

Changes are underlined

Local field potentials and spikes in the human medial temporal lobe are selective to image category

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Abstract

Local field potentials (LFPs) reflect the averaged dendrosomatic activity of synaptic signals of large neuronal populations. In this study we investigate the selectivity of local field potentials and single neuron activity to semantic categories of visual stimuli in the medial temporal lobe of nine neurosurgical patients implanted with intracranial depth electrodes for clinical reasons. Strong selectivity to the category of presented images was found for the amplitude of LFPs in 8 % of implanted microelectrodes and for the firing rates of single and multi units in 14 % of microelectrodes . There was little overlap between the LFP- and spike-selective microelectrodes. Separate analysis of the power and phase of LFPs revealed that the mean phase was category-selective around the θ frequency range and that the power of the LFPs was category-selective for high frequencies around the γ rhythm. Of the 36 microelectrodes with amplitude-selective LFPs, 30 were found in the hippocampus. Finally, it was possible to read-out information about the category of stimuli presented to the patients with both spikes and LFPs. Combining spiking and LFP activity enhanced the decoding accuracy in comparison with the accuracy obtained with each signal alone, especially for short time intervals.

Introduction

Existing experimental approaches to investigate the functions of the brain by recording electrical signals vary from single cell recordings in animals to surface electroencephalography (EEG) in humans. Animal electrophysiology usually relies on the analysis of spiking activity of neurons, while placing less emphasis on local field potentials (LFPs), a slow non-spiking component of the recorded electrical signal. In contrast, studies of brain electrical activity in humans typically use electrical potentials recorded from the surface of the skull. Such EEG signals represent the electrical activity of large fraction of cortical and sub-cortical tissue but can still be linked to different behavioral states or cognitive functions.

Certain neurological conditions, in particular pharmacologically intractable epilepsy, require, on occasion, the implantation of either subdural electrodes that rest on the surface of cortex, or of depth electrodes that are implanted into the brain parenchyma (Bechtereva & Abdullaev, 2000; Engel, Moll, Fried, & Ojemann, 2005; Lachaux, Rudrauf, & Kahane, 2003). The signals obtained with these intracranial electrodes represent average activity of the brain with temporal and spatial resolution on the order of milliseconds and centimeters respectively. The size and impedance of these clinically used electrodes typically do not permit the recording of spiking activity of neurons. Here we present data obtained with microelectrodes implanted in the human medial temporal lobe (MTL) of epilepsy patients. Their impedance ($\sim 0.5\text{M}\Omega$) and the size (size of the tip $\sim 40\mu\text{m}$) enabled us to record spiking activity of single neurons as well as LFPs (Fried, et al., 1999). The relationship between LFPs and spiking activity in non-primates was already addressed as early as in the sixties (see e.g., (Buchwald, 1965; Fromm, 1967;

Haberly & Shepherd, 1973; John, 1967; John, 1972) and (Logothetis, 2003) for review).

There were also several studies dealing with the correlation of spiking and epileptiform activity in epileptic patients (Verzeano, Crandall, & Dymond, 1971; Wyler, Ojemann, & Ward, 1982). In monkey electrophysiology, LFPs and their relationship to spiking activity have been actively studied only in the last few years (Henrie & Shapley, 2005; Kreiman, et al., 2006; Liu & Newsome, 2006; Mehring, et al., 2003; Pesaran, Pezaris, Sahani, Mitra, & Andersen, 2002; Scherberger, Jarvis, & Andersen, 2005). For example, the LFPs activity in parietal cortex of macaque monkeys was found to discriminate between preferred and anti-preferred directions of reach plans (Pesaran, Pezaris, Sahani, Mitra, & Andersen, 2002). The LFPs in monkey motor cortex were also predictive of hand movements (Mehring, et al., 2003). Scherberger et al reported that the monkey behavioral state can be decoded with LFPs better than with the spiking activity (Scherberger, Jarvis, & Andersen, 2005). Increasing stimulus contrast was found to cause an increase in power, in particular in gamma band, of the LFP recorded in macaque V1 (Henrie & Shapley, 2005). Moreover, it occurred over a contrast range within which the spike rates of cortical cells were saturating. Recently, LFPs as well as spiking activity recorded in inferior temporal cortex (IT) of macaque monkeys were shown to be object-selective but barely related to each other (Kreiman, et al., 2006). The aim of our study is to investigate in humans, the selectivity of spiking and LFP activity recorded simultaneously from the same microelectrodes during a simple perceptual task, and the relationship between their selectivities.

The selectivity of spiking activity recorded in the human medial temporal lobe to visual categories has already been reported by our group (Kreiman, Koch, & Fried, 2000). In

this study, we investigated visual category selectivity of the spiking activity and LFPs in a different group of patients using the amplitude, power, and mean phase of the LFPs. In addition, we compared selectivity properties of the LFP and spiking activity and found a weak correlation between them. We were also able to decode information about the category of a presented stimulus using the spiking and LFP activity separately and together. Decoding accuracy in a short time window was found to be optimal using the LFPs and spiking activity simultaneously as an input to the decoding algorithm. The weak correlation of object selectivity properties of the LFPs and spiking activity and augmented decoding accuracy using both of them support the hypothesis that the LFPs contain additional information about the category of a visual stimulus.

Methods

Subjects and recordings

The data came from 12 sessions in 9 patients with pharmacologically intractable epilepsy (all right handed, 4 males, 17 to 47 years old). This set of patients is overlapping with the one used for the invariance study (Quiñan Quiroga, Reddy, Kreiman, Koch, & Fried, 2005), but we used data from different experimental sessions. For these patients extensive non-invasive monitoring did not yield concordant data corresponding to a single resectable epileptogenic focus. Therefore, they were implanted with chronic depth electrodes for 7-10 days to determine the seizure focus for possible surgical resection (Fried, MacDonald, & Wilson, 1997). Here we report data from microelectrodes in the hippocampus, amygdala, entorhinal cortex and parahippocampal gyrus. All studies conformed to the guidelines of the Medical Institutional Review Board at UCLA. The electrode locations were based exclusively on clinical criteria and were verified by magnetic resonance imaging (MRI) or by computer tomography co-registered to preoperative MRI. Each electrode probe had a total of 9 micro-wires at its end, eight active recording channels and one reference. The differential signal from the micro-wires was amplified using a 64-channel Neuralynx system (Tucson, Arizona), filtered between 1-9000 Hz and sampled at 28 kHz. Spike detection and clustering was done using recordings high-pass filtered between 300 and 3000 Hz. The local field potentials were obtained by low-pass filtering the same recordings between 1 and 100 Hz and down-sampling them to 256 Hz.

Each recording session lasted about 30 minutes. Subjects lay in bed, facing a laptop computer, on which pictures of individuals, animals, landmarks or objects were shown. The images covered about 1.5°, were centered on a laptop screen and were displayed 6 times each in pseudo-random order for 1 sec. Images were photos of animals, landmarks, celebrities which were partially chosen according to the patients preferences and photos of people and places unknown to the patients. More details about the stimulus set are available from (Quiñ Quiroga, Reddy, Kreiman, Koch, & Fried, 2005). The inter stimulus interval (ISI) was also randomized with the minimum ISI equal to 1.5 seconds. In order to encourage subjects to attend to the picture presentations, subjects had to respond whether the pictures contained a face or something else by pressing the ‘Y’ and ‘N’ keys, respectively.

Data analysis

We analyzed the signals recorded from 568 microelectrodes implanted in different locations of human medial temporal lobe. For the LFPs data we initially applied a digital notch filter at 60 Hz and the first two harmonics (4th order elliptic filter, 0.1 db peak-to-peak ripple, 40 db stopband attenuation). Recordings that showed either peaks at harmonics of 60 Hz on the power spectrum, or high frequency noise were discarded, thus obtaining a final set of 451 ‘clean’ microelectrodes for LFPs analysis, 384 of which showed spiking activity.

We discarded from the analysis trials which had more than 5 points outside of the mean plus/minus 5 standard deviations range. The mean and the standard deviation were calculated across all trials for each sample point.

Analysis of the LFPs amplitude selectivity

To quantify category selectivity of the LFPs amplitude we applied a sample-by-sample one-way analysis of variance (ANOVA) (Blair & Karniski, 1993; Guthrie & Buchwald, 1991) with the category identity as a main factor to the LFP values. The sample-by-sample ANOVA test gives a time resolved significance level of how the LFP values are different across categories. We considered a microelectrode to be selective to a category if at least 15 consecutive points (~60 msec) of the ANOVA trace crossed the significance threshold of 0.001. To check whether the ANOVA traces crossed the 0.001 threshold by chance we applied two control tests. First, we applied the ANOVA test to the second preceding stimulus onset. Secondly, we applied a bootstrap procedure by shuffling the pictures in-between categories. Such shuffling destroys information about the category but preserves time correlations of LFPs and correlations between different presentations of the same picture. Since the bootstrap procedure is time consuming, we applied it only to the selective microelectrodes (one-way ANOVA, $p < 0.001$, 15 consecutive points). The goal of the bootstrap test was to confirm that the significant p-values obtained with the ANOVA test were reflecting differences in averaged LFPs for semantic categorization but not for other possible ones, for example the cases when there was only one very strong response to a single picture.

