic control between normal subjects and patients with OSAS across different sleep-wake states. Application of CPAP also ensured that all individuals could be studied under relatively similar patterns of respiration in stable stages of REM and NREM sleep. In contrast, previous investigations of autonomic control in OSAS during sleep were carried out under uncontrolled conditions, in which the episodes of obstructive apnea were associated with profound swings in sympathetic and parasympathetic activity as well as transient state changes (19, 39). Large swings in respiration and cardiovascular variables can lead to severe distortions of the HRV and blood pressure variability spectra and, thus, contribute to interpretational difficulties if these spectral measures are used for making inferences about autonomic function (21).

On the other hand, CPAP application in the OSAS patients during sleep may have led inevitably to some confounding influences. For example, in subjects with normal heart function, acute application of CPAP is known to decrease left ventricular preload more than left ventricular afterload, leading to a reduction in cardiac output (13). However, our OSAS patients did not show any change in mean SBP or DBP during sleep relative to wakefulness when only minimal CPAP was applied. Furthermore, CPAP is likely to have increased lung volume, which is known to increase vagal activity and reduce sympathetic drive (11). Therefore, our technique is likely to have underestimated the extent of autonomic abnormality in the OSAS patients during sleep.

Dynamics of the estimated model components. The forms of the estimated impulse responses corresponding to the three mechanisms appear to be consistent with the underlying physiology. The PSR impulse response, which represents the vagal feedback from PSR, shows a fast initial negative peak and a delay between V and RRI, indicating that pulmonary inflation increases heart rate reflexly but with some delay after the initiation of inspiration, as has been reported previously (37). The RCC impulse response, which represents the central component of RSA, shows a fast initial negative peak and a lead (negative delay) between heart rate and P_{mus}, reflecting an increase in heart rate during inspiratory effort. The ABR impulse response increases initially to positive values, reflecting a very rapid decrease in heart rate (or, equivalently, increase in RRI) in response to an increase in blood pressure. All three impulse responses showed time courses that decayed within 15 s. The rapidity of the dynamic behavior suggests that all three components of HRV are mediated primarily by the parasympathetic system.

Effects of OSAS and sleep state on HRV components. The main finding in this study is that, in wakefulness and sleep, RCC and ABR gains were reduced in patients with OSAS compared with normal subjects. The second major finding is that, in both subject groups, ABR and RCC gains increased from wakefulness to sleep. These results are consistent with the fact that sympathetic tone is chronically elevated in patients with OSAS (36). Elevated sympathetic tone, such as in patients with congestive heart failure (36) or postmyocardial infarction (15), is associated with diminished baroreflex function. Increased sympathetic tone induced by mental stress is also known to reduce the parasympathetic influence on heart rate regulation (9). For example, voluntary override of the spontaneous respiratory pattern generator by controlled breathing has been shown to decrease high-frequency power RRI substantially in normal subjects (29). On the other hand, our results also show that sleep leads to increases in RCC and ABR gains, consistent with the notion that parasympathetic tone becomes elevated during sleep (24).

In contrast, there were no significant differences in PSR gain between normal subjects and OSAS patients; nor did either group demonstrate any dependence of PSR gain on sleep-wake state. There was substantial interindividual variability within each group in the PSR gain estimates (see sizes of standard errors in Table 2), and this may have obscured any real intergroup differences. The relative contribution of PSR to overall HRV was also uninfluenced by subject group or sleep-wake state. These findings suggest that the PSR mechanism is little affected by differences in sympathetic tone resulting from disease or differences in parasympathetic tone arising from changes in sleep-wake state. Why this should be so remains unclear and requires further testing using a larger pool of subjects. A further observation is that, in OSAS patients and normal subjects, PSR contributes the least of all three mechanisms to overall HRV. Similarly, previous studies in humans have shown that the contribution of the lung vagal reflexes to RSA is small (2, 3, 35) but nonnegligible (38).

A potential limitation of this study is that although we categorized "moderate-to-severe" OSAS as a single group, disease severity covered a broad range of apnea-hypopnea index values (from 21 to 88). It could be argued that a portion of the interindividual variability found in the OSAS group might well have been related to differences in disease severity. On the other hand, two observations make this hypothesis unlikely. 1) The interindividual variability (as represented by the standard errors of the estimated parameters in Table 2) in the control group was as large as the variability in the OSAS group. 2) We computed linear correlation coefficients between each parameter and apnea-hypopnea index. No significant correlations were found.

Estimation of baroreflex gain: comparison with previous studies. In our previous study in which we employed a dual-input model, we also found that ABR gain was significantly lower in OSAS patients than in normal subjects across all states, but the difference was largest during sleep. These amplitudes increased almost twofold in sleep vs. wakefulness in normal subjects but were less affected by state changes in...
OSAS. The inclusion of a third input (i.e., $P_{\text{mus}}$) in the present model did not significantly alter our estimates of baroreflex gain from the same datasets. The similarity of our present and previous estimates of ABR gain validates our prior assumption that the test protocol produces input sequences (i.e., $P_{\text{mus}}, V$, and SBP time series) that are not strongly correlated with one another.

We previously showed that our model-based estimates of ABR gain correlate significantly with baroreflex sensitivity calculated using conventional methods, such as the “sequence technique” and the bolus technique (7). These conventional methods, however, do not take into account the contribution of respiration to the fluctuations in RRI. A recent study has shown that excluding the effect of respiration on RRI can lead to the introduction of a significant amount of bias into the computation of baroreflex sensitivity (23). In our model-based approach, the component of the RRI fluctuations linearly correlated to respiration is removed before the remaining values are related to SBP fluctuations for the estimation of baroreflex gain. As well, our model makes the more realistic assumption that baroreflex action is associated with dynamic characteristics (e.g., response time); the aforementioned pharmacological and sequence techniques do not take into account baroreflex dynamics.

In conclusion, the model proposed in this study allows for the independent quantitative characterization of the three main autonomically mediated mechanisms involved in control of heart rate: vagal feedback from the PSR, central coupling between the respiratory and cardioregulatory centers, and the ABR. This can be achieved by using the model to analyze the cardiovascular responses to a test protocol in which positive airway pressure is applied in the form of a randomized sequence of breaths. The test protocol is nonintrusive and is reasonably well tolerated during sleep. Our results show that the adverse autonomic effects of OSAS include impairment of baroreflex sensitivity and the central mechanism linking respiratory drive with heart rate regulation. However, the component of RSA that is mediated by lung vagal feedback remains intact in OSAS. In normal subjects and OSAS patients, this third mechanism of HRV is relatively unaffected by changes in sleep-wake state, whereas the other two components demonstrate an increase in gain during sleep. Future studies could include the extension of the model to incorporate other potentially important factors affecting HRV, such as the contributions of the cardiopulmonary receptors and the chemoreflexes. As well, a determination of the relative impairment of these different contributions to HRV in patients with less severe manifestations of the disease could yield important insights into the long-term development of autonomic abnormality in OSAS.

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