Sound sensitivity of neurons in rat hippocampus during performance of a sound-guided task

Pavel M. Itskov,1* Ekaterina Vinnik,1* Christian Honey,2 Jan Schnupp,2 and Mathew E. Diamond1

Cognitive Neuroscience Sector, Scuola Internazionale Superiore di Studi Avanzati, Trieste, Italy; and 2Auditory Neuroscience Group, Department of Physiology, Anatomy and Genetics, University of Oxford, Oxford, United Kingdom

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MATERIALS AND METHODS

Ethics statement. All experiments were conducted in accordance with National Institutes of Health, international, and institutional standards for the care and use of animals in research and were approved by the Bioethics Committee of the International School for Advanced Studies (permit n.5575-III/14) and supervised by a consulting veterinarian.

Subjects. Three Wistar male rats weighing about 350 g were housed together and maintained on a 14:10-h light-dark cycle. To ensure that the animals did not suffer dehydration as a consequence of water restriction, they were allowed to continue the behavioral testing to satiation (250–350 trials per session) and were given access ad libitum to drinking water for 1 h after the end of each training session. The animals’ body weight and general state of health was monitored throughout.

Stimuli. Artificial vowels were chosen as a more “naturalistic” class of stimuli than pure tones. Artificial vowels are a simplified version of vocalization sounds used by many species of mammals and have been used in studies of the ascending auditory pathway from the auditory nerve (Cariani and Delgutte 1996; Holmberg and Hemmert 2004) to the auditory cortex (Bizley et al. 2009, 2010). The spectra of natural

HIPPOCAMPAL NEURONS REPRESENT the animal’s location in the environment (Leutgeb et al. 2004, 2005a, 2005b; O’Keefe and Burgess 1996; O’Keefe and Nadel 1978; Wills et al. 2005; Wilson and McNaughton 1993). A growing body of work documents that hippocampal neurons also represent individual sensory stimuli and other nonspatial events (Christian and Deadwyler 1986; Eichenbaum et al. 1987; Hampson et al. 1999, 2004; Ho et al. 2011; Hok et al. 2007; Itskov et al. 2011; Komorowski et al. 2009; Lenck-Santini et al. 2008; Manns and Eichenbaum 2009; Moita et al. 2003; Wiener et al. 1989; Wood et al. 1999). Since hippocampus contains a prominent representation of location in space, one can specify and quantify a stimulus representation only after teasing it apart from head and body position-dependent responses associated with the stimulus-guided behavior. Despite the body of work demonstrating the representation of nonspatial, sensory stimuli in the hippocampus, few studies have focused on the distinction between spatial representation and stimulus representation.

We recently demonstrated that hippocampus contains a robust representation of tactile stimuli: many neurons fired at different rates when the rat encountered surfaces of different roughness even when the action associated with those stimuli was identical (Itskov et al. 2011). Still, in separate locations within the same experimental room, independent populations of neurons responded to the same tactile stimuli, indicating that responses to the tactile stimuli were modulated by the location of the animal. We also found that the explicit behavior of the animal (the direction of turn associated with reward) and tactile stimuli were represented by independent populations of neurons. The goal of the current work is to confirm and extend the previous results from tactile stimuli to sound stimuli and to further elaborate on underlying principles governing the neuronal representation of sounds in the hippocampus.

We therefore trained animals to associate four different sounds with two possible response directions. Because the behavioral responses to the two different stimuli associated with the same response direction were identical, any difference in the neuronal activity evoked by these two stimuli must reflect the coding of stimulus identity, rather than some aspect of behavior. Our experimental design allowed us to “isolate” stimulus-specific auditory responses and to examine how such sensory responses depend on spatial and behavioral context.

ADDRESS FOR REPRINT REQUESTS AND OTHER CORRESPONDENCE: P. Itskov, Department of Physiology, Anatomy and Genetics, University of Oxford, Oxford, United Kingdom

* P. M. Itskov and E. Vinnik contributed equally to this work.

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vowels are characterized by “formant” peaks that result from resonances in the vocal tract of the vocalizing animal (Schnupp et al. 2011). Formant peaks therefore carry information about both the size and the configuration of the vocal tract, and human listeners readily categorize vowels according to vowel type (e.g., /a/ vs. /o/) (Peterson and Barney 1952), as well as according to speaker type (e.g., male vs. female voice) or speaker identity (Gelfer and Mikos 2005). Many species of animals, including rats (Eriksen and Villa 2006), chinchillas (Burdi and Miller 1975), cats (Dewson 1964), monkeys (Kuhl 1991), and many bird species (Dooling and Brown 1990; Klunder et al. 1987) readily learn to discriminate synthetic vowels.