Latencies

The latency of the selectivity for the averaged LFPs was defined as the time-point when the LFPs for three categories started to be significantly different from each other, i.e. when the ANOVA trace first crossed significance level of $p = 0.001$ for at least 15 consecutive points. Analogously, for the definition of the latency of responsiveness we applied a sample-by-sample t-test comparing the distribution of LFPs values for each

category and each time point during stimulus presentation, with the distribution of all LFPs values during the baseline interval. The moment when the t-test trace first crossed significance level of $p=0.001$ for at least 15 consecutive points for one category was defined as the latency of the responsiveness.

Phase and power analysis

To estimate the instantaneous phase and power we used the continuous wavelet transform. The LFP of each trial was convoluted with complex Morlet wavelets $\Psi(f_0, t) = (\sigma^2 \pi)^{-1/4} \exp(-t^2/2\sigma^2) \exp(2\pi i f_0 t)$, where f_0 is the central frequency and σ specifies the width of the wavelet in time domain. The Morlet wavelet is a complex sine wave whose amplitude is tapered by a Gaussian function. A wavelet family is characterized by a constant $n_c = 2\pi f_0 \sigma$, which we set equal to 6. The convolution with a complex Morlet wavelet gives a series of complex wavelet coefficients $W(f_0, t) = \int \Psi(f_0, t-\tau) s(\tau) d\tau = A(f_0, t) \exp(i\phi(f_0, t))$. From the wavelet coefficients corresponding to each frequency and time-point it is possible to define the instantaneous power as $|W(f_0, t)|^2$, and the instantaneous phase as $\phi(f_0, t)$. To resemble the main EEG frequency bands we used 11 frequencies spaced on the approximately logarithmical scale between 2.5–85 Hz. Since the distribution of the power was found to be significantly different from Gaussian, we used a non parametric sample-by-sample ANOVA (Kruskal-Wallis test) to test for selectivity of the LFP power. To test for mean-phase selectivity of the LFP we used a sample-by-sample analysis of a common mean direction. This test is a generalization of a t-test analog for circular data to more than two variables.

Since, in general, the correlation between consecutive time points is higher for the lower frequencies, both for the power and phase analyses we required that the significant

difference between categories (with $p < 0.001$) lasted for at least two-periods (Rizzuto, et al., 2003). Due to the fact that a two-period interval may be relatively short, especially for high frequencies (e.g., for 85 Hz it is only about 23 msec, which corresponds to 6 points if the sampling frequency is 256 Hz), the test for selectivity ($p < 0.001$, two periods) was validated with a bootstrap procedure where the pictures were shuffled in-between the categories.

Decoding

A trial-by-trial decoding was done with a linear Fisher algorithm (Duda, Hart, & Stork, 2001). We employed a one versus all strategy; i.e. for each trial a decision about its category was made based on the distributions of all other trials. The decoding accuracy was defined as the relative number of correct predictions. The chance level was equal to the inverse number of categories (0.33). Time profiles of decoding accuracies were calculated using the number of spikes or mean LFP values in sliding time windows as an input to the decoding algorithm. The sliding windows had 50% overlap. To increase the number of inputs to the classifier we decreased the significance threshold to 0.01 in comparison to 0.001 which was used for selectivity analysis. We also calculated the decoding accuracies with a very loose significance threshold of 0.1, which did not change significantly the results.

Results

We studied the spiking activity and LFPs recorded from the same microelectrodes implanted in medial temporal lobe of human subjects with pharmacologically intractable epilepsy. The placement of the electrodes was determined exclusively by clinical criteria (Fried, MacDonald, & Wilson, 1997). In 12 experimental sessions with 9 patients, we recorded activity from 568 microelectrodes. Only “clean” recordings (451 microelectrodes) were used for further analysis (Methods). The microelectrodes were located in the amygdala (120), hippocampus (182), entorhinal cortex (102), and parahippocampal gyrus (47). Anatomical locations of microelectrodes were selectively estimated from the fused image of structural MRI taken before implantation of the electrodes and CT taken while the electrodes were implanted (Fried, et al., 1999).

Selectivity of the amplitude of local field potentials to categories

All images were divided into three semantic categories: faces, places, and animals. LFPs were time-aligned to the stimulus onset. Intervals of one second before and one second after stimulus onset were used in the analysis. Examples of averaged LFPs, raster plots, and post-stimulus time histograms for three categories are shown in Figure 1. Here the spiking activity was clearly responsive and selective to the category “places”, increasing from its background rate of about 0.2 Hz to approximately 2 Hz. Yet surprisingly, the amplitude of the averaged LFPs for the category “places” was the smallest among the three categories. The averaged amplitudes of LFPs for the two other categories (“animals” and “faces”) were significantly different from baseline. Another example is shown in Figure 2. Here the spiking activity as well as LFPs for the category “faces”

were significantly different from the responses to images from the other two categories. The LFPs for the three categories from Figure 2 are plotted in Figure 3a. The corresponding ANOVA trace (Methods) is shown in Figure 3b. Note that after stimulus onset, there are two intervals corresponding to positive and negative reflections of the averaged LFPs from baseline where the ANOVA trace is far above the chosen significance value of 0.001. To verify the category selectivity of the LFPs we applied two control tests (Methods). Note that for the example in Figure 3, there was no single point before stimulus onset where the amplitude of the LFPs was selective. In fact, none of the 451 microelectrodes showed selectivity during the baseline interval. In total, we found that 36 of the 451 microelectrodes (8.0%) produced LFPs with a significant category selectivity (one way ANOVA, $p < 0.001$, 15 consecutive points, bootstrap 1000 shuffles; Methods). Thirty of these microelectrodes were located in the hippocampus, three in the amygdala, three in the entorhinal cortex, and none in the parahippocampal gyrus (Figure 4). We did not find any significant difference ($p > 0.05$, binomial test) in the number of LFP- amplitude selective microelectrodes between different hemispheres (left 19/251, right 17/200). Equal number of selective microelectrodes were found in the epileptogenic temporal lobe and in contra-lateral lobe (18/270-18/181).

The latency of the selectivity, i.e. the moment when the LFPs of the three categories started to be significantly different from each other, was 460 ± 15 msec (mean \pm S.E.M.). For the selective microelectrodes, we also calculated the latency of the LFP responsiveness, i.e. the when LFPs of a category started to be significantly different from the baseline (Methods) and found it to be 369 ± 53 msec (mean \pm S.E.M.) which is significantly shorter than the latency of the selectivity ($p < 0.01$, t-test). This difference is

explained by the observation that very often the LFPs of all three categories first start to deviate from the baseline and only later from each other (Supplementary Figure 1).

Selectivity of the mean phase and power of the LFPs

In addition to the selectivity of the LFPs amplitudes, we studied whether their phase and power were category-selective. The instantaneous phase and power of each category were defined for every time point in the interval [-1:2] sec using a complex Morlet wavelet transform (Grossmann, Kronland-Martinet, & Morlet, 1989), widely used in EEG analysis (TallonBaudry, Bertrand, Delpuech, & Pernier, 1997) (Methods). We applied a sample-by-sample, one way, non-parametric analysis of variance to the values of the log-transformed power in different frequency bands. This gave a non-parametric ANOVA trace for each microelectrode and frequency band. Similarly, to quantify the category selectivity of the mean phases, we applied a sample-by-sample test for a common mean direction (Fisher, 1995) (Methods). In Figure 5 an example of a power-selective microelectrode is presented. The upper panel shows the averaged log-transformed power in the γ band (the central frequency of the Morlet wavelet was 45 Hz) for the three categories. Around 320 msec, the power of the category “faces” starts to be clearly different from the baseline power and from the power in the other two categories, the latter being reflected in the ANOVA trace plot (lower panel). An example of a mean phase-selective microelectrode is presented in Figure 6. The averaged LFPs filtered in θ band (the central frequency of the Morlet wavelet was 6 Hz) for the three categories are shown in Figure 6a. Figure 6b shows the significance of the sample-by-sample test for a common mean direction. The test reached its maximum significance around 300 msec. The phase distributions and their mean direction at this particular time are shown in

Figure 6c. The length of the mean direction vector is proportional to the difference between the phase distribution for a given category from the uniform distribution (Rayleigh test). Figure 7a plots the total number of the microelectrodes showing selectivity to a category with the power and Figure 7b with the mean phase across different frequency bands. The percentage of microelectrodes that showed selectivity with power (24; 5.3%) or mean phase (27; 6.0%) was relatively small. Comparing the data presented in Figure 7a and b one can see a higher percent of mean phase selective LFPs for the lower frequencies and an opposite trend for the power selective LFPs, namely higher percentage of power selective LFPs for higher frequencies (30-100Hz). It leads to the intriguing hypothesis that there are two possible different mechanisms for the selectivity of LFPs. One involves phase locking in the lower frequencies and another one engages power increase in higher frequencies.

