

Accurate Sound Localization in Reverberant Environments Is Mediated by Robust Encoding of Spatial Cues in the Auditory Midbrain

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SUMMARY

In reverberant environments, acoustic reflections interfere with the direct sound arriving at a listener's ears, distorting the spatial cues for sound localization. Yet, human listeners have little difficulty localizing sounds in most settings. Because reverberant energy builds up over time, the source location is represented relatively faithfully during the early portion of a sound, but this representation becomes increasingly degraded later in the stimulus. We show that the directional sensitivity of single neurons in the auditory midbrain of anesthetized cats follows a similar time course, although onset dominance in temporal response patterns results in more robust directional sensitivity than expected, suggesting a simple mechanism for improving directional sensitivity in reverberation. In parallel behavioral experiments, we demonstrate that human lateralization judgments are consistent with predictions from a population rate model decoding the observed midbrain responses, suggesting a subcortical origin for robust sound localization in reverberant environments.

INTRODUCTION

The ability to localize sound sources can be important for survival and facilitates the identification of target sounds in multisource environments (Darwin, 2008; Kidd et al., 2005; Shinn-Cunningham, 2008). The auditory scenes that we perceive unfold in environments full of surfaces like walls, trees, and rocks (Huisman and Attenborough, 1991; Sakai et al., 1998). When an acoustic wave emanating from a sound source strikes a boundary surface, a fraction of the energy is reflected. The reflected waves themselves generate second-order reflections, with the process repeating ad infinitum. The myriad of temporally overlapping reflections, perceived not as discrete echoes but as a single acoustic entity, is referred to as reverberation.

Reverberation poses a challenge to accurate sound localization. To estimate the location of a sound source with low-frequency energy, such as speech, human listeners rely principally on tiny interaural time differences (ITDs) that result from the separation of the ears on the head (Macpherson and Middlebrooks, 2002; Wightman and Kistler, 1992). In a reverberant environment, reflected acoustic waves reach the listener from all directions, interfering with the direct sound. Under such conditions, the ear-input signals become decorrelated (Beranek, 2004) and the instantaneous ITD fluctuates (Shinn-Cunningham and Kawakyu, 2003). Because reverberant energy builds up over time, the directional information contained in the ear-input signals has a characteristic time course, in that ITD cues represent the true source location relatively faithfully during the early portion of a sound, but become increasingly degraded later in the stimulus.

In principle, listeners could accurately localize sounds in reverberation by basing their judgments on the directional information in the uncorrupted onset of the signals reaching the ears. Although human listeners can robustly localize sound sources in moderate reverberation (Hartmann, 1983; Rakerd and Hartmann, 2005), localization accuracy degrades in stronger reverberation (Giguere and Abel, 1993; Rakerd and Hartmann, 2005; Shinn-Cunningham et al., 2005b), suggesting that listeners are not immune to the ongoing, corrupted directional cues. To date, no one has studied the directional sensitivity of auditory neurons using stimuli with realistic reverberation. Thus, the degree to which auditory neurons maintain robust directional sensitivity in reverberation is unknown.

ITDs are initially coded in the auditory pathway as differences in relative spike timing between auditory nerve fibers on the left and right sides of the head. These timing differences are transformed to a rate code in the medial superior olive (MSO), where morphologically and physiologically specialized neurons (Grothe and Sanes, 1994; Scott et al., 2005; Smith, 1995; Svirskis et al., 2004) perform coincidence detection on convergent input from both sides of the head (Goldberg and Brown, 1969; Yin and Chan, 1990). Theoretically, the average firing rate of these coincidence detectors is equivalent to a crosscorrelation of the input spike trains (Colburn, 1973).

The majority of neurophysiological studies of spatial processing have targeted the inferior colliculus (IC), the primary nucleus

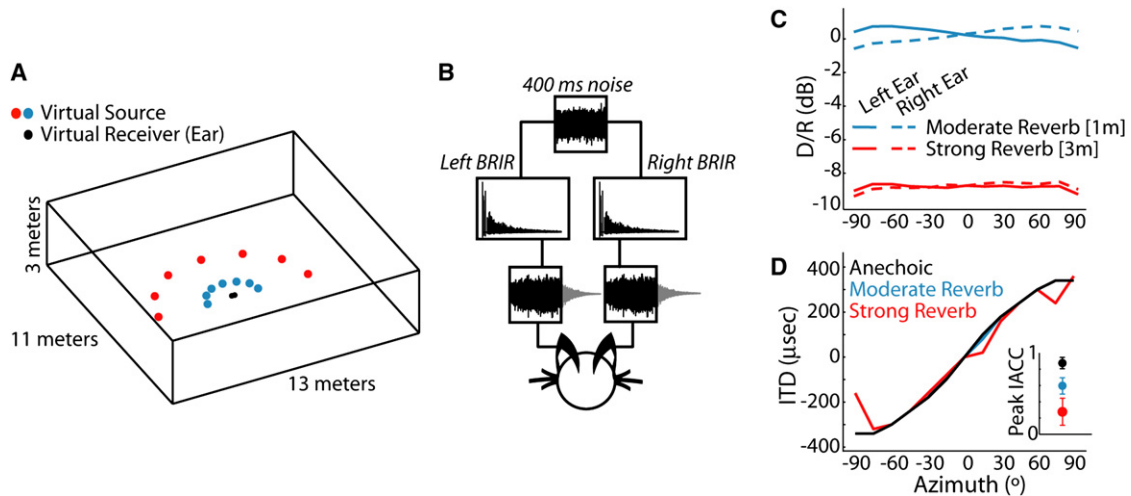


Figure 1. Properties of the Virtual Auditory Space Simulations

(A) Geometry of the virtual auditory environment. Reverberant binaural room impulse responses (BRIR) were simulated at two distances between source and receiver (1 m and 3 m). Anechoic (i.e., “no reverb”) BRIR were created by time-windowing the direct wavefront from the 1 m reverberant BRIR. (B) To simulate a sound source at a given azimuth, a reproducible 400 ms broadband noise burst is convolved with the left and right BRIR and presented to the experimental subject over headphones. (C) Direct to reverberant energy ratio (D/R) versus azimuth for reverberant BRIRs. (D) Broadband ITD versus azimuth for each room condition, estimated as the time delay corresponding to the peak normalized interaural correlation coefficient (IACC). Inset, peak IACC for each room condition. Error bars represent ± 1 SD across azimuths.

comprising the auditory midbrain (Aitkin et al., 1984; Delgutte et al., 1999; Joris, 2003; Kuwada et al., 1987; Kuwada and Yin, 1983; McAlpine et al., 2001; Rose et al., 1966; Stillman, 1971a; Yin et al., 1986). Multiple parallel sound-processing pathways in the auditory brainstem converge in the IC (Adams, 1979; Oliver et al., 1995), making it a site of complex synaptic integration. Despite this complexity, the rate responses of low-frequency, ITD-sensitive IC neurons to broadband signals with a static interaural delay resemble the responses of ITD-sensitive neurons in the MSO (Yin et al., 1986) and are well modeled as a crosscorrelation of the acoustic ear-input signals, after accounting for cochlear frequency filtering (Hancock and Delgutte, 2004; Yin et al., 1987).