In the present study we used single formant artificial vowels, which were generated using binary click trains at the fundamental frequency of 152 Hz and 200-ms duration and were bandpass filtered with a bandwidth equal to 1/50th of the formant frequency using Malcolm Slaney’s Auditory Toolbox (http://cobweb.ecn.purdue.edu/malcolm/). Formant center frequencies were 1,520, 2,280, 3,420, 5,130 Hz, respectively. The spectra of all four stimuli are shown in Fig. 1B. The formant frequencies chosen here lie below the point of maximum sensitivity in the rat’s audiogram (Heffner et al. 1994), but previous work (Eriksson and Villa 2006) has shown that rats can easily learn to categorize vowels that are distinguished by formant peaks entirely below 4 kHz. The stimuli were ramped on and off using a Hanning window and played three times in succession (see Fig. 1A; sound stimuli are in Supplemental Files S1–S4) such that the stimuli presented to the animals were 600 ms long. (Supplemental data for this article is available online at the Journal of Neurophysiology website.) The stimuli were presented at a sound level of 70 dB SPL simultaneously from two speakers located symmetrically on the both sides of the platforms, so the sound level at the position of the animal’s ears was equal when the animal was in the “nose poke” position at the start of the trial (see Fig. 1D).

For the training and for analysis of neuronal responses, it was crucial for the sounds to be easily distinguishable. For this reason, we ensured that the distance between formant peaks (0.585 octave) was well above the frequency difference at which rats reach plateau performance in tone discrimination (∼0.07 octave; Sloan et al. 2009). Sounds were presented through Visaton FRS 8 speakers, which have a flat frequency response (less than ± 2 dB) between 200 Hz and 10 kHz. The speakers were driven through a standard personal computer sound card controlled by LabView (National Instruments, Austin TX).

**Apparatus.** The arena consisted of two platforms, each with a layout as illustrated in Fig. 1C. The platforms were elevated 40 cm to encourage the animals to remain on the platform during the experiment. A small wall with a nose-poke hole was positioned at one end of each platform. Reward water spouts were located on either side of the nose poke, 3.5 cm from the lateral platform edge on the level of the platform. The two platforms faced each other with their nose-poke walls at a distance of 25 cm. The two platforms differed only by the texture of the floor and their spatial position and orientation. The whole setup was in a Faraday room and was illuminated with infrared light (880 nm) during testing sessions.

Auditory categorization task. The day before behavioral testing, animals were placed on a water-restricted dry food diet. They were trained to initiate each trial with a nose poke. This triggered the presentation of one of the four stimuli. Different sounds were presented in random order. The animals did not have to wait for the end of a stimulus presentation but were free to respond at any time after stimulus onset. Right turns after presentation of the 1,520- or 3,420-Hz formants were considered correct responses, as were left turns after presentation of 2,800 or 5,130-Hz formants. Correct responses triggered delivery of a small water reward (∼50 µl) from the spout on the corresponding side. Incorrect responses triggered no water delivery but caused a 3-s timeout during which no new trial could be initiated. Each session lasted until the rat stopped doing the task.

The reaction times (from sound onset to the approach to the reward site) were measured. Since our setup required the animal to move just 13 cm from nose poke to reward spout, reaction times are a better measure of behavior than video tracking, where errors can exceed ± 2 cm (Jensen and Lisman 2000). One key aim of this study was to dissociate neuronal coding of stimulus identity from any aspect of explicit behavior (either ongoing or planned), for example, when the animals’ response causes them to move into or out of hippocampal place fields. We therefore considered a neuron to be sensitive to the acoustic stimuli only if its response patterns differentiated cases where the stimuli differed even though the animal’s behavioral responses were the same, i.e., it turned in the same direction, and did so with comparable speed, for both stimuli. By disregarding all the sessions in which the reaction times for the two stimuli associated with either response direction on either platform differed significantly (P < 0.05, Wilcoxon rank sum test, 14% of cases), we excluded the possibility...
that any systematic variability in the animals’ motor response, which might be differentially encoded by hippocampal place cells, could be misinterpreted as stimulus-related neuronal activity. Before and after the training session, the room was illuminated by visible light, so the rats had the opportunity to use visual cues to orient themselves relative to the layout of the experimental chamber and apparatus. S

Surgery and electrophysiological recordings. Once the rats reached ≥70% correct performance on both platforms, they were implanted with chronic recording electrodes. To do so, they were anesthetized with a mixture of Zoletil (30 mg/kg) and xylazine (5 mg/kg) delivered intraperitoneally. A craniotomy was made above left dorsal hippocampus, centered 3.0 mm posterior to bregma and 2.5 mm lateral to the midline. A microdrive loaded with 12 tetrodes (25-μm diameter Pt/Ir wire; California Fine Wire) was mounted over the craniotomy. Tetrodes were positioned perpendicular to the brain surface and individually advanced in small steps of 40–80 μm per day until they reached the CA1 area, indicated by the amplitude and the shape of the sharp wave/ripples. All recordings were performed at a depth of 2,000–2,500 μm below the brain surface. After surgery, animals were given antibiotic (Baytril; 5 mg/kg delivered through the water bottle) and the analgesic caprofen (Rimadyl; 2.5 mg/kg by subcutaneous injection) for postoperative analgesia and prophylaxis against infection, and they were allowed to recover for 1 wk to 10 days after surgery before testing resumed.