Half (13 microelectrodes) of the mean phase-selective microelectrodes were also selective for the amplitude of the LFPs (dark blue bars in Figure 7b), while only four microelectrodes showed selectivity for both power and the amplitude of the LFPs (dark blue bars in Figure 7a). This is not very surprising since the activity phase locked to the stimulus onset is mostly preserved in the averaged LFPs whereas the induced, non-stimulus locked, activity is averaged out and is revealed only in the averaged power.

Selectivity of spiking activity to categories

The spiking activity recorded with the same microelectrodes used to record the LFPs was preprocessed using a novel spike sorting algorithm (Quiñero, Nadasdy, & Ben-Shaul, 2004). To quantify the category-selectivity of the spiking activity, we applied an one-way analysis of variance with the category identity as a main factor, and the number

of spikes in the interval [300:1000] msec (Quiñan Quiroga, Reddy, Kreiman, Koch, & Fried, 2005) as repeated measures. A t-test comparison with the baseline interval [-1000:-300] msec was used as *post hoc* test to define the responsive category. Additionally, we performed a bootstrap test. The pictures were randomly shuffled between categories and an ANOVA test was applied. We found that 66 out of the 591 recorded units (11.2%) had a spiking response with significant category-selectivity (ANOVA, $p < 0.001$, t-test, $p < 0.001$, bootstrap 1000). Fourteen units showed a significant decrease in firing rate, and 8 units showed a significant increase to one category and a significant decrease in firing rate to another one.

Units selective to at least one of the three semantic categories were recorded from 56 of the 384 microelectrodes (14.6%) used in the analysis. These numbers are comparable with those reported in (Kreiman, Koch, & Fried, 2000). As a control, the baseline and stimulus presentation intervals were exchanged and the same analysis was repeated. In this case, only one significant response was found. The distribution of the spike-selective microelectrodes across different brain regions was more uniform in comparison to that of the LFPs. We found 18 selective microelectrodes in hippocampus, 25 in the amygdala, 8 in entorhinal cortex, and 5 in parahippocampal gyrus (Figure 4).

To compare the latency of the selective spiking and LFP activities, we convolved each spike train with a Gaussian kernel (100 msec width at half height) and repeated the same analysis used for the selectivity of the LFPs amplitudes (sample by sample one way ANOVA, $p < 0.001$, 15 consecutive points). The latency of selectivity was defined as the first time-point when there was a significant difference between the categories. It was equivalent to the latency of responsiveness. The average value for the latency of

selectivity was found to be 341 msec (14 msec S.E.M.). It was significantly ($p < 0.001$, t-test) earlier than the latency of selectivity of the LFPs. Yet, the latency of the spiking responses was found to be not significantly different from the latency of the LFP responses ($t > 0.3$, t-test).

For the spiking activity we found 24/315 category selective units in the left hemisphere and 42/276 in the right one, and 20/343 category selective units in the epileptogenic hemisphere and 46/248 on the contralateral one. We found more category selective units in the contralateral side. Since we do not have extensive patient statistics (9 patients) we can not make any conclusive claims about lateralization of the category selectivity effect.

In total, we found 85 microelectrodes which produced either selective LFPs or selective spiking responses and which passed the bootstrap test which shuffles pictures between categories. However, only 7 of them were selective for both the LFPs and the spiking activity. Six microelectrodes showed mean phase and spiking selectivity and 10 spike-selective microelectrodes power-selectivity. The distribution of these channels across the different frequency bands is indicated with red bars in Figure 7.

Decoding with LFPs and spiking activity

We applied a linear decoding algorithm to the LFPs and the spiking activity recorded simultaneously from many microelectrodes in the MTL in order to ascertain how much information pertaining to the semantic category of the images can be inferred read-out from the neural data. Here we use the term “decoding” in the computational sense, namely, we studied how reliably one can predict in each single trial the category identity of the stimulus given the firing of the neurons or the LFP activity. We studied the time profile of the decoding accuracy (Methods) in the time interval [-1:2] sec. The inputs to

the decoding algorithm were the number of spikes for each category or the mean value of the amplitude of the LFP defined in sliding windows of different sizes with 50% overlap. Only the activity of amplitude-selective microelectrodes (one-way ANOVA, $p < 0.01$, 15 consecutive points) was taken as an input.

We found that the time profile of the decoding accuracy using spiking data increased with the length of the moving window (Figure 8b) and saturated for windows longer than 200 msec. For LFPs, the time profile remained approximately at the same level (Figure 8a) for different window sizes. Both for spiking and LFP data, and for all durations of the moving window, decoding accuracy during the baseline interval did not differ from chance (t-test, $p > 0.05$). LFPs slightly outperformed the spiking activity only for very small window sizes of 10 and 20 msec. Moreover, for 10 msec windows the classifier could barely distinguish between the categories using only spikes. This can be explained by the typically low firing rates of these neurons, considering that they may not produce a single spike during a 10 msec window. Combining the LFP and spiking activities increased the decoding accuracy for all window sizes between 10 and 200 msec (Figure 8c). The traces in Figure 8d show the decoding accuracies obtained with the LFPs, with the spikes and with both the LFPs and spikes for a window length of 20 msec. The red crosses indicate time-points when decoding using the combination of the LFPs and spikes was significantly better (paired t-test, $p < 0.01$) than the decoding accuracy using only the amplitude of the LFPs, while, the blue circles show the comparison between accuracy of the combined spikes and LFPs signal with the one using only spikes. For longer windows, decoding accuracy with combined LFPs and spikes input was also slightly better than accuracy obtained with each signal alone. It is notable that the time when the

decoding accuracy became significantly different from chance was very similar for the LFPs and the spiking activity, around 300 msec after stimulus onset. The fact that we can infer category identity by applying a classifier to LFPs does not imply that the brain makes use of this information, but only that this information is present in the MTL and could be used by post-synaptic processes.

Discussion

In previous studies, the spiking activity of single neurons in human MTL was found to be selective to different categories of visual stimuli (Kreiman, Koch, & Fried, 2000) and even invariant to different views of the same person or object (Quian Quiroga, Reddy, Kreiman, Koch, & Fried, 2005). The analysis of the LFPs recorded from the surface of the brain revealed face-selective LFPs in extrastriate cortex (Allison, et al., 1994; Allison, McCarthy, Nobre, Puce, & Belger, 1994). Later studies using intracranial depth electrodes localized the source of the face-selective LFPs near the fusiform gyrus (Allison, Puce, Spencer, & McCarthy, 1999; Lachaux, et al., 2005), in good agreement with fMRI findings (Kanwisher, McDermott, & Chun, 1997). In this paper, we undertook a combined study of the LFPs and spiking activity recorded by the same microelectrodes to assess their object- and face-selectivity properties. We found that the spiking activity recorded from 56 microelectrodes was selective for semantic categories. The amplitude of the LFPs from a smaller number of microelectrodes (36) also showed category-selectivity, but not necessarily to the category “faces”.

We also found that the power in the γ band was discriminative between categories but only in a small number of microelectrodes. This finding is similar to one reported by (Oya, Kawasaki, Howard, & Adolphs, 2002), who showed selectivity of the LFP γ power

in the amygdala in response to emotional faces. Although we did not find any power selectivity in lower frequency bands, for a small number of microelectrodes the mean phases of three categories were different in the δ (1-4 Hz) and θ (4-8 Hz) bands, but not in the γ band (higher than 30 Hz following the definition in (Engel, Fries, & Singer, 2001)). This suggests two possible mechanisms for selectivity of LFPs: via phase-locking in the lower frequencies bands and/or via power-increase at high frequencies. Since the power of δ and θ oscillations is significantly larger than the power of γ oscillations (see Supplementary Figure 3), it is more efficient, i.e. less energy consuming and faster, to transmit information by modulation of γ power. On the other hand, γ oscillations are much faster than δ or θ ones, therefore a small jitter in γ oscillations will destroy their synchronization but would hardly influence slow δ and θ oscillations. Previous studies with rats performing spatial tasks also found a phase locking of spikes with the ongoing activity in the θ band (Siapas, Lubenov, & Wilson, 2005). Synchronization of γ oscillations was also suggested as possible mechanism for information processing (Singer, 1999). The vast majority of the LFP amplitude-selective microelectrodes were found in the hippocampus (30 out of 36). The relative number of hippocampal selective microelectrodes (16%) was three times larger than in all other areas (Figure 4). At the same time, the relative number of spike-selective electrodes was evenly distributed across all four investigated areas. This discrepancy suggests that the category-selective LFPs are either the result of local processing within hippocampus or that the hippocampus receives a category specific input from adjacent areas.

Only a small overlap was observed between spikes and LFPs selective microelectrodes.

This lack of correlation supports the view that neurons in MTL are only weakly spatially

clustered in terms of the semantic categories (faces, places, animals). This result is in line with the weak correlation between the object-selective LFPs and object-selective spiking activity recently reported in monkey IT (Kreiman, et al., 2006). It also suggests that spiking activity and LFPs contain different information about stimulus category.