Here, we investigate the effects of reverberation on the directional sensitivity of low-frequency ITD-sensitive IC neurons. Consistent with the buildup of reverberation in the acoustic inputs, we show that directional sensitivity is better near the onset of a reverberant stimulus and degrades over time, although directional sensitivity is more robust than predictions from a traditional crosscorrelation model of binaural processing that is insensitive to temporal dynamics in the reverberant sound stimuli. We further show that human lateralization judgments in reverberation are consistent with predictions from a population rate model for decoding the observed midbrain responses, suggesting that robust encoding of spatial cues in the auditory midbrain can account for human sound localization in reverberant environments.

RESULTS

Effects of Reverberation on Neural Azimuth Sensitivity

We used virtual auditory space simulation techniques (Figure 1; Experimental Procedures) to study the directional response properties of 36 low-frequency, ITD-sensitive neurons in the IC

of anesthetized cats. The virtual space stimuli simulated the acoustics of a medium-size room (e.g., a classroom) and were designed to contain only ITD cues, without any interaural level differences or spectral cues. Stimuli were synthesized for two distances between the sound source and the virtual ears (1 m and 3 m) in order to vary the amount of reverberation (“moderate” and “strong”). The ratio of direct to reverberant energy (D/R) decreased with increasing distance and was largely independent of azimuth for each distance simulated (Figure 1C). Reverberation did not systematically alter the broadband ITD, estimated as the time delay yielding the maximum normalized interaural correlation coefficient (IACC) between the left and right ear-input signals (Figure 1D). However, increasing reverberation did cause a systematic reduction in the peak IACC (Figure 1D, inset), indicating increasing dissimilarity in the ear-input waveforms.

Figures 2A–2C illustrates anechoic (i.e., “no reverb”) and reverberant rate-azimuth curves for three IC units. For anechoic stimuli (Figures 2A–2C, black curves), the shape of the rate-azimuth curve was determined by the unit’s sensitivity to ITD within the naturally occurring range (see Figures S1A and S1B available online), which corresponds to ± 360 μ s for our virtual space simulations for cats. In many neurons, the discharge rate increased monotonically with azimuth (Figures 2A and 2B), particularly in the sound field contralateral to the recording site, which corresponds to positive azimuths. Units with a nonmonotonic dependence of firing rate on azimuth (Figure 2C) generally peaked within the contralateral hemifield, consistent with the contralateral bias in the representation of ITD in the mammalian midbrain (Hancock and Delgutte, 2004; McAlpine et al., 2001; Yin et al., 1986).

In reverberation, there was an overall tendency for the range of firing rates across azimuths to decrease with increasing

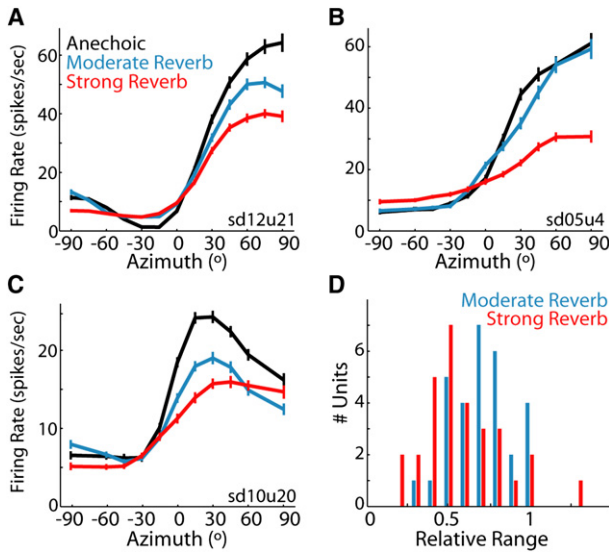


Figure 2. Reverberation Causes Compression of Neural Rate-Azimuth Curves

Anechoic and reverberant rate-azimuth curves (mean \pm 1 standard error) for three IC neurons with CFs of (A) 817 Hz, (B) 569 Hz, (C) 1196 Hz. (D) Population histogram of relative range for each D/R (moderate reverb, $n = 30$; strong reverb, $n = 30$).

reverberation, although the exact dependence varied across units. Typically, the effect of reverberation was graded (Figures 2A and 2C); however, there were units for which moderate reverberation had essentially no effect on the rate response (Figure 2B). Generally, the reduction in response range primarily resulted from a decrease in the peak firing rate; increases in minimum firing rates were less pronounced.

We quantified the overall compression of the rate-azimuth curves in reverberation using the *relative range*, which expresses the range of firing rates for a reverberant rate-azimuth curve as a fraction of the range of firing rates for that unit's anechoic rate-azimuth curve. In reverberation, the relative range is generally less than 1 (Figure 2D) and is significantly lower for the strong reverb than for the moderate reverb condition (paired t test, $p = 0.001$, $n = 24$). An information theoretic measure of directional sensitivity that is sensitive to the variability in spike counts as well as the mean firing rates showed a similar dependence on reverberation strength (Figure S2).

Reverberation could also alter the sharpness of azimuth tuning and—for units having a best ITD within the naturally occurring range—shift the best azimuth (Figure S3). However, changes in these tuning parameters occurred in either direction and were not consistently observed in all units. The most consistent effect of reverberation across our neural population was the compression of the response range.

Directional Sensitivity Is Better Near Stimulus Onset in Reverberation

Reverberant sounds have a characteristic temporal structure that is ignored when firing rates are averaged over the entire stimulus duration, as in Figure 2. At the onset of a sound in