After recovery from electrode implantation, the animals were retrained for one to two sessions to get them accustomed to performing the task with a microdrive and cable attached. They then performed regular behavioral tests with the categorization paradigm described above, during which neuronal responses were recorded from the tetrode array using Neuralynx data acquisition equipment (analog-wired Cheetah acquisition system). Tetrodes had impedances between 100 and 500 kΩ. We usually recorded between one and six single units per tetrode. Spikes were presorted automatically using KlustaKwik, with the use of waveform energy on each of the four channels of the tetrode as coordinates in a four-dimensional feature space. The result of the automatic clustering was inspected visually after the data were imported into MCCLust. Only well-isolated single units that exhibited a clear refractory period were included in further analysis. In addition, we used two measures of spike separation, the L-ratio and the isolation index, to define the separation between the clusters in the multidimensional space (Schmitzer-Torbert et al. 2005). In the vast majority of cases, each single unit was recorded only in one session. In rare cases, the same unit could be distinguished on 2 consecutive days (judged by the spike waveform, distribution of the interspike intervals, and the response pattern), in which case only the single session with better isolation quality was considered in the analysis.

Statistical procedures. Firing rates usually varied over time, and the latency of the response varied across individual cells. Therefore, to quantify the effect of sound identity (or response direction) on neuronal activity, we used a method that would I) take into account multiple time points, 2) provide a single measure of statistical significance of the effect, and 3) quantify the strength of the effect in each case. From among the many statistical methods that could be applied (ANOVA, classification techniques such as linear discriminant analysis, etc.), we opted to use information theory measures because they provide a uniform scale that can be used to compare results from different experiments, their properties and biases have been extensively studied analytically, and robust bias correction procedures have been described and successfully implemented (Panzeri and Treves 1996; Panzeri et al. 2007; Pola et al. 2003).

Shannon mutual information (Shannon 1948) was used to quantify the statistical interdependence between trial parameter X (either stimulus identity or response direction) and neuronal spike count in a given temporal window. The information measures were computed in Matlab (MathWorks) using the information breakdown toolbox (Magri et al. 2009). Mutual information is given by the following equation:

\[
I(X;Y) = \sum_{y \in Y} \sum_{x \in X} p(x,y) \log \left( \frac{p(x,y)}{p(x)p(y)} \right) \tag{1}
\]

This quantifies the reduction of uncertainty about the trial parameter (X) gained by a single-trial observation of the spike count (Y). The probabilities in the above formulas are not known a priori and must be estimated empirically from the observed frequencies in a limited number of trials N. Inevitable inaccuracies in the resulting estimates of response probabilities can lead to an upward bias in the estimate of mutual information (Panzeri and Treves 1996; Panzeri et al. 2007; Pola et al. 2003). An approximate expression for the bias has been formulated (Panzeri and Treves 1996) and can be subtracted from the information estimates calculated with Eq. 1 above, provided that N is at least two to four times greater than the number of different stimuli or behavioral responses (Pola et al. 2003). In our data, N depended on the number of trials performed by the animals in a particular session but was never less than 12 in any condition, whereas the number of classes was always 2 (corresponding to either the 2 response directions or the 2 different sound stimuli associated with 1 response direction). This allowed us to use the above-mentioned bias correction procedure (Panzeri and Treves 1996). The information carried by spike counts was calculated in a 200-ms-wide window sliding in steps of 25 ms along the entire duration of the trial, from the sound onset to 1 s after.

Since for each neuron we took into account many time points, multiple comparisons had to be accounted for. This was done using a permutation test, which tested the null hypothesis that the mean value of information obtained across all the time points could be expected by chance (see below).

Statistical significance of sound-related information. Sound-related information quantified how reliably a neuron’s firing rate could distinguish between the two stimuli associated with the same response direction. First, the information about sound identity was measured independently at every time point from the sound onset until 400 ms after the sound offset. Each integration window was 200 ms long; the step size was 25 ms, yielding 41 sequential information values. Next, to determine whether the mutual information carried by the neuron’s firing rate was statistically significant, we used a permutation test that compared the observed mean information value (averaged across the 1-s analysis window) to a distribution of simulated values obtained after sound labels had been randomly reassigned to neuronal responses and the mean mutual information for these shuffled data had been computed. This resampling, repeated 200 times, provided an estimate of the distribution of average information values expected by chance if sound stimuli had no systematic influence on neuronal firing. P values for the significance of sound-related information were then calculated by comparing the information value for the original data to a Gaussian curve fitted to the distribution of 200 information values obtained by resampling. Neurons that were found to carry significant information about stimulus identity for at least one of the possible reward locations are referred to as “sound-sensitive neurons.”