Halgren and coworkers (Halgren, et al., 1980) recorded LFPs in the human MTL during an “oddball” paradigm and found a P300 evoked potential well known from surface EEG measurements. Some studies argue for a generation of the P300 in the hippocampal formation and amygdala (Halgren, Marinkovic, & Chauvel, 1998; McCarthy, Wood, Williamson, & Spencer, 1989). To check whether the evoked potentials which we observed were task-dependent, we repeated our experiment in one patient without a task, i.e. the patient was asked to passively look at the pictures presented on the screen for 500 msec. We found two (out of 24 microelectrodes analyzed for this session) LFPs amplitude-selective microelectrodes (Supplementary Figure 2). This corresponds to the 8 % of LFP selective microelectrodes found with task and argues in favor of task independence.

The LFPs represent the average dendrosomatic activity of pre-synaptic signals of large neuronal populations (Logothetis, 2003; Mitzdorf, 1985). Therefore, the observed selectivity of LFPs in the MTL might be caused by specific pattern of dendritic activity arising e.g., from prefrontal cortex, which has been reported to be involved in categorical representation of visual stimuli (Freedman, Riesenhuber, Poggio, & Miller, 2001) or by local synaptic circuitry which is differentially activated for different categories of visual stimuli. It is not possible to distinguish among these possibilities with the current data.

The weak correlation in the selectivity properties of LFPs and spikes suggest that they reflect two different aspects of brain activity. A similar disassociation between LFPs and spikes was also observed e.g. in macaque V1 (Henrie & Shapley, 2005). These authors hypothesized that the network activity captured by LFPs originates from inhibitory interneurons whereas single unit activity is largely biased towards pyramidal neurons. This relative independence observed in the selectivity of the spiking and LFPs activities is compatible with our decoding analysis. Reading out both the mean number of spikes as well as the mean amplitude of the LFP allowed us to infer the identity of the category of the visual stimulus significantly better than using either measure by itself. This effect was more pronounced when information from short time windows was used for classification.

Acknowledgments

We thank all patients for their participation and Eric Behnke, Tony Fields, Emily Ho, Eve Isham, and Charles Wilson for technical assistance. We also would like to thank Gabriel Kreiman and Daniel Rizzuto for helpful discussions. This work was supported by grants from the NINDS, NIMH, NSF, DARPA, EPSRC, the Office of Naval Research, the W.M. Keck Foundation Fund for Discovery in Basic Medical Research, the Gordon Moore Foundation, the Sloan Foundation, and the Swartz Foundation for Computational Neuroscience.

Figure Captions

Figure 1 Spike and LFP responses from a microelectrode in the left medial hippocampus

For each of the three image categories shown, (a) raster plots, (b) post stimulus time histogram, and (c) average LFPs (red) plus/minus one standard error of the mean across trials (blue, thin) are plotted. Vertical dashed lines indicate onset (zero on time axis) and offset (1 sec) of the image. The number of pictures in a category is specified by the number in the brackets following the name of the category. Each picture was shown six times. The baseline firing rate was about 0.2 Hz for all three categories. Only during presentation of “place” pictures the firing activity increase to 2 Hz. In contrast, the average LFPs for “places” had the smallest amplitude, whereas the maximum amplitude of the average LFPs for the categories “animals” and “faces” was about 50 μ V. The baseline amplitudes for all three categories were about 15 μ V.

Figure 2 Spike and LFP responses from a microelectrode in the left anterior hippocampus.

The baseline rate of this unit was 10.2 Hz. It decreased significantly ($p < 0.001$, t-test) upon presentation of animal pictures, remained unchanged for “places” and increased significantly ($p < 0.001$, t-test) for “faces”. The LFPs for “faces” were significantly different from baseline ($p < 0.001$, ANOVA, see next Figure).

Figure 3 Time resolved selectivity analysis

Selectivity of the LFP from the microelectrode of Fig. 2. (a) Averaged LFPs (thick line) for the 3 categories plus/minus one standard error of the mean (thin lines). (b)

Normalized p-values ($-\log_{10}(p)$) obtained from the ANOVA test. The dash-dotted line

corresponds to a significance of $p=0.001$. The significance values are in logarithmic scale, significance less than 0.001 corresponds to the values on y-axis larger than $-\log_{10}(0.001)=3$. The ANOVA test showed a significant difference among categories in two intervals of about 50 and 100 msec duration, corresponding to positive and negative reflections of the averaged LFPs for the category “faces”.

Figure 4 Localization of selective microelectrodes

(a) Total number of category-selective microelectrodes across different brain regions (hippocampus (Hipp), amygdala (Am), parahippocampal gyrus (PG), entorhinal cortex (EC)); (b) the number of selective microelectrodes normalized to the number of electrodes implanted in the area. The vast majority of the amplitude-selective microelectrodes were found in hippocampus. Spike-selective microelectrodes were more evenly distributed across these four regions. Blue bars indicate the LFP-amplitude selective microelectrodes and red bars are for spike-selective microelectrodes.

Figure 5 Example of a microelectrode with selective LFP power

(a) Averaged log-transformed power in the γ band with a central frequency of the complex Morlet wavelet at 45 Hz (thick line) for three categories plus/minus one standard error of the mean (thin lines); (b) Normalized p-values obtained from ANOVA test. The dash-dotted line corresponds to the significance level $p=0.001$. The averaged instantaneous power is larger for the category “faces” than the power for the other two image categories for about 100 msec starting at ~ 300 msec.

Figure 6 Example of a microelectrode with selective LFP mean phase

(a) Averaged LFPs filtered in the θ band with the central frequency of the complex Morlet wavelet at 6 Hz (thick line) for three categories plus/minus one standard error of

the mean (thin lines); (b) normalized p-value retrieved from a test for a common mean direction. The black vertical line indicates the time point at which the mean phases were most different. The dash-dotted line corresponds to the significance level $p=0.001$; (c) phase distributions across trials taken at the moment of the largest difference among categories, as shown in (a) and (b) with the solid vertical line. Colored vectors show the mean phase direction. The length of the vector is proportional to the significance of the difference of the phase distribution for a given category from a uniform circular distribution (Rayleigh test). P-values are given next to the category names. Here, the phase distribution of the category “faces” significantly differs from a uniform circular distribution ($p < 10^{-7}$).

Figure 7 Number of microelectrodes with showing selectivity with LFP power and mean phase

The distribution of the (a) power and (b) mean phase-selective microelectrodes across different frequency bands of the LFPs (cyan bars). A total of 20 microelectrodes out of 475 showed category-selectivity using the power analysis and 29 microelectrodes showed category-selectivity using mean phase analysis. In principle, the same microelectrodes could be selective in different frequency bands. For the total number of selective microelectrodes we count such electrodes only once. The dark blue bars on both plots indicate the proportion of microelectrodes which also showed selectivity in the LFP amplitude analysis. A large overlap is observed in selective microelectrodes according to LFP amplitude analysis and the mean phase analysis (15 microelectrodes). The red bars on both plots show the number of microelectrodes which were found to be selective with spikes.

Figure 8 Time profile of decoding accuracy, using the spiking activity, the LFPs, and both signals together.

Decoding accuracy using (a) the average value of LFPs amplitude, (b) firing rates (SPK) and (c) their combination (SPK/LFP). The different curves on the (a,b,c) subplots correspond to window sizes of 10, 20, 50, 100 and 200 msec. Decoding accuracy with LFPs is nearly independent of the window size in this range, while accuracy with spikes (with or without LFP) increased with the window size. All curves were smoothed with a 5 points moving average. Dash-dotted lines show the confidence intervals obtained for the time points from a t-test comparison with chance level (1/3, black line), $p < 0.01$.

Decoding with LFPs, spikes and their combination is shown in panel (d) for a window of size 20 msec. Optimal accuracy is achieved with the combination of LFPs and spiking activity. Red crosses (blue circles) indicate time-points where decoding for combination of the LFPs and spiking activity was significantly better than decoding using only LFPs (only spikes) (paired t-test, $p < 0.01$).

Supplementary Figure Captions

Figure S1 Latency of selectivity vs latency of responsiveness

(a) Averaged LFPs (thick line) for the 3 categories plus minus one standard error of the mean (thin lines) corresponding to an electrode implanted in the right amygdale. (b) Normalized p-values obtained with the ANOVA test of selectivity (black thick line) and the t-tests or responsiveness for each category (color lines). A sample-by-sample t-test was applied to the distribution of LFPs values for each category and each time point during stimulus presentation with the distribution of all LFPs values during the baseline interval. The dash-dotted line corresponds to a significance of 0.001. The ANOVA trace first crosses the significance level of $p=0.001$ at ~300 msec whereas the t-test reached significance earlier, at ~230 msec.