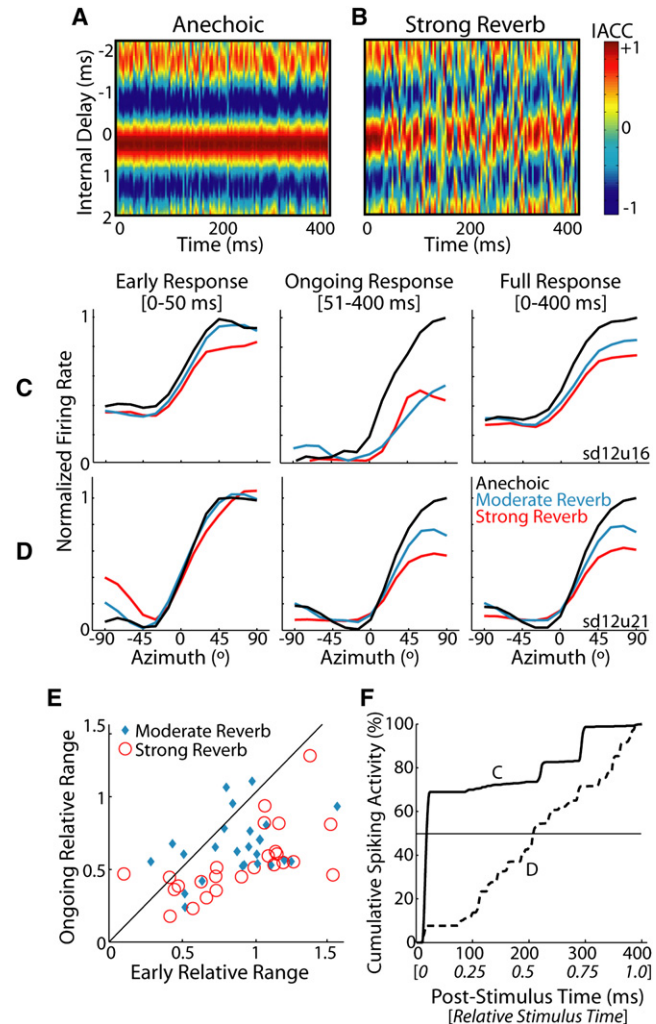


Figure 3. Temporal Dynamics of Directional Sensitivity in Reverberation

(A) Short-term IACC across time for the 45° anechoic virtual space stimulus; hot colors indicate high correlation. Ear-input signals were simulated as in Figure 1B and subsequently band-pass filtered (fourth-order Gammatone filter centered at 1000 Hz) to simulate peripheral auditory processing. Short-term IACC was computed using a sliding 4 ms window.

(B) Short-term IACC for the 45° strong reverb virtual space stimulus. (C and D) Rate-azimuth curves for two IC neurons computed using the early (0–50 ms), ongoing (51–400 ms), and full (0–400 ms) neural responses. To facilitate comparison across time periods, firing rates have been normalized to the maximum firing rate in the anechoic condition, separately for each time period. Unit CFs are (C) 747 Hz and (D) 817 Hz.

(E) Ongoing versus early relative range for IC neuron population. Solid line indicates identity i.e., $y = x$.

(F) Average cumulative peristimulus time histograms (cPSTHs) for the two neurons in panels (C) (solid line) and (D) (dashed line).

a reverberant environment, the energy reaching a listener's ears contains only the direct sound. Thus, the directional cues near the stimulus onset are similar for anechoic and reverberant virtual space stimuli (Figures 3A and 3B). As reverberation builds up over time, reflections increasingly interfere with the direct sound energy at a listener's ears and the directional cues for

the reverberant stimuli become more corrupted. Accordingly, we expected neural directional sensitivity to be better during the early, as opposed to the ongoing, portion of a sound stimulus in reverberation. Figures 3C and 3D shows rate-azimuth curves for two IC neurons computed from the early (0–50 ms), ongoing (51–400 ms), and full (0–400 ms) neural response. The rate-azimuth curves have been normalized to the maximum rate within each time period to facilitate comparison. Consistent with the buildup of reverberation, the rate-azimuth curves computed from the early response are similar across room conditions (Figures 3C and 3D, left), whereas substantial rate compression occurs for reverberant stimuli in the ongoing response (Figures 3C and 3D, middle). This trend holds across our sample of low-frequency, ITD-sensitive neurons (Figure 3E). Directional sensitivity in both moderate and strong reverberation is significantly higher during the early, as compared to the ongoing, neural response epoch (paired t test, *moderate reverb*: $p = 0.007$, $n = 24$; *strong reverb*: $p < 0.001$, $n = 25$).

Previous studies of ITD sensitivity in the mammalian IC have reported that neural onset responses show poorer ITD-tuning than ongoing neural responses (Geisler et al., 1969). Here, we have defined the “early” response epoch as the first 50 ms of the neural response, which is substantially longer than what is generally considered the “onset” response of a cell. Nonetheless, to prevent nondirectional early responses from biasing our results, we removed units that showed no significant change in early discharge rate across azimuth (Kruskal-Wallis test, $p > 0.05$); 6/36 units were removed from the statistical analysis and are not included in Figure 3E.

Role of Temporal Response Dynamics

The relative contribution of the early and ongoing responses to the directional sensitivity measured over the entire stimulus duration (Figures 3C and 3D, right) is determined by the distribution of spiking activity over the course of the stimulus. Many low-frequency, ITD-sensitive IC neurons exhibit spike rate adaptation in response to a sustained acoustic stimulus, such that firing rates are higher during the earlier portion of the stimulus and decrease over time (Ingham and McAlpine, 2004; Nuding et al., 1999; Rees et al., 1997; Stillman, 1971b). Such “onset dominance” in neural processing reduces the contribution of less-reliable ongoing reverberant stimulus energy to temporally-integrated measures of directional sensitivity.

Figure 3F shows anechoic cumulative peristimulus time histograms (cPSTHs; see Experimental Procedures) for the same two units as in Figures 3C and 3D. A unit with strong onset dominance (Figure 3F, solid line) has a cPSTH that rises rapidly shortly after stimulus onset. Accordingly, the full response for this unit is determined primarily by the early response (Figure 3C). In contrast, a unit that fires in a sustained manner throughout the stimulus has a more linear cPSTH (Figure 3F, dashed line); in this case, the full response exhibits a stronger resemblance to the ongoing neural response (Figure 3D).

To quantify onset dominance in single units, we computed T50—the time post-stimulus onset at which the cPSTH reaches 50% of its final value (Figure 3F). A strongly onset-dominated unit has a small T50 (Figure 3F, solid line), while a sustained unit has a T50 near the stimulus midpoint (Figure 3F, dashed line). Across

the neural population, the median T50 is significantly less than 0.5 (Wilcoxon signed-rank test, $p < 0.001$, $n = 36$), with the interquartile range spanning [0.31, 0.47]. This suggests that early directional responses typically contribute more to the overall directional sensitivity than the more-degraded ongoing directional responses.

If the response to a reverberant stimulus were governed primarily by neural response dynamics, we would expect onset-dominated units to show better directional sensitivity in reverberation than units with a sustained response. That is, we should observe a negative correlation between T50 and relative range. However, the correlation was not significant for either condition (*moderate reverb*: $p = 0.624$; *strong reverb*: $p = 0.517$), suggesting that other neural properties in addition to onset dominance influence directional sensitivity in reverberation.