Meta-test for the significance of sound-related information in the neuronal population as a whole. Although the permutation test just described yielded small (highly significant) P values for many neurons in our data set, it was important to ask whether the number of significant outcomes observed across the population exceeded that expected from possible “false positives” that occur within any large sample. To ascertain this, we compared the distribution of sound information P values obtained across the entire data set (all 261 neurons) to a distribution of P values obtained from simulated neuronal responses created by 200 shuffles of the stimulus labels. We found that the distribution of observed P values was significantly smaller (Wilcoxon rank sum test, P = 0.0013) than that expected by chance. This means that the neuronal population carried a far stronger
sound identity signal than could be expected based on firing rate fluctuations unrelated to the stimuli.

**Statistical significance of response direction-related information.** Response direction is the side to which the rat turned to obtain water. When the “turn right” and the “turn left” trials are compared, any difference in firing rate could be due to the place field of the cell, head direction, or predictive coding of the direction of the turn. The statistical significance of such differences was measured in the same way as was done for sound identity-related differences in activity: at each time point, the firing rate in the trials with correct right turn was compared with the firing rate at the same time point in the trials with correct left turn. The information across all time points was averaged and compared with the distribution of average information values obtained by random shuffling of the trial labels. Neurons carrying a significant quantity of information about the animal’s response direction on at least one of the two test platforms are referred to as “response direction-sensitive neurons.”

In theory, the difference between right-turn and left-turn trials could also be due to sound-dependent responses: for instance, if a neuron strongly responds to sound 1 (right) but does not respond to sounds 2, 4 (left), or 3 (right), it could result in spurious significant response direction information. To exclude cases in which sound-related responses might cause an apparent response direction response, a neuron was classified as sensitive to response direction only if it showed a significant left vs. right firing difference for all four left vs. right stimulus pairings: sound 1 vs. sound 2, sound 1 vs. sound 4, sound 3 vs. sound 2, and sound 3 vs. sound 4.

**Context dependence of sound sensitivity.** The design of the experiment allowed us to test for unambiguous neuronal sensitivity to differences in sounds in four different conditions: discrimination of sounds 1 and 3, both of which directed the rat to the right water reward spout, and discrimination of sounds 2 and 4, both of which directed the rat to the left spout. Both discriminations were made on platforms 1 or 2. Consequently, one recorded neuron could exhibit sensitivity to sound identity (that is, carry a significant quantity of information about the sound stimulus) under a total of 0, 1, 2, 3, or 4 conditions. We compared the observed instances of sound sensitivity of each neuron to the number expected by chance if the sound sensitivity under each condition were independent.

The distribution expected by chance was calculated as follows. Under the assumption of independence, the number of sound-sensitive neurons associated with all four response directions is expected to equal \(Np^4\), where \(N\) is the number of neurons in the sample and \(p\) is the probability that a neuron will carry a significant quantity of information about sound at one of the four possible reward locations. Similarly, the number of neurons expected not to be sensitive associated with any of the four response directions equals \(N(1-p)^4\).

There are four different ways in which a neuron can be sensitive to sound associated with only one response direction (i.e., sensitive to the sound associated with the 1st, 2nd, 3rd, or 4th spout), and each of these possible outcomes has a probability of \(p(1-p)^3\), yielding an expected number of neurons sensitive at just one spout equal to \(4Np(1-p)^3\). Similarly, the expected number of neurons sensitive at three spouts equals \(4Np^3(1-p)\). Finally, there are six different ways in which a neuron can be sensitive to sound associated with all four response directions (i.e., sensitive to the sound associated with the 1st, 2nd, 3rd, or 4th spout), and each of these possible outcomes has a probability of \(p^4(1-p)^0\), yielding an expected number of neurons sensitive at all four spouts equal to \(6Np^4(1-p)^0\). Once the expected values were calculated in this manner, their confidence intervals were obtained from the critical values of the inverse binomial distribution (Matlab function “binoinv”).