Figure S2 Time resolved selectivity analysis

Selectivity analysis of LFPs recorded during a passive viewing task. Each stimulus was presented for 500 msec. (a) Averaged LFPs (thick line) for the 3 categories plus minus one standard error of the mean (thin lines). (b) Normalized p-values obtained with the ANOVA test. The dash-dotted line corresponds to a significance of $p=0.001$. All three categories show significant response but there is a clear difference between the category “faces” and two other categories

Figure S3 Average LFPs power spectrum

Red (blue) solid line shows a power spectrum averaged across stimulus presentation [0:1] sec (baseline [-1:0] sec) interval for all analyzed LFPs. Dash-dotted lines depict least-squares logarithmic fit, $f^{-\alpha}$. For stimulus presentation interval average α was found to be $\langle \alpha \rangle_{microelectrodes} = 2.01 \pm 0.02$, (mean \pm S.E.M.) and for the baseline interval

$\langle \alpha \rangle_{\text{microelectrodes}} = 2.04 \pm 0.02$, (mean \pm S.E.M.). Note that the red and blue curves practically almost overlap. Drop at 60 Hz is due to the digital notch filter applied at 60 Hz.

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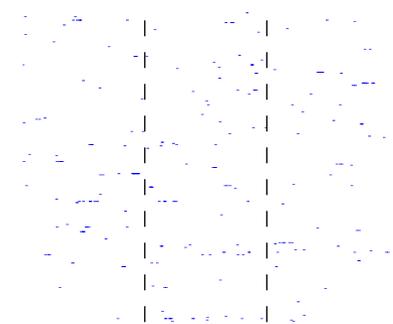
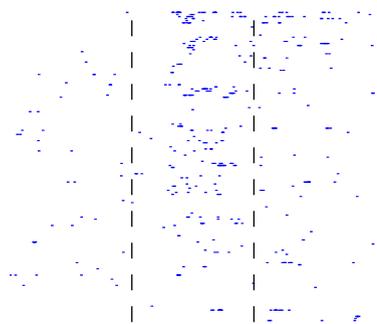
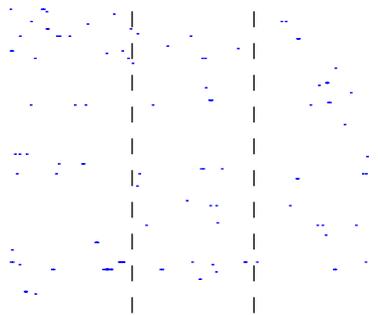
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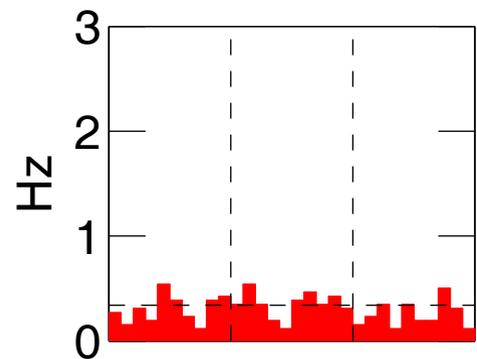
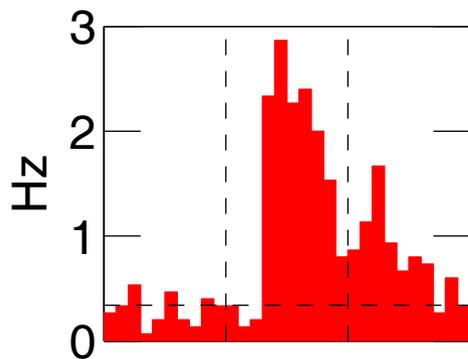
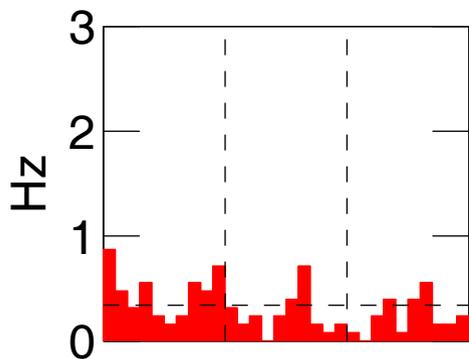
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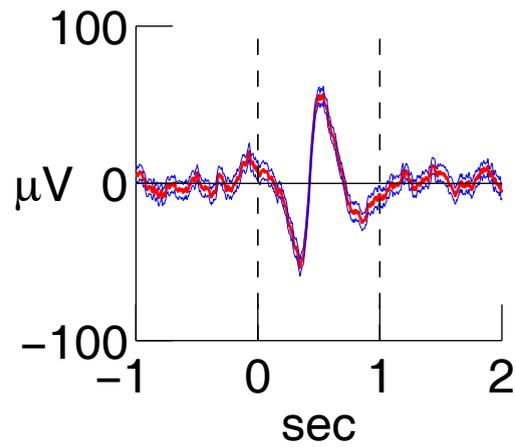
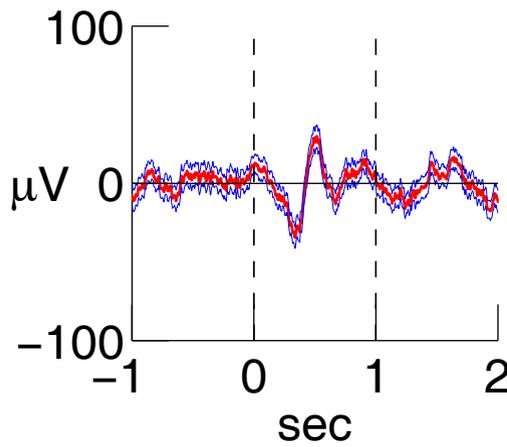
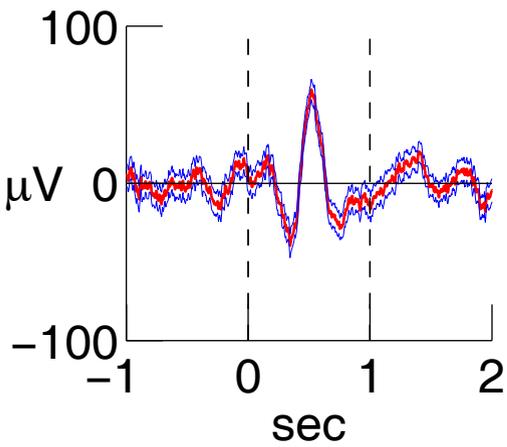
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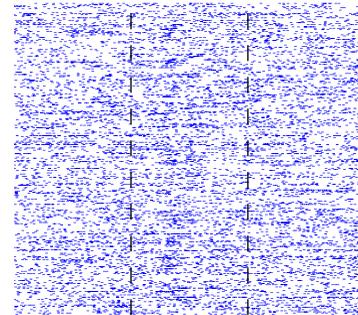
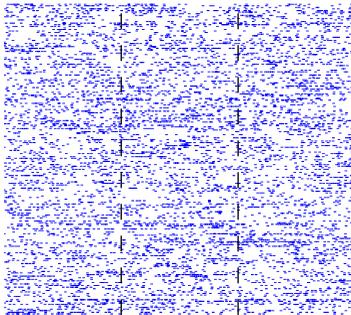
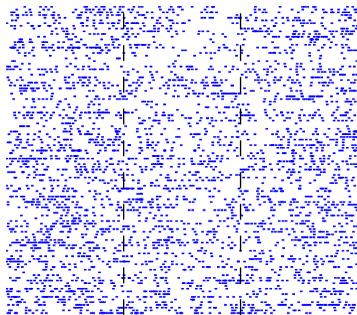