Comparison to a Crosscorrelation Model

Previous investigations of low-frequency ITD-sensitive IC neurons have established that the rate response to interaurally delayed broadband noise is well-described by a crosscorrelation of the left and right ear-input signals, after accounting for peripheral frequency filtering and the nonlinear relationship between interaural correlation and firing rate (Hancock and Delgutte, 2004). Crosscorrelation models essentially reduce all binaural processing (including interaural delays) to a change in the effective IACC computed over the entire duration of the stimulus. In general, firing rate changes monotonically with IACC in low-frequency IC neurons, although there is substantial variability in the degree of nonlinearity in the relationship (Albeck and Konishi, 1995; Coffey et al., 2006; Shackleton et al., 2005).

In a reverberant environment, reflections interfere with the direct sound wave, resulting in decorrelation of the ear-input signals (Figure 1D, inset; see also Hartmann et al., 2005; Shinn-Cunningham et al., 2005a). According to the crosscorrelation model, this would *qualitatively* result in a compression of neural rate-azimuth curves, as observed in our neural data. We investigated whether a traditional crosscorrelation model could *quantitatively* account for the degradation of directional sensitivity in reverberation.

We used a modified version of the Hancock and Delgutte (2004) crosscorrelation model of ITD-sensitive IC neurons to generate predictions of reverberant rate-azimuth curves (see Experimental Procedures). The model is a cascade of linear peripheral frequency filtering and binaural crosscorrelation followed by a nonlinear transformation of IACC to firing rate (Figure 4A). The model parameters were fit for each individual unit using the rate-ITD and anechoic rate-azimuth data (Figure 4B), and then fixed to predict responses to reverberant stimuli.

Figures 4C–4E show model predictions of reverberant rate-azimuth curves for the same three IC units as in Figures 2A–2C. As expected, the model rate-azimuth curves are qualitatively similar to the measured reverberant rate-azimuth curves in that increasing reverberation causes more compression of the response. We quantified overall differences between observed and predicted directional sensitivity using the relative range (Figure 4F). Across the population, the model predicts substantial variability in the relative range, which originates from

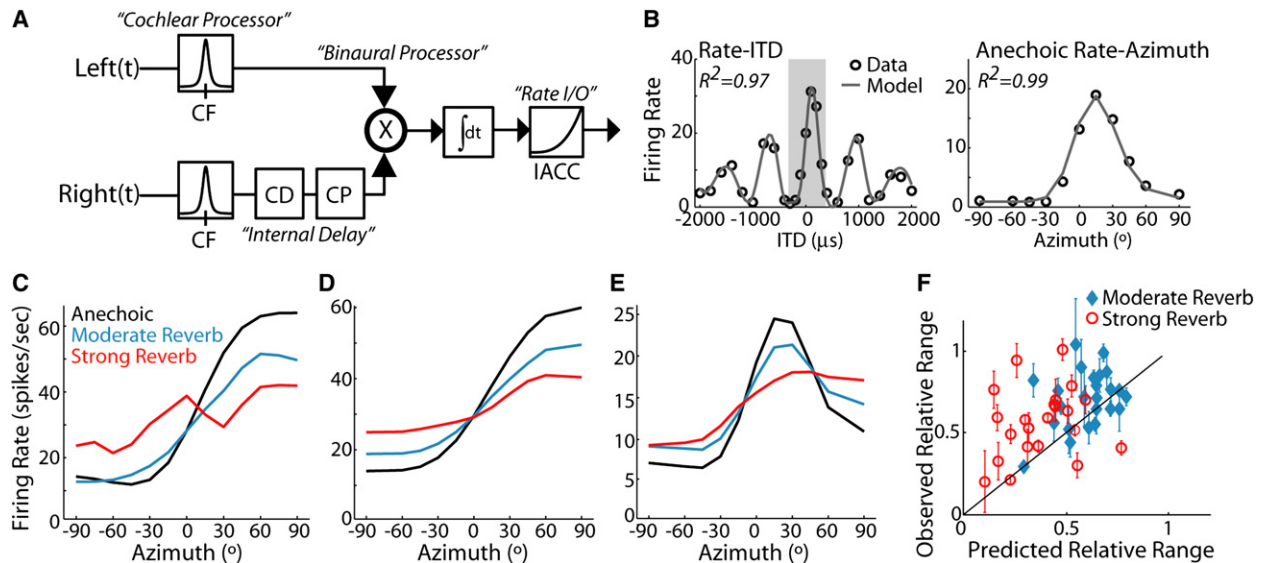


Figure 4. Average Effective IACC Poorly Predicts Directional Sensitivity in Reverberation

(A) Block diagram of the crosscorrelation model, after Hancock and Delgutte (2004). Left and right ear-input signals are band-pass filtered to simulate cochlear processing. Right-ear signal is internally delayed through a combination of pure time delay (CD) and phase shift (CP), and the resulting IACC is converted to firing rate using a power-law nonlinearity.

(B) Example model fits to the rate-ITD (left) and anechoic rate-azimuth (right) data for one IC unit (CF = 1312 Hz). The shaded region in the left panel delineates the range of ITDs corresponding to $\pm 90^\circ$ in the right panel.

(C–E) Model predictions of rate-azimuth curves for three IC neurons (same units as in Figures 2A–2C). For each neuron, model parameters were adjusted to minimize least-squared error between observed and predicted rate-ITD and anechoic rate-azimuth curves and subsequently fixed to generate predictions of reverberant rate-azimuth curves.

(F) Observed versus predicted relative range across the IC neuron population. Solid line indicates identity i.e., $y = x$. Error bars represent bootstrap estimates of ± 1 std. of relative range for observed responses.

variations in both frequency tuning and the nonlinear dependence of firing rate on IACC. Accurate model predictions for individual units would yield data points close to the identity line $y = x$ in Figure 4F; however, there is a great deal of spread in the data with no significant correlation between observed and predicted relative range for either reverberation condition (*moderate reverb*: $p = 0.174$, *strong reverb*: $p = 0.532$). Moreover, a majority of the data points fall above the identity line, indicating that observed directional sensitivity is generally more robust (i.e., better) than model predictions. For both reverberation conditions, predicted directional sensitivity is significantly worse than observed directional sensitivity (one-tailed paired t test, *moderate reverb*: $p = 0.02$, $n = 24$, *strong reverb*: $p = 0.005$, $n = 24$).

The crosscorrelation model is not sensitive to the exact time course of short-term IACC; rather, its output depends only on the IACC averaged over the entire stimulus. In contrast, we have shown that onset dominance in neural responses emphasizes the earlier segments of the stimulus which, in reverberation, contain less-degraded directional information. Such neural processing would effectively attenuate the contribution of ongoing reverberant stimulus energy to the IACC measured at the output of the integrator in Figure 4A. Thus, we hypothesized that neural onset dominance could account for the inability of the model to predict directional sensitivity in reverberation.