**Separation and comparison of putative interneurons and pyramidal cells.** In keeping with established methods, three criteria were used to distinguish neuronal cell types according to physiological criteria: mean firing rate (Fox and Ranck 1981; Ranck 1973), spike duration (Skaggs et al. 1996), and the autocorrelation function (Csisvari et al. 1999). Firing rate was measured over the whole session. Spike duration was measured at 25% of maximum spike amplitude (Csisvari et al. 1999). The autocorrelation-derived index is the difference between the numbers of spikes that occurred in a 3- to 5-ms postspike window and a 20- to 80-ms postspike, divided by the sum of the two. Spike duration and the autocorrelation-derived index yielded bimodal distributions. On the basis of firing rate, spike duration, and the autocorrelation-derived index, the complete set of neurons was clustered into two classes using a \(K\)-means algorithm. In keeping with the conventions established by others (Csisvari et al. 1999; Fox and Ranck 1981; Wiebe and Staubli 2001), we refer to the units with the broader action potentials and lower firing rates as putative pyramidal cells and to those with the narrower spikes and higher firing rates as putative interneurons. We stress that a definitive classification of cell types requires information beyond that which can be obtained with extracellular recordings alone. We found that narrow spike-firing, putative interneurons exhibited a broader distribution of spike widths and firing rates than putative pyramidal cells. This may be attributable to the greater diversity of the interneurons compared with pyramidal cells (Csisvari et al. 1999; Klausberger and Somogyi 2008). For the two physiologically distinguished cell types, the statistical significance of the information about sound stimulus identity or...
response direction was determined using the permutation test described above. To quantify the amount of information carried by each neuron, we took the peak value of the information across time among those neurons that passed the significance test. These values were used to compare sound and response direction coding differences between the putative pyramidal cells and interneurons.

**Histology.** After conclusion of the recording experiments, the animals were overdosed with intraperitoneal injection of the anesthetic urethane and transcardially perfused with 10% formalin. Brains were sectioned in the coronal plane and stained with cresyl violet. Electrode tracks were localized on the serial sections (Fig. 2).

**RESULTS**

**Neuronal representation of sound identity.** In this study we set out to test how neurons in rat hippocampus represent auditory stimuli in the context of a behavioral task. Rats triggered a sound stimulus (Fig. 1, A and B) by extending their head through an infrared light barrier placed at the front of the platform (the “nose-poke” position, see Fig. 1C). When the light barrier was crossed, one of the four artificial vowel triplets (Fig. 1D) was played bilaterally. Depending on stimulus identity, animals had to turn either right or left to receive a water reward (Fig. 1E). Two sounds were associated with the reward on the right, and two other sounds with the reward on the left. Since the animals performed the same action for both sounds associated with the same reward spout, this design allowed us to look for stimulus-specific neuronal responses. The rats were initially trained on one of the two nearly identical training platforms until they performed at ≥70% correct for all four sounds. This took between 1.5 and 2 mo. Next, they were trained in the same task on the second platform with the same stimulus-response associations (in rat-centered coordinates). Although the behavioral task and sound stimuli were identical, the transfer of the task to the second platform was not immediate. About 3–5 days (~400 trials) of additional training were required for the animals to become competent, and for the remaining sessions, they persistently achieved lower performance on the second platform (median performance across all recording sessions: 80.2% on first platform and 70.6% on second platform; Wilcoxon sign rank test, P < 10⁻⁵). The lack of immediate transfer of reward contingencies suggests the animals may have formed two distinct representations of the task, one for each of the two training platforms. Alternatively, it may stem from the conflict between a room-centered representation of the arena and a self-centered representation.

To quantify the representation of sound identity, we measured the information carried by neuronal firing rate on correct trials (see MATERIALS AND METHODS for details). To ensure that any observed differences in firing for the two different sounds associated with each response direction reflected differences in sound, and not differences in behavior (see MATERIALS AND

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**Fig. 3.** Examples of sound-sensitive responses recorded from 4 neurons (A–D). Diagrams at left indicate on which platform and with which reward direction the activity was recorded. Top plots show mean firing rates (±SE) after stimulus onset. Responses to the 2 different sound stimuli (S1 or S3; S2 or S4) tested for each response direction are color coded as indicated. Middle plots show the responses as dot rasters (each dot corresponds to the time of a spike; sequential trials of 1 stimulus are plotted on separate rows). Bottom plots quantify the stimulus dependence of firing rate over time as bits of mutual information in 200-ms sliding time windows. Solid black lines indicate actual information quantity; red dashed lines represent average information across 200 permutations. Bin size = 200 ms, step = 25 ms.
METHODS for details), we excluded recording sessions in which reaction times differed between the two sounds tested at any of the four response directions (left or right on either platform 1 or 2). Accordingly, 261 of the total of 429 recorded neurons were included in further analysis.

Examples of responses from sound-sensitive neurons are shown in Fig. 3. A total of 55 of 261 tested neurons (21%) exhibited significant sensitivity to sound identity. The median quantity of information about sound identity carried by neurons with sound sensitivity was 0.12 bits (interquartile range 0.08–0.16 bits). The upper limit, in a discrimination between two sounds, would be 1.0 bit. A meta-test (see MATERIALS AND METHODS for details) confirmed that the observed sensitivity across the population was statistically significant (Wilcoxon rank sum test, \( P = 0.0013 \)).