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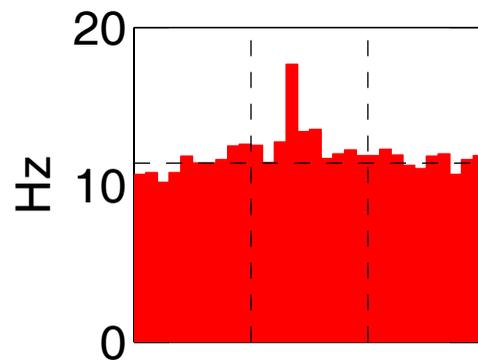
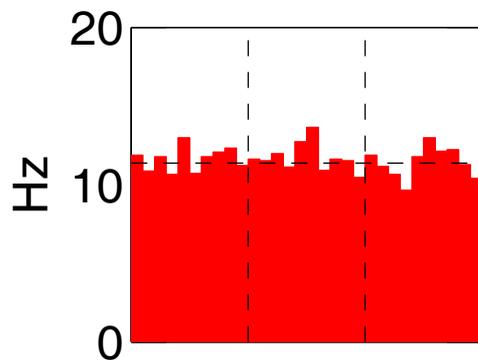
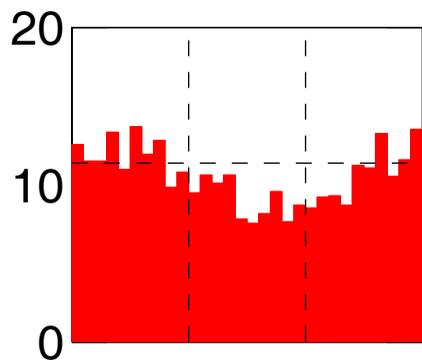
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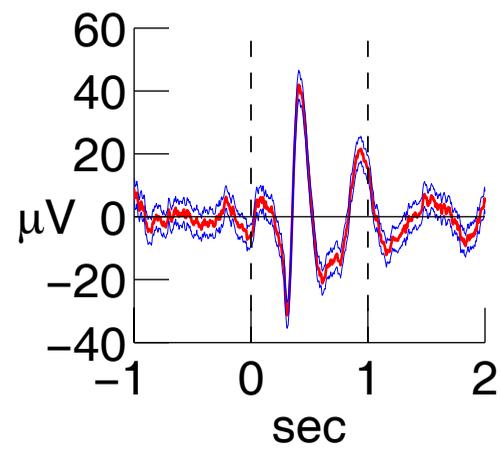
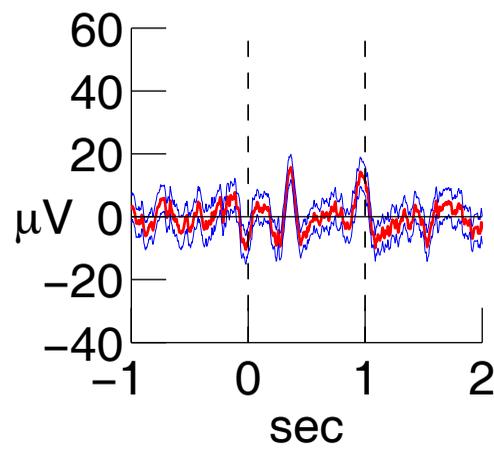
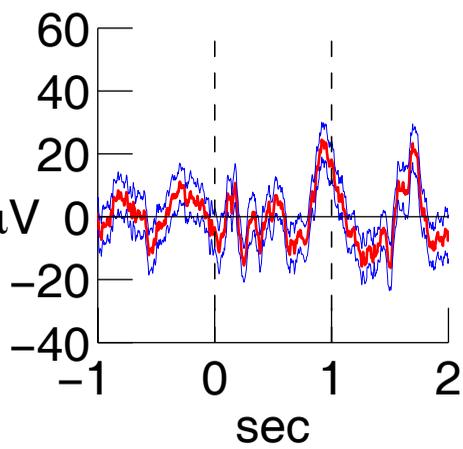
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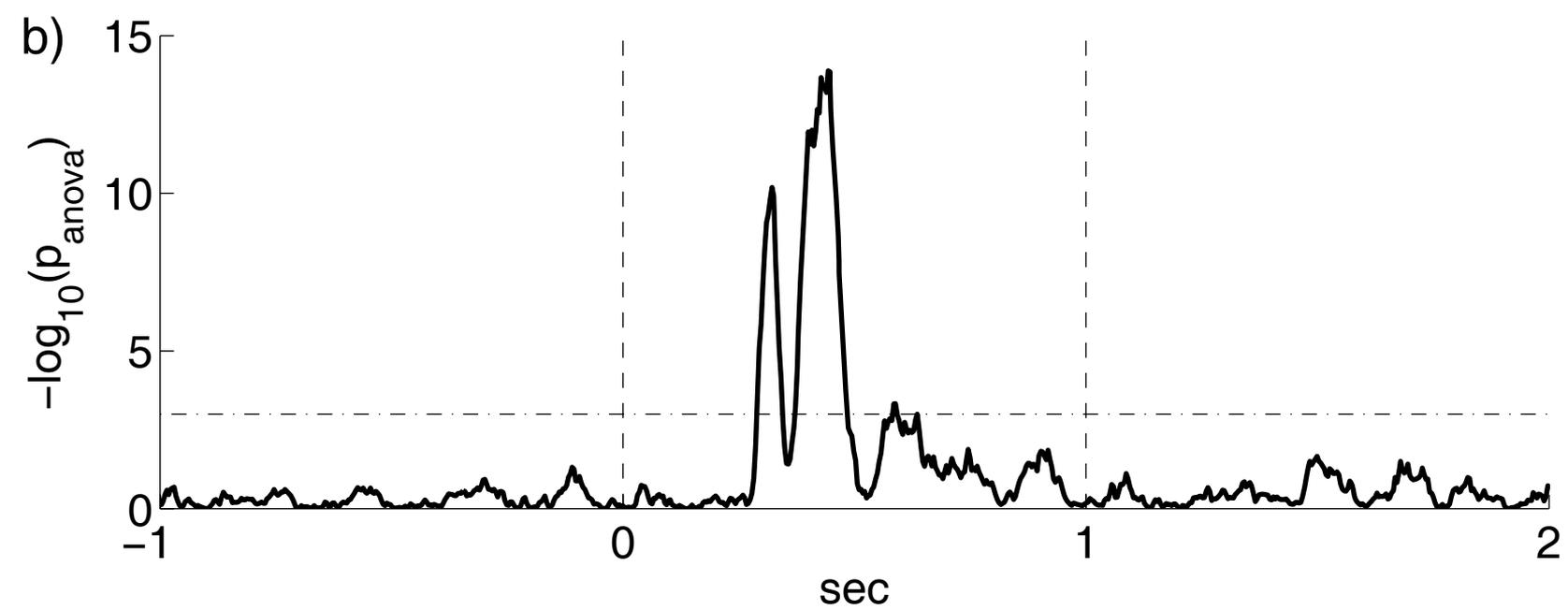
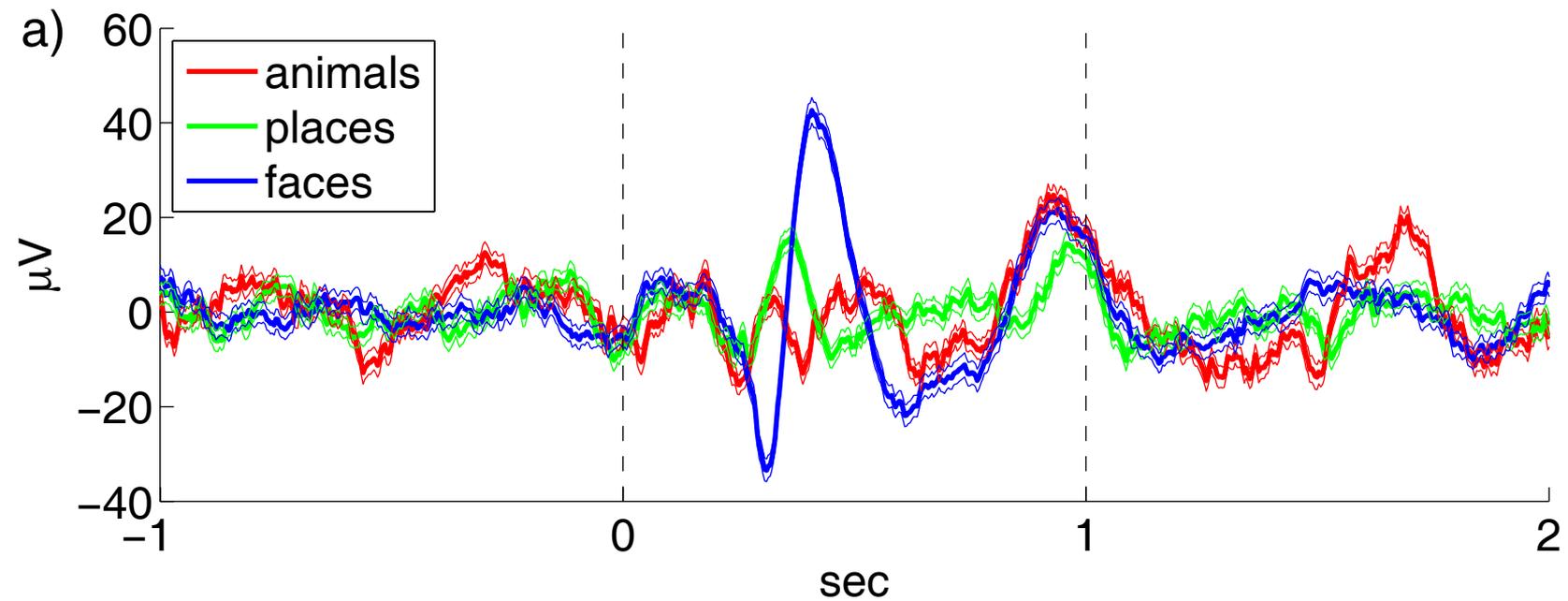