To test the hypothesis, we examined the relationship between T50 and crosscorrelation model error (defined as the difference between observed and predicted relative ranges, ΔRR). Positive

values of ΔRR indicate robustness to reverberation (i.e., the crosscorrelation model predicts more compression than was actually observed). Figure 5B shows a scatter plot of ΔRR versus T50; the filled symbols correspond to the cPSTHs plotted in Figure 5A. There is a significant negative correlation between the two metrics for both reverberation conditions (*moderate reverb*: $r = -0.534$, $p = 0.007$; *strong reverb*: $r = -0.612$, $p = 0.003$). Namely, units with smaller T50 (i.e., the most onset-dominated units) tend to be more robust to reverberation relative to model predictions than units with longer T50.

Despite the correlation, the substantial spread in the data suggests that onset-dominance cannot completely account for the inability of the crosscorrelation model to predict directional sensitivity in reverberation. The crosscorrelation model may be a poor predictor of directional sensitivity for stimuli with dynamic interaural time differences, in general (see Discussion). Nevertheless, these results suggest that onset dominance can improve directional sensitivity in reverberation.

Comparison to Human Psychophysics

We measured human behavioral lateralization of virtual space stimuli nearly identical to those used in the neurophysiology experiments. Because they contain only a single binaural cue, the virtual space targets are generally perceived on an internal interaural axis and are not externalized outside the head; hence, they are said to be lateralized instead of localized. Listeners adjusted the ILD of high-frequency narrowband noise until its

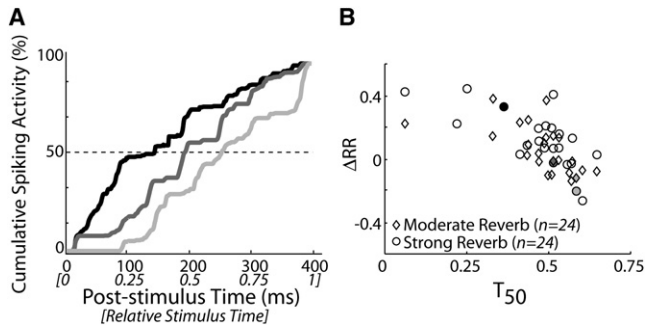


Figure 5. Onset Dominance Is Related to Robust Directional Sensitivity in Reverberation

(A) cPSTHs for three IC neurons with CFs of 150 Hz (black), 741 Hz (dark gray), and 1551 Hz (light gray). T50 is defined as the time at which the cPSTH reaches 50% of its final value (intersection of cPSTH with dashed line).

(B) Model prediction error (ΔRR) versus T50 across the IC neuron population, where positive ΔRR indicate robustness to reverberation. The two metrics are inversely correlated (*moderate reverb*, $p = 0.007$; *strong reverb*, $p = 0.003$). Shaded symbols correspond to the units shown in (A).

perceived laterality subjectively matched that of each virtual space stimulus. Because the absolute range of pointer ILDs for azimuths spanning $\pm 90^\circ$ varied from subject to subject, we normalized the subjective lateral positions to their maximum for each subject. Figure 6A shows the normalized subjective lateral position as a function of stimulus azimuth. For all conditions, mean lateralization judgments vary nearly monotonically with virtual source azimuth. Listener judgments of source laterality are similar for the anechoic and moderate reverberation conditions. However, in strong reverberation, the range of lateralization judgments is noticeably compressed. This compression of perceived laterality resembles the reduction in relative range measured in single IC neurons.

In order to directly compare neural responses to the behavioral results, we implemented a hemispheric-difference decoding model (Hancock, 2007; McAlpine et al., 2001; van Bergeijk, 1962) using the empirically measured rate-azimuth curves from our neurophysiology experiments. The model (Figure 6B, inset) estimates the lateral position of a sound source from the difference in the total activation between the two ICs. The choice of such a code (as opposed to a labeled line code, e.g., Jeffress [1948]) was motivated by the prevalence of monotonic rate-azimuth curves in our neural population, where a neuron's best ITD lies outside of the naturally occurring range of ITDs (Figures S1C and S1D).

The total population activity is computed for the ipsilateral IC by summing weighted rate-azimuth curves for all units in our sample of ITD-sensitive neurons. The weighting factors were used to adjust for slight differences between our empirical CF distribution and that found in a larger sample of low-frequency, ITD-sensitive IC neurons (Hancock and Delgutte, 2004). The weighting function was $w(CF) = P_{HD}(CF)/P_{Pres}(CF)$, where $P_{HD}(CF)$ is the lognormal distribution of CFs (with $\mu = 6.5$ and $\sigma = 0.31$) fit to the Hancock and Delgutte (2004) data, and $P_{Pres}(CF)$ is the empirical CF distribution in our population. Assuming symmetry with respect to the sagittal plane in the neural activation patterns produced by sound sources located on opposite sides of the

midline, the total population activity in the contralateral IC is derived by reflecting the ipsilateral population rate signal about the midline. The model output (*hemispheric difference signal*) is computed as the difference in population activity between the two ICs.

The main panel in Figure 6B shows the hemispheric difference signal for the anechoic and reverberant conditions. In all conditions, the hemispheric difference signal varies monotonically with stimulus azimuth. With increasing reverberation, the hemispheric difference signal becomes more compressed, as expected from the rate compression observed in individual units and consistent with the main trend in the behavioral responses. However, for both anechoic and reverberant conditions, the hemispheric difference signal saturates more quickly for lateral source positions than the human laterality judgments (see Discussion).

We quantified compression of the hemispheric difference signal using the relative range. Figure 6C shows the relative range of the hemispheric difference signal (open circles) plotted as a function of the decoder integration time i.e., the time interval from stimulus onset over which we averaged the individual neuron's firing rates to compute the hemispheric difference signal. The data are well fit by a single decaying exponential (solid curves). Because directional sensitivity is better during the earlier segment of a reverberant stimulus (Figure 3E), the relative range is initially close to 1 and decreases over time, consistent with the buildup of reverberant energy in the stimulus.

The symbols at the right of Figure 6C show the relative range of the lateralization estimates for individual human subjects. Both perceptual and decoder compressions show a similar dependence on reverberation strength. Quantitatively, the behavioral estimates show less compression than the hemispheric difference signal computed from the full neural response (0–400 ms), but more compression than that computed from only the early response (0–50 ms), suggesting that listener's lateralization judgments are influenced by late-arriving stimulus energy. To the extent that listeners integrate information over early and ongoing response segments, onset dominance may reduce the effective contribution of the ongoing population response.

DISCUSSION

Our neurophysiological results show that the directional sensitivity of ITD-sensitive auditory midbrain neurons degrades over the duration of a reverberant stimulus, consistent with the buildup of reflected sound energy at a listener's ears. We further find that onset dominance in temporal response patterns emphasizes the more reliable directional information in the early response, suggesting a role for this general feature of neural processing in improving directional sensitivity in reverberant environments. By comparing neural responses with human lateralization judgments, we find that the temporally integrated population rate response forms a possible neural substrate for robust sound localization in reverberation.