To evaluate the latencies of sound-specific neuronal responses, we analyzed the distribution of the first bin after stimulus onset carrying a significant amount of information about sound identity \( (P < 0.05) \). Neurons started to carry sound information at a median latency of 310 ms after the sound onset (Fig. 4A). There were insufficient numbers of error trials to permit an analysis of whether the representation of specific sounds was equivalent on correct and incorrect trials.

**Passive sound exposure.** The use of sounds as stimuli allowed us to ask whether their representation in CA1 was related to the stimuli as mere physical signals or as meaningful events dependent on behavioral context. To date, it has not been possible to show unambiguously whether rat hippocampal neurons truly encode nonspatial sensory stimuli or whether they fire to some particular aspect of the behavior associated with the discrimination. To answer this question, we recorded neuronal activity just after the animals had completed the test session and were resting on the second platform. These data were obtained for 282 of the total of 429 cells in our data set. The animals were exposed to the same sounds as during the immediately preceding behavioral task. The sounds were triggered with no relation to the animal’s posture or movement, in a random sequence lasting \( \sim 15 \) min with a random interstimulus interval ranging from 0.8 to 1.5 s. The rats did not respond to these sounds with head turns toward the water spouts or any other overt behavior. We analyzed the neuronal discharges in relation to sounds by the same methods as when the sounds were presented during the discrimination behavior. Seventeen of 282 neurons (6%) recorded during the passive conditions were classified as sound sensitive at a \( P < 0.05 \) significance level, a number close to the expected “false alarm” rate of 5%. We compared the frequency of such responses across the population using the stimulus label shuffling meta-test described in MATERIALS AND METHODS and found no evidence for statistically significant sound coding across the population as a whole \( (P = 0.39) \). Moreover, among the neurons that did exceed the criterion, the median quantity of sound information was significantly smaller than the quantity found during the active task \( (0.05 \text{ bits in the passive condition vs. } 0.12 \text{ bits for the active task; Wilcoxon rank sum test, } P < 10^{-7}) \). By visual inspection, none of the 282 neurons recorded in the passive condition showed convincing, reproducible sound selectivity. Two examples of neurons that distinguished between two sounds within the behavioral task but not in the passive condition are shown in Fig. 5. Together, these results indicate that under our experimental conditions, neurons in the CA1 region of hippocampus were not sensitive to differences in acoustic stimuli presented in isolation, even if the neurons showed clear discrimination between the same acoustic stimuli in the context of sound-guided behavior.

**Neuronal representation of response direction.** A wealth of literature documents that many neurons in rat hippocampus are highly sensitive to the animals’ location in space, and one might therefore expect a number of neurons in our sample to be sensitive to whether the animal turned left or right to collect water reward. We tested for such sensitivity by applying the same mutual information measures used to test for stimulus coding (see MATERIALS AND METHODS for details). Examples of neurons that showed response direction-dependent firing on one of the two platforms only are shown in Fig. 6, A and B, whereas Fig. 6, C and D, shows examples of neurons that fired more during turns toward the reward spout on the left regardless of platform.

A total of 113 of 261 tested neurons (43%) exhibited significant sensitivity to response direction. Neurons with response direction sensitivity carried a median value of 0.22 bits of information about response direction (interquartile range 0.15–0.38 bits). Neurons that encoded response direction on one platform only (67 putative pyramidal cells and 24 putative interneurons) were more common than those that encoded response direction on both platforms (12 putative pyramidal cells and 10 putative interneurons). Neurons started to carry response direction information at a median latency of 350 ms after the sound onset (Fig. 4B).

Whenever there was a sufficient number of incorrect trials to characterize neuronal firing on those trials, the firing of the response direction-sensitive neurons was found to be associated with the animals’ actions on those trials. The simplest interpretation is that these neurons discriminated the actual direction of the movement rather than the stimulus category.

**Relationship of sensitivity across different platforms.** The fact that animals were tested on two nearly identical but differently located platforms enabled us to investigate whether sound-related information present in the neuronal discharges...
depended on the animal’s spatial location, and how coding for sound identity and place interacted. Our data indicate that in the large majority of the cases, sound-related information was observed only when the rat performed the task on one of the two platforms, suggesting that sound coding depended on spatial context.

The relationship between sound coding and spatial location was assessed as follows. When tested on two platforms, each with two response spouts, a neuron could in principle show sound sensitivity in its firing associated with each of the four possible response directions (i.e., during both left and right turns on each of the 2 platforms), or it might be sensitive to sound in association with only a subset of the four response directions or at none at all. We found that sensitivity to sound was linked to the position of the animal within the room, i.e., the neuronal responses to sounds on one platform could not be predicted based on the sensitivity of the same neuron on the other platform. Of the 261 neurons in our sample, 48 gave significant sound identity information for only one of the four response directions or at none at all. We found that sensitivity to sound was linked to the position of the animal within the room, i.e., the neuronal responses to sounds on one platform could not be predicted based on the sensitivity of the same neuron on the other platform. Of the 261 neurons in our sample, 48 gave significant sound identity information for only one of the four response directions, 7 gave significant information for two response directions, and none gave significant information for more than two response directions. The filled bars in Fig. 7A summarize these results. The open bars show the distribution that is predicted if one assumes that the presence in a given neuron of a sound identity code for any one of the four tested reward spout locations does not predict whether the same neuron will discriminate between sounds for other response directions. The observed distribution matches the distribution predicted by the model of independence.