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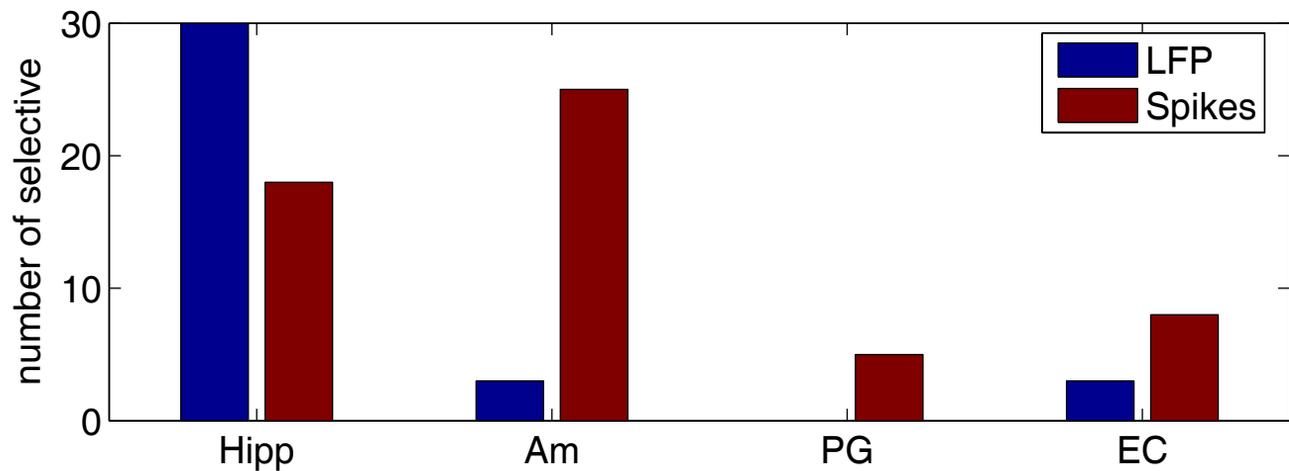


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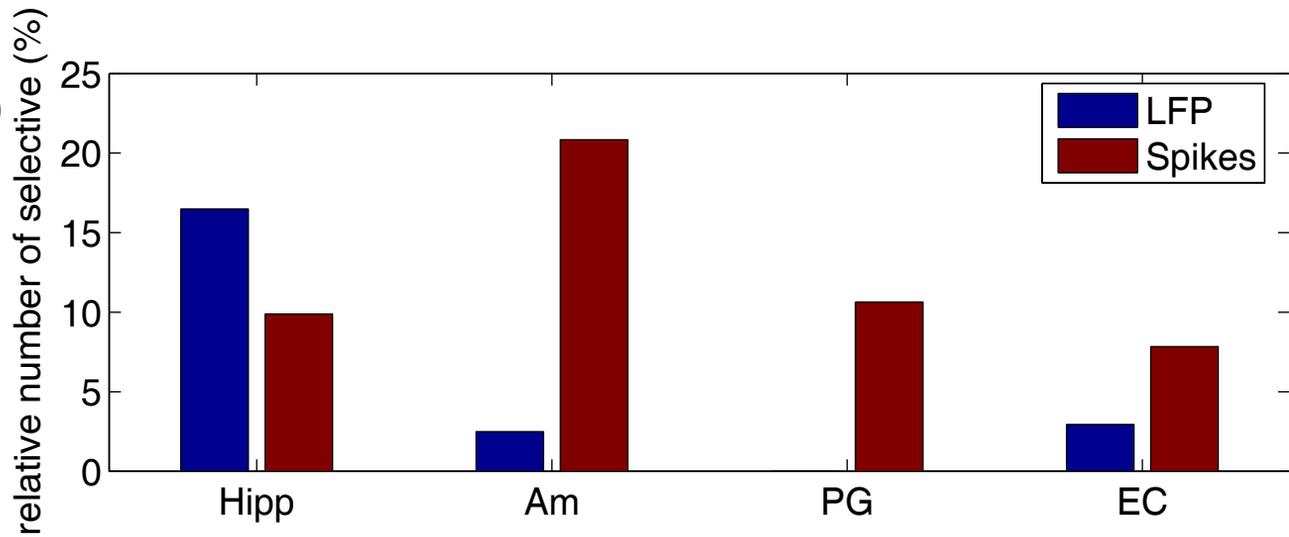


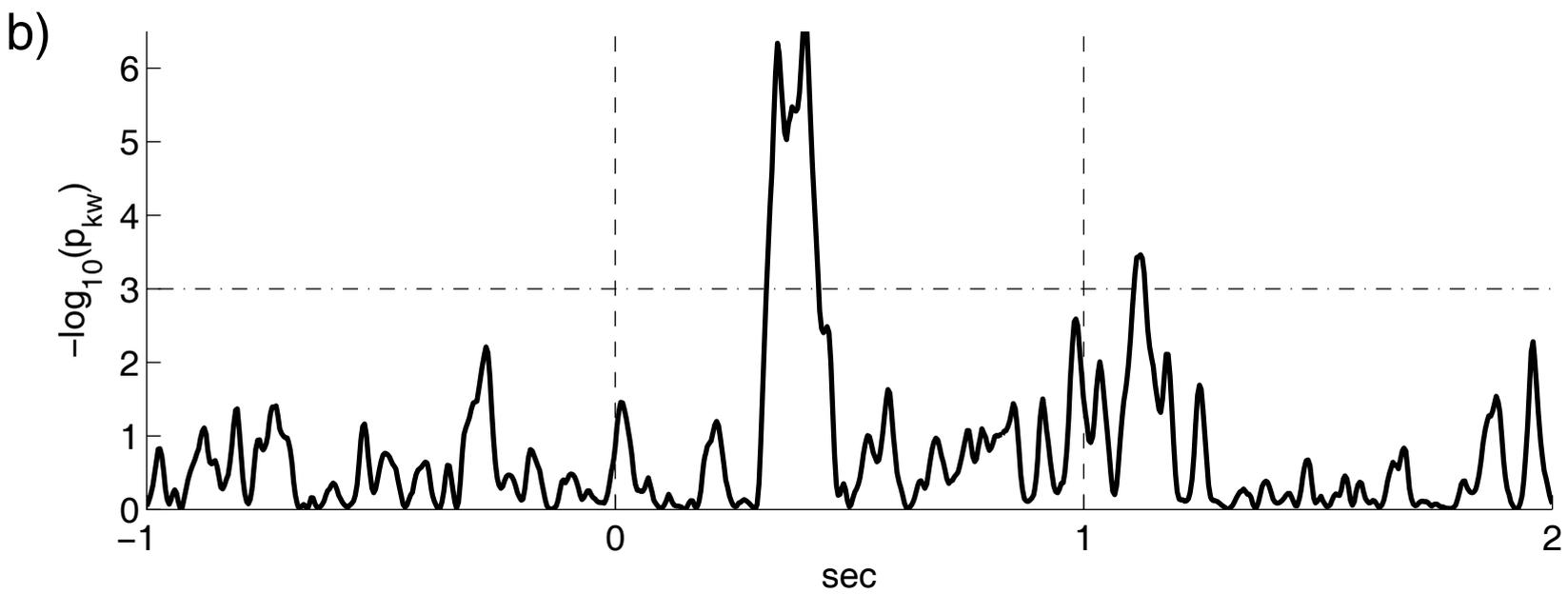
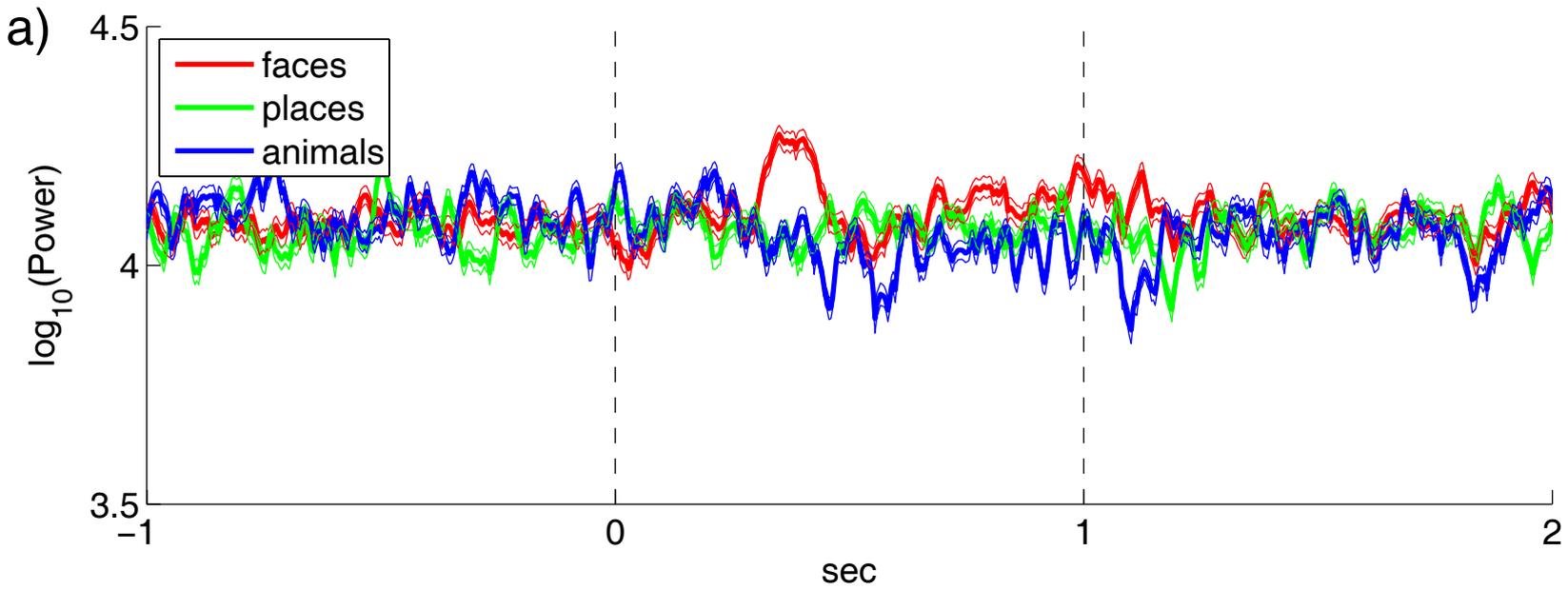