Dynamics of Directional Sensitivity in Reverberation

In a reverberant environment, reflections interfere with the direct sound arriving at a listener's ears, causing the ear-input signals

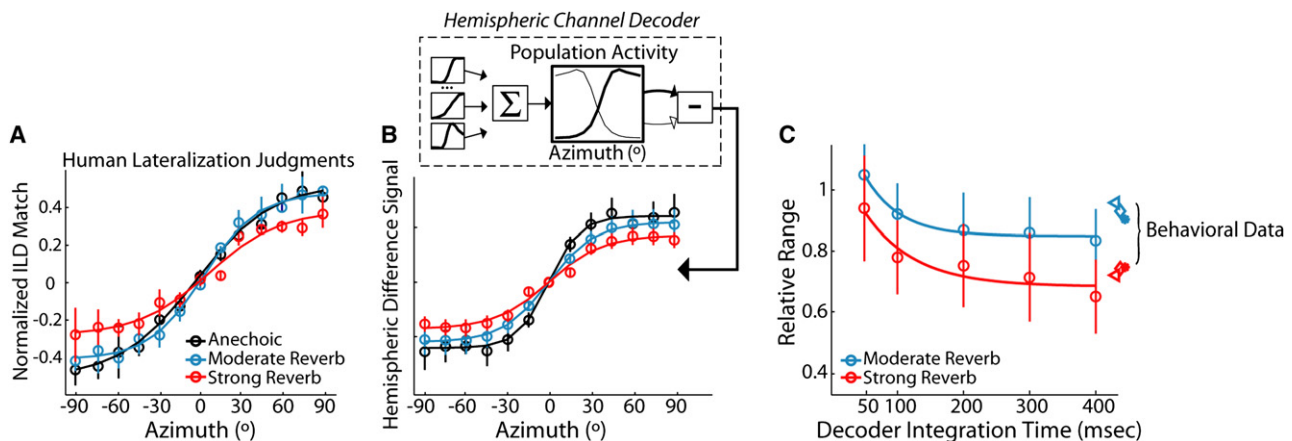


Figure 6. Hemispheric Decoding of IC Neural Responses Accounts for Lateralization Behavior of Human Listeners

(A) Human lateralization judgments. Across-subject ($n = 3$) mean (± 1 std) estimate of lateral position (i.e., normalized ILD-match) versus stimulus azimuth.

(B) Upper panel, schematic of the population decoding model (see text for description). Lower panel, hemispheric difference signal versus azimuth. Error bars indicate bootstrap estimates of ± 1 SD.

(C) Comparison of decoder and perceptual compression. Relative range of hemispheric difference signal (open circles) versus the time interval over which firing rate is integrated in the hemispheric decoding model; solid lines indicate fits by decaying exponential. Error bars represent bootstrap estimates of ± 1 SD. Relative range of human behavioral responses is plotted at the right edge of the panel (different symbols represent individual subjects).

to become decorrelated. Thus, it is not surprising that we observed a more severe degradation in directional sensitivity with increasing reverberation for both single neurons in the auditory midbrain (Figure 2D) and the crosscorrelation model (Figure 4). However, the directional information in reverberation has a characteristic time course: it is relatively uncorrupted near the sound onset, before the arrival of reflections at a listener's ears and becomes more degraded as reverberation builds up over time (Figures 3A and 3B). Our results show that neural directional sensitivity parallels this temporal pattern of cues in reverberation: Sensitivity is better during the early response than during the ongoing neural response (Figure 3E).

The overall directional sensitivity computed from the average rate response will depend on the distribution of spiking activity over time. Since directional information is better near the stimulus onset, a beneficial processing strategy would be to give proportionally more weight to the response near the onset of a stimulus. This could be achieved by any mechanism that reduces responsiveness in the later portions of the stimulus. A majority of neurons in our population exhibited onset dominance in their temporal response patterns, where firing rates are initially high and decay over time. When directional sensitivity is computed by integrating spike activity over time, onset dominance is a basic mechanism for emphasizing the earliest activity periods, when directional information is most reliable.

The sound stimulus used in the present experiments was a sustained noise, hence had a single onset. Many natural sounds, including human speech and animal vocalizations, are characterized by prominent amplitude modulations in the 3–7 Hz range (Houtgast and Steeneken, 1973; Singh and Theunissen, 2003), which functionally create multiple “onsets” over the duration of the stimulus. Indeed, the responses of IC neurons to sinusoidally amplitude modulated (SAM) sound stimuli typically show adaptation on every modulation cycle at low

modulation frequencies (Krishna and Semple, 2000; Nelson and Carney, 2007; Rees and Moller, 1983). While onsets in natural sounds are thought to be crucial for speech reception in reverberant rooms (Longworth-Reed et al., 2009), they may also provide a listener with multiple “onset-dominated” epochs over which to integrate directional information and make localization judgments (so long as the reverberation time does not exceed the period of dominant amplitude modulations in the stimulus).

Physiologically, onset-dominance in the IC could be realized through any of several neural mechanisms, including synaptic depression (Wu et al., 2002), intrinsic dynamics of active membrane channels (Sivaramakrishnan and Oliver, 2001), delayed, long-lasting inhibition (Kuwada et al., 1989; McAlpine and Palmer, 2002; Nelson and Erulkar, 1963; Pecka et al., 2007; Tan and Borst, 2007) or adaptation already present in the inputs to the IC (Smith and Zwillocki, 1975). The present physiological data do not allow us to discriminate among these possible mechanisms.

Relationship between Onset Dominance and Echo Suppression

The ability of a listener to localize sounds accurately in reverberant environments is often attributed to the precedence effect, a phenomenon in which the perceived source location is dominated by the initial portion of a stimulus (Litovsky et al., 1999). Numerous studies have reported neurophysiological correlates of classic precedence phenomena in the IC (Fitzpatrick et al., 1999; Litovsky and Yin, 1998; Pecka et al., 2007; Spitzer et al., 2004; Tollin et al., 2004; Yin, 1994). The stimuli used in these studies consisted of a leading source (representing the direct sound) followed by a lagging source (representing a single acoustic reflection). Because most of these studies used very brief stimuli, the leading and lagging sounds did not overlap in

time. Such conditions are an extreme oversimplification of realistic reverberation, in which thousands of reflections contribute to the energy at a listener's ears over hundreds of milliseconds.

Typically, neurophysiological studies of the precedence effect report that responses to the lagging sound are suppressed over a range of delays between the leading and lagging sounds, consistent with the dominance of the leading sound in the perceived location. The present result suggests that onset dominance in neural responses helps provide a robust representation of the location of sound sources in reverberation when the neural response is averaged over much longer times than the separation between individual reflections. While there is a superficial similarity between onset dominance and echo suppression, the two sets of results are not comparable because we cannot isolate the response to individual reflections as done in studies of the precedence effect.