Relationship of sensitivity to sound and response direction. In principle, a neuron could be sensitive to only response direction (left/right), only sound identity, or both. The sensitivity of the neuron to one of these factors may depend on the other factor, or, alternatively, sound sensitivity and response direction sensitivity could be independent properties. We derived a distribution of cell counts expected to occur in a population, given the assumption of this independence (see MATERIALS AND METHODS).

In the experiment represented in Fig. 7B, we investigated the relationship between sensitivity to sound stimulus identity and sensitivity to response direction across the sample of 261 neurons. A neuron was considered sound sensitive (S+) if in at least one of the four conditions (platform 1: sound 1 vs. sound 3; platform 1: sound 2 vs. sound 4; platform 2: sound 1 vs. sound 3; platform 2: sound 2 vs. sound 4) it carried a significant quantity of information about sound identity. Otherwise, it was considered sound insensitive (S−). Likewise, a neuron was considered sensitive to response direction (D+) if it distinguished left and right response direction on either platform. Otherwise, it was considered to be insensitive to re-

![Fig. 5. Examples of responses for 2 neurons sensitive to sound identity during the behavioral task but not during passive listening. Top plots show mean firing rates (±SE) after stimulus onset. Middle plots show the responses as dot rasters. Bottom plots quantify the stimulus dependence of the firing rate over time as bits of mutual information in 200-ms sliding time windows. Black lines indicate actual information quantity; red dashed lines represent average information across 200 permutations. Bin size = 200 ms, step = 25 ms. The first neuron fired more vigorously to S3 than to S1 during the discrimination task (A) but did not show significant sound coding when the same stimuli were presented passively (B). The second neuron was excited by the onsets of each of the sounds during the discrimination task (the triplets of bursts in C are aligned to the triplets of sounds) but showed no stimulus-related modulation during passive listening (D).](http://jn.physiology.org/identifier/doi:10.1152/jn.00404.2011)
neurons as putative pyramidal cells and 82 as putative interneurons (Csicsvari et al. 1999; Fox and Ranck 1981; Itskov et al. 2011; Ranck 1973). We then asked whether the putative cell type was related to functional properties of neurons in the behavioral task.

Putative interneurons were more likely to be sensitive to stimulus identity, and carried more sound-related information, than putative pyramidal cells. Among putative interneurons, 33% were sound sensitive, whereas only 16% of putative pyramidal cells were sound sensitive (27 of the 82 putative interneurons and 28 of the 179 putative pyramidal cells). To quantify information in these cells, we took a peak value of sound identity information. The median value of the peak among all sound-sensitive putative interneurons was 0.15 bits (interquartile range 0.11–0.19 bits), significantly higher than the value of 0.10 bits (interquartile range 0.07–0.13 bits) in putative pyramidal cells (Wilcoxon rank sum test, \( P < 0.0008 \)).

Seventy-nine of 179 (44%) putative pyramidal cells and 34 of 82 (41%) putative interneurons carried significant amounts of information about response direction. The median of the
The study was designed in a manner that can dissociate stimulus-related firing from firing related to behavioral responses or spatial location, by training animals to associate more than one sound stimulus with each behavioral response. Moreover, we required rats to carry out identical stimulus-reward pairing experiments at different locations in the room. We found that some CA1 neurons clearly distinguished between different acoustic stimuli, and because the actions associated with the paired sounds were identical, the sound sensitivity cannot be easily explained by any explicit behavioral correlate. Acoustic sensitivity was in most cases contingent on behavioral context and the animal’s location in space. Neurons that discriminated between a sound pair when the animal was on one platform frequently failed to discriminate between the same sound pair when the animal was on the other platform. As a further demonstration of the role of context, we found that during passive listening, neurons did not discriminate between sounds. The modulation of neural responses to the presented sounds that is commonly found in hippocampus thus appears to be much greater in extent and qualitatively different from that normally seen either in auditory cortex as a function of changes in behavioral context (Fritz et al. 2007) or in the midbrain as a function of spatial variables (Campbell et al. 2006).

A degree of dependence on spatial context of sensory responses in hippocampus has been observed previously. For example, studies by Wood et al. (1999) and Komorowski et al. (2009) reported that hippocampal responses to odontor stimuli often differed at different locations, although some neurons appeared to “prefer” certain particular odors regardless of location. Our experiment was not designed to map sound preferences across space, but the results nevertheless confirm and extend the previous findings that coding for sensory stimuli and coding for space appear to interact in hippocampus. In particular, a neuron’s tendency to differentiate two sound stimuli at one location in space did not predict whether the neuron would differentiate the same sounds at a different location.