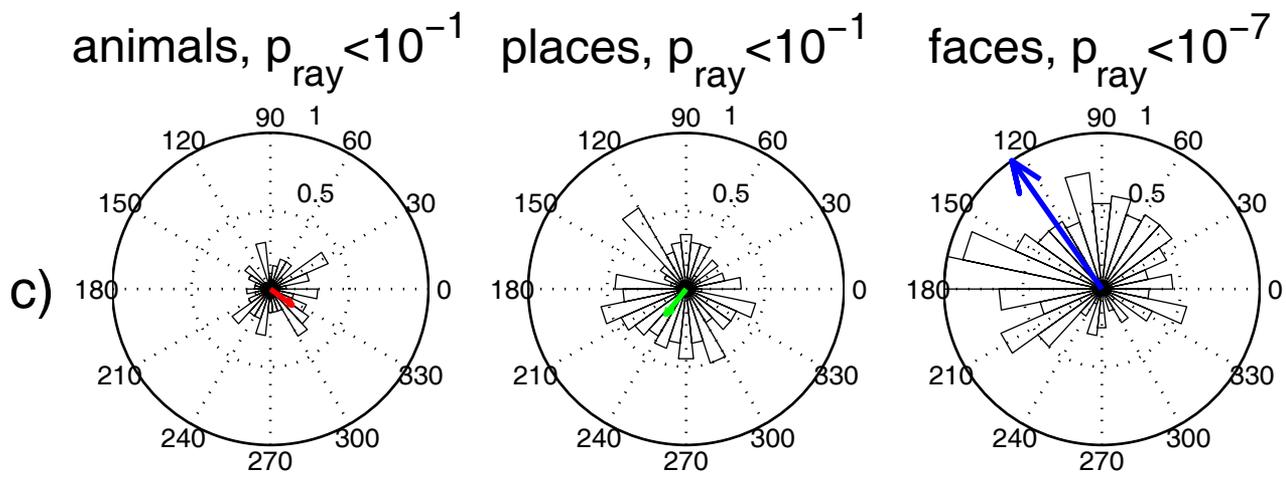
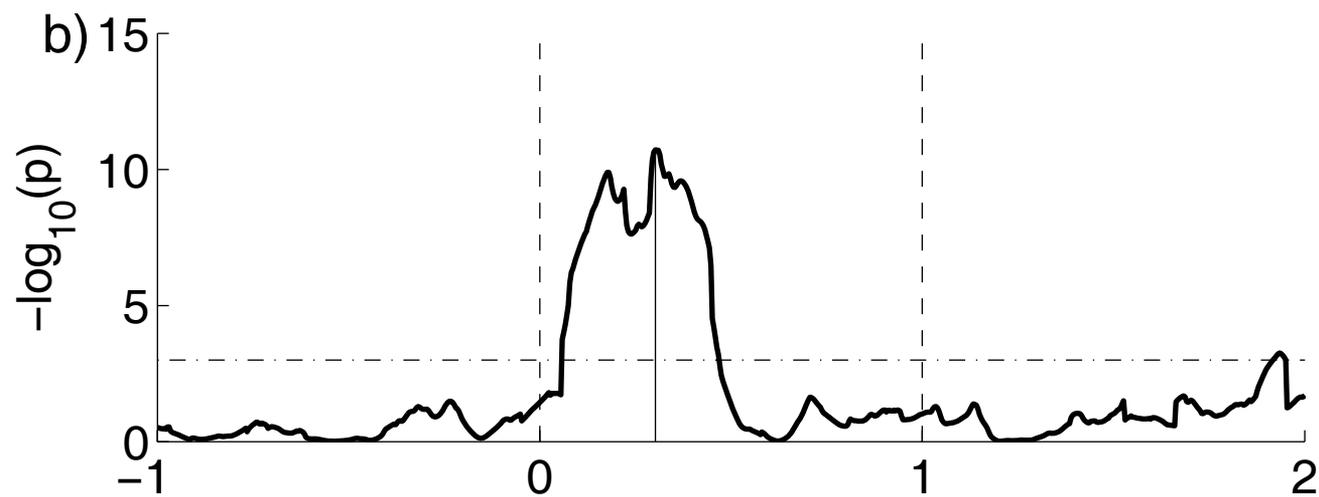
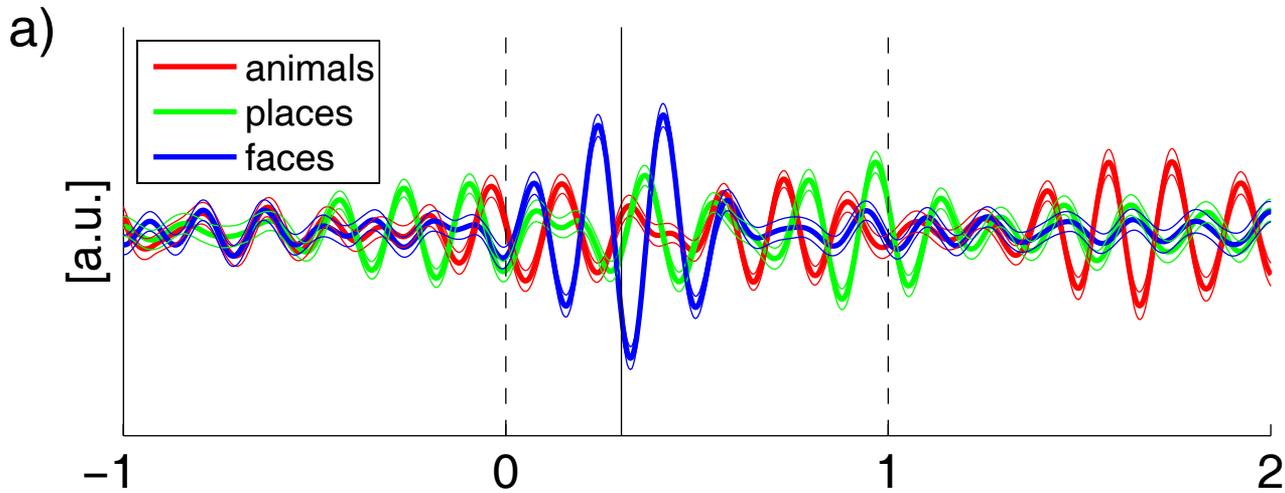
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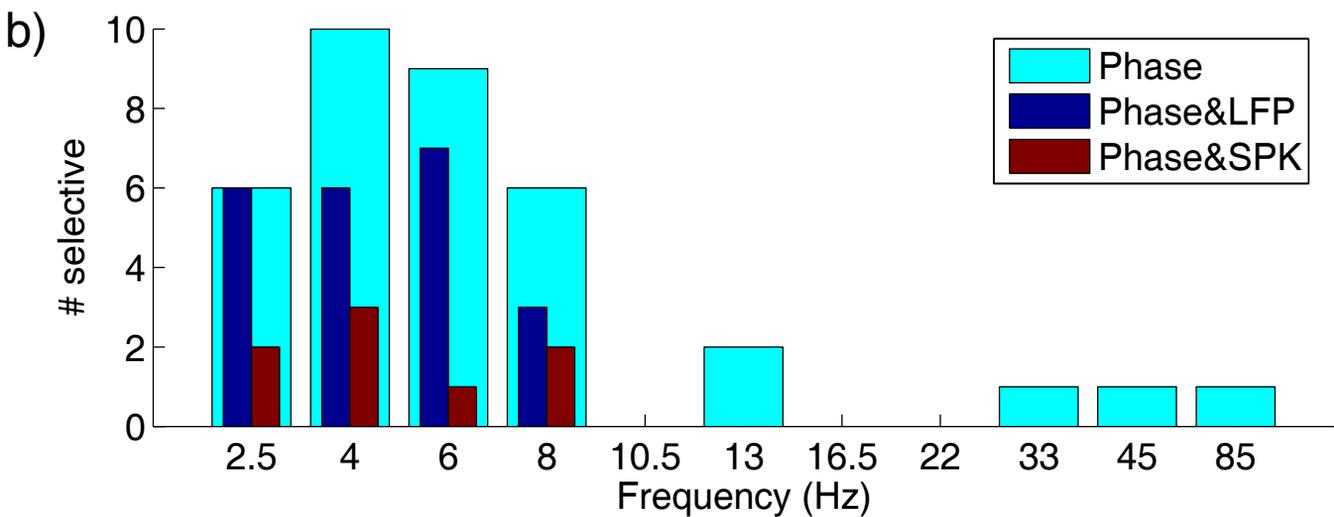