A possible dissociation between neural echo suppression and onset dominance is suggested by the effects of anesthesia. The time course of recovery from neural echo suppression is faster in unanesthetized compared to anesthetized animals (c.f. Tollin et al., 2004; Litovsky and Yin, 1998). In contrast, ongoing experiments in our laboratory suggest that the effects of reverberation on azimuth sensitivity are comparable in the IC of awake rabbit and anesthetized cat (Devore and Delgutte, 2008). Moreover, the dynamics of spike-rate adaptation, a possible mechanism underlying onset dominance appear not to be strongly affected by anesthesia in the IC (Ter-Mikaelian et al., 2007).

While robust encoding of ITD in reverberation and neural suppression of discrete echoes each embody the seminal notion of the "law of the first wavefront" (Wallach et al., 1949), they operate on different time scales. In fact, onset dominance and neural echo suppression may contribute independently to robust encoding of azimuth in reverberant environments. The neural mechanisms underlying echo suppression in transient stimuli undoubtedly affect the neural response in the early portion of reverberant stimuli. However, there is likely an additional process, operating over longer time scales, that integrates directional information over time, emphasizing the early, reliable spatial cues over ongoing cues that are more degraded by reverberation.

Other Factors Influencing Directional Sensitivity in Reverberation

Qualitatively, the effect of reverberation on neural responses is consistent with a crosscorrelation model of binaural processing (Hancock and Delgutte, 2004; Yin et al., 1987), which predicts the average firing rate of IC neurons as a function of the effective IACC of the input signals (Figure 3A). However, a quantitative comparison reveals that the predicted reduction in directional sensitivity is not correlated with the observed reduction, indicating that the model does a poor job at predicting directional sensitivity in reverberation. Moreover, the observed reduction in directional sensitivity was generally less than the predicted reduction (Figure 4F), suggesting that additional mechanisms not included in the model provide neural robustness to reverberation. The difference between observed and predicted directional sensitivity was systematically related to onset dominance in neural temporal responses (Figure 5B); however, the relation

between onset dominance and model misprediction showed a lot of scatter, suggesting that additional factors beyond neural response dynamics play a role in the model's shortcoming.

The crosscorrelation model functionally reduces all processing of ear-input signals, including internal delay and reverberation, to changes in the effective interaural correlation. However, there is growing evidence that ITD-sensitive IC neurons receive convergent inputs from multiple brainstem coincidence detectors exhibiting different frequency and delay tuning (Fitzpatrick et al., 2000; McAlpine et al., 1998). Moreover, in addition to corrupting directional cues, reverberation also distorts the temporal envelopes of each ear-input signal. Temporal processing of stimulus envelope in the IC interacts with binaural processing in that manipulation of the stimulus envelope can cause changes in the firing rate of ITD-sensitive IC neurons even when IACC is unchanged (D'Angelo et al., 2003; Lane and Delgutte, 2005). Differences between model predictions and observed responses might be explained by differences between a single effective interaural correlation computation (as assumed in the model) and the actual computation performed by the IC cell on multiple inputs with different spectral, binaural, and temporal tuning characteristics.

Comparison to Psychophysics

The present results suggest that reverberation produces similar effects on the lateralization judgments of human listeners and on the directional sensitivity of IC neurons. A direct comparison of neural responses with human behavior requires explicit assumptions about how azimuth information is decoded from the rate responses of the neural population. Two basic classes of decoding models for sound lateralization have been analyzed: labeled-line models and hemispheric channel models. In labeled-line models (Fitzpatrick et al., 1997; Jeffress, 1948; Shackleton et al., 1992), the lateral position of a sound is determined by reading out the ITD corresponding to the centroid of activity in an array of neurons tuned to different ITDs. Such models require each tuned channel to transmit a label (i.e., the best ITD) to the decoder. In contrast, a hemispheric channel model determines the lateral position of a sound source by computing the difference of activity in two broadly tuned spatial channels, each representing subpopulations of neurons that preferentially respond to sound sources in one hemifield (Hancock, 2007; McAlpine et al., 2001; Stecker et al., 2005; van Bergeijk, 1962). Consistent with previous studies (Brand et al., 2002; Hancock and Delgutte, 2004; McAlpine et al., 2001), the majority of units in our population had monotonic rate-azimuth functions (Figure S1C), with best delays outside the naturally-occurring range and almost exclusively in the contralateral hemifield (Figure S1D), motivating our decision to implement a hemispheric channel decoder.

The model hemispheric difference signal was computed directly from the rate-azimuth curves measured in our sample of IC neurons. The range of the hemispheric difference signal decreased with increasing reverberation, mirroring the compression of human lateralization judgments (Figure 6C). Ideally, human listeners would use only the information at the onset of the stimulus to make the lateralization judgment and would therefore be minimally affected by reverberation. The fact that lateralization judgments do show compression suggests that

there may be an obligatory window of integration over which the lateral position is estimated. This possibility is intriguing, in that it suggests listeners may behave “suboptimally” given the available acoustic information. However, such behavior may be appropriate, considering that onset information can be unreliable due to masking by other sounds or internal noise. Thus, in everyday environments, optimal behavior may be to emphasize onsets, when detectable, but to also make use of ongoing information in case no onset information is available. Moreover, previous behavioral experiments have shown that human listeners are relatively insensitive to fast fluctuations in interaural correlation and appear to integrate binaural information over tens to hundreds of milliseconds when judging source direction (Grantham and Wightman, 1978). Psychophysical estimates of the length of the so-called “binaural temporal window” generally fall in the vicinity of 100 ms (Boehnke et al., 2002; Kollmeier and Gilkey, 1990). When we compared the human lateralization judgments to the hemispheric difference signal computed with different integration times (Figure 6C), we found that decoder compression best matches perceptual compression for an integration window of 100–200 ms. To the extent that lateralization judgments result from the integration of population rate responses over time, onset dominance will emphasize the early stimulus segments during this integration, as was shown for individual units (Figure 3C).

The azimuth dependence of the hemispheric difference signal was shallower at lateral azimuths than that of the human lateralization judgments (c.f. Figures 6A and 6B). However, this result is very sensitive to model assumptions including the exact distribution of CFs and best ITDs, as well as the mapping between azimuth and ITD. Moreover, species differences may also play a role since we are comparing human psychophysical data with model predictions based on cat neural data.

Hemispheric channel models have been criticized due to the lack of anatomical and physiological evidence for this type of operation, with simpler, single hemisphere rate codes offered as an alternative (Joris and Yin, 2007). With the present data, we found that the inter-channel comparison was necessary to avoid nonmonotonic responses at the most lateral source positions in the population rate response of each IC.