Interactions between coding for sensory stimuli and coding for spatial location do not of course preclude the possibility that spatial or sensory coding may be achieved by somewhat separate, specialized subpopulations of hippocampal neurons.
A number of studies by Sakurai (1994, 1996, 2002) have explored hippocampal population responses during various sound-related tasks. Sakurai’s experimental and analytical approach differs substantially from ours, but our work nevertheless confirms and extends several of Sakurai’s findings. In one study (Sakurai 1996), animals were trained in three different tasks: auditory discrimination, visual discrimination, and a configural auditory plus visual discrimination. The populations of neurons that were engaged in each of these tasks were found to be partially overlapping. The author put forward the idea that these overlapping assemblies arose from independent coding. Independence here is to be understood in the statistical sense, meaning that the probability that any one neuron carries significant amounts of visual information is not influenced by whether or not the same neuron encodes auditory or spatial information. However, Sakurai did not demonstrate such statistical independence quantitatively. In comparison, our current experiment, as well as a previous experiment using tactile stimuli (Itskov et al. 2011), revealed distinct and statistically independent representations of sounds associated with different locations in the spatial arena (Fig. 7A) and also showed that neuronal populations involved in coding different aspects of the task, sound identity and the action associated with it, overlap in proportions consistent with statistical independence (Fig. 7B).

However, we observed some statistical dependence of coding for space or sound on cell type. At present, there is no consensus in the literature on the relative contributions made by putative hippocampal interneurons and putative pyramidal cells to the representation of auditory stimuli or the animal’s location in space. Whereas one study showed a clear functional dissociation between the two neuronal cell types (Christian and Deadwyler 1986), another observed no clear differences in sound-related firing (Moita et al. 2003). In our experiment, both types of neurons participated in the representation of sounds and places, but putative interneurons carried more stimulus-related information, whereas putative pyramidal cells carried more response direction-related information. These differences were statistically highly significant, but their neuroanatomical and functional significance remains difficult to evaluate at present, given that distinguishing anatomically defined neural populations on the basis of physiological criteria...
such as spike width and firing rate is not without difficulty (Vigneswaran et al. 2011) and given that these observed differences between the two physiologically distinguished cell types, although statistically highly significant, were nevertheless not very large, and it is important to remember that.

Although our results therefore agree with previous studies of rodent hippocampus, parallels with work on human hippocampus are harder to draw. Neurons in human hippocampus fire according to the stimulus category (Quiroga et al. 2005, 2009), which suggests a highly abstracted representation of the semantics of stimuli (Gelbard-Sagiv et al. 2008) that is largely independent of context. This contrasts sharply with the strong spatial and behavioral context dependence of sensory coding that we observed in our experiments. The sounds in our experiments were identical on both platforms and in the active and passive conditions, so if the neurons had formed abstract and invariant representations of the sounds per se, they should have exhibited similar responses on both platforms and in the passive as well as active condition. An important difference in experimental design is that in the human studies so far, the subjects were always tested in just a single location. Future studies in humans may reveal whether this categorical representation is also influenced by the subject’s position and behavioral context.

The spatial context dependence of sensory responses that we observed is certainly compatible with the hypothesis that hippocampus may create “object in place” representations by linking salient events to the spatial context in which they occur (Cohen and Eichenbaum 1993; Eacott and Norman 2004; Eichenbaum et al. 1999; Wood et al. 1999). This may also relate to the phenomenon referred to as “remapping” (Anderson and Jeffery 2003; Bostock et al. 1991; Colgin et al. 2008; Muller and Kubie 1987), whereby place fields of hippocampal neurons may change radically if the context in which an animal finds itself is altered or when the behavioral context is altered (Kennedy and Shapiro 2009; Moita et al. 2004). The fact that different sensory events appear to be encoded by partially overlapping, statistically independent populations of neurons agrees with the predictions of theoreticians such as Marr (1971), who considered this a desirable property in a memory system designed to collect “snapshots” of experience.

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DISCLOSURES

No conflicts of interest, financial or otherwise, are declared by the author(s).

AUTHOR CONTRIBUTIONS

Author contributions: P.M.I., E.V., C.H., J.W.H.S., and M.E.D. conception and design of research; P.M.I. and E.V. performed experiments; P.M.I., E.V., J.W.H.S., and M.E.D. analyzed data; P.M.I., E.V., C.H., J.W.H.S., and M.E.D. interpreted results of experiments; P.M.I., E.V., C.H., and J.W.H.S. prepared figures; P.M.I. and E.V. drafted manuscript; P.M.I., E.V., C.H., J.W.H.S., and
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