Theoretically, Jeffress-type models become more computationally powerful for animals with larger head sizes, including humans, while hemispheric decoding models work best for smaller animals such as cats (Harper and McAlpine, 2004). Because our neural data were not amenable to a straightforward implementation of a Jeffress-type decoding model for sound localization, we cannot say whether labeled-line models can explain lateralization performance in reverberation. However, our results show that hemispheric decoding models can indeed account for human lateralization in reverberant environments.

Conclusions

Our results show that reverberation degrades directional sensitivity both in single neurons and human listeners alike. Neural directional sensitivity is better during the earlier stimulus segments, when the signals at a listener’s ears are more reliable and less corrupted by reverberation. To the extent that listeners integrate directional information over time in estimating the posi-

tion of a sound source, we have shown that onset dominance in neural responses enhances spatial cues that are most reliable, resulting in more robust estimates of source position. Overall, our findings suggest that robust encoding of directional information in the rate responses of subcortical auditory neurons is sufficient to account for the lateralization performance of human listeners.

EXPERIMENTAL PROCEDURES

Surgical Preparation

Healthy, adult cats were anesthetized with dial-in-urethane (75 mg/kg, i.p.) and prepared for acute single-unit recording from the auditory midbrain using surgical procedures described in Hancock and Delgutte (2004). All surgical and experimental procedures were approved by the Institute Animal Use and Care Committees at both the Massachusetts Eye and Ear Infirmary and the Massachusetts Institute of Technology.

Virtual Space Stimuli

Binaural room impulse responses (BRIRs) were simulated using the room-image method (Allen and Berkley, 1979; Shinn-Cunningham et al., 2001) for a pair of receivers separated by 12 cm slightly displaced from the center of a virtual room measuring 11 × 13 × 3 m (Figure 1A). The interreceiver distance was chosen so that the range of ITDs in the direct sound spanned the range typically experienced by cats ($\pm 360 \mu\text{s}$, Figure 1D). Because we did not include a model of the cat head in the simulations, the resulting BRIRs contained ITD but essentially no interaural level difference (ILD) cues. BRIRs were calculated for azimuths spanning the frontal hemifield (-90° to $+90^\circ$) at distances of 1 and 3 m with respect to the midpoint of the receivers. Anechoic impulse responses were created by time-windowing the direct sound from the 1 m reverberant BRIRs. Virtual auditory space stimuli were created by convolving the BRIRs with a 400 ms burst of exactly reproducible Gaussian broadband noise gated with 4 ms \sin^2 ramps (Figure 1B).

Single-Unit Recordings

Experimental procedures for recording activity from single units in the auditory midbrain were as described in Hancock and Delgutte (2004). When a single unit was isolated, we estimated its characteristic frequency (CF) using an automatic tracking procedure (Kiang and Moxon, 1974) and then determined the intensity threshold for diotic broadband noise. ITD-sensitivity for 200 ms broadband noise bursts ($2/s \times 10$ repeats) was characterized at ~ 15 dB above threshold. Typically, ITD was varied between $\pm 2000 \mu\text{s}$ in $200 \mu\text{s}$ steps. Only ITD-sensitive units with low CFs (< 2.5 kHz) were further studied with the virtual space stimuli, with responses for each of the three room conditions obtained in pseudorandom order ($1/s \times 16$ repeats at each azimuth).

Data Analysis

A rate-azimuth curve for each room condition was computed by averaging the number of spikes that occurred in a fixed temporal window, defined relative to stimulus onset, across all trials for each azimuth. Rate-azimuth curves were smoothed using a three point triangular smoothing filter having weights [1/6 2/3 1/6]. We computed average cumulative peristimulus time histograms (cPSTH) for each unit to obtain a metric of onset dominance in the response. Each 1 ms bin in the cPSTH represents the cumulative number of spikes up to the bin time in the anechoic PSTH. The cPSTH was computed over a 400 ms duration, with time zero corresponding to the first bin in the anechoic PSTH having an across-trial spike count distribution significantly different from that of spontaneous activity. Only azimuths that evoked mean firing rates $\geq 90\%$ of the maximum rate across all azimuths were included in the average cPSTH in order to avoid including onset responses that often occur at unfavorable azimuths.

Single-Neuron Crosscorrelation Model

We used a crosscorrelation model to predict reverberant rate-azimuth curves of IC units. For each unit, we fit the rate-ITD curve with a modified version of the

Hancock and Delgutte (2004) crosscorrelation model (Figure 4A). The original model used a parabolic function to transform IACC into firing rate. We modified this transformation to be a power function of the form:

$$R(\rho) = a \cdot \left(\frac{1 + \rho}{2} \right)^p + b,$$

where a , b , and p are free parameters (Coffey et al., 2006; Shackleton et al., 2005). This modification improved the model fits (as evaluated using R^2).

To predict neural responses in reverberation, we first fit the six-parameter model to each unit's rate-ITD curve using the *lsqnonlin* function in Matlab (The Mathworks, Natick, MA). We then refit the scaling parameters (a and b) to the anechoic rate-azimuth function (to compensate for differences in the duty cycle with which the measurements were made). Finally, we generated predictions of reverberant rate-azimuth curves by running the model with the appropriate virtual space stimuli as inputs. We only included units for which the goodness-of-fit (R^2) for both rate-ITD and anechoic rate-azimuth data was at least 0.75 (8/36 units excluded).

Psychophysics

Four paid human subjects with normal hearing participated in the behavioral experiment. One of the four subjects failed the preliminary training procedure and was dismissed from the experiment. Experimental procedures were approved by the Boston University Charles River Campus Institutional Review Board.

Stimuli

BRIRs were created using the same methods and room characteristics as in the physiology experiments, except that the receivers were separated by 23 cm to achieve ITDs spanning the range typically encountered by a human ($\pm 690 \mu\text{s}$). Virtual space stimuli were created by convolving the BRIRs with random 400 ms Gaussian low-pass noise bursts (fourth order Butterworth filter with 2500 Hz cutoff) with 4 ms \sin^2 ramps.

Task

We used an acoustic pointing task to obtain a quantitative measure of stimulus laterality using the method of Best et al. (2007). Briefly, subjects adjusted the ILD of an acoustic pointer (200 Hz band noise centered at 3.0 kHz) until its perceived laterality matched that of a virtual space target. On each trial, the initial pointer ILD was randomly chosen from ± 20 dB. The target and pointer were then played in alternation (500 ms interstimulus interval) until the subject indicated a match with a button press.

Data Analysis

We computed the mean ILD-match at each azimuth, for each condition, after rejecting outlying trials (defined as estimates more than ± 3 standard deviations from the mean). We then fit sigmoid functions (using *lsqnonlin* in Matlab) to the individual subject responses and computed statistics using the fitted functions.

SUPPLEMENTAL DATA

The Supplemental Data include three figures and Supplemental Experimental Procedures and can be found with this article online at [http://www.neuron.org/supplemental/S0896-6273\(09\)00163-9](http://www.neuron.org/supplemental/S0896-6273(09)00163-9).

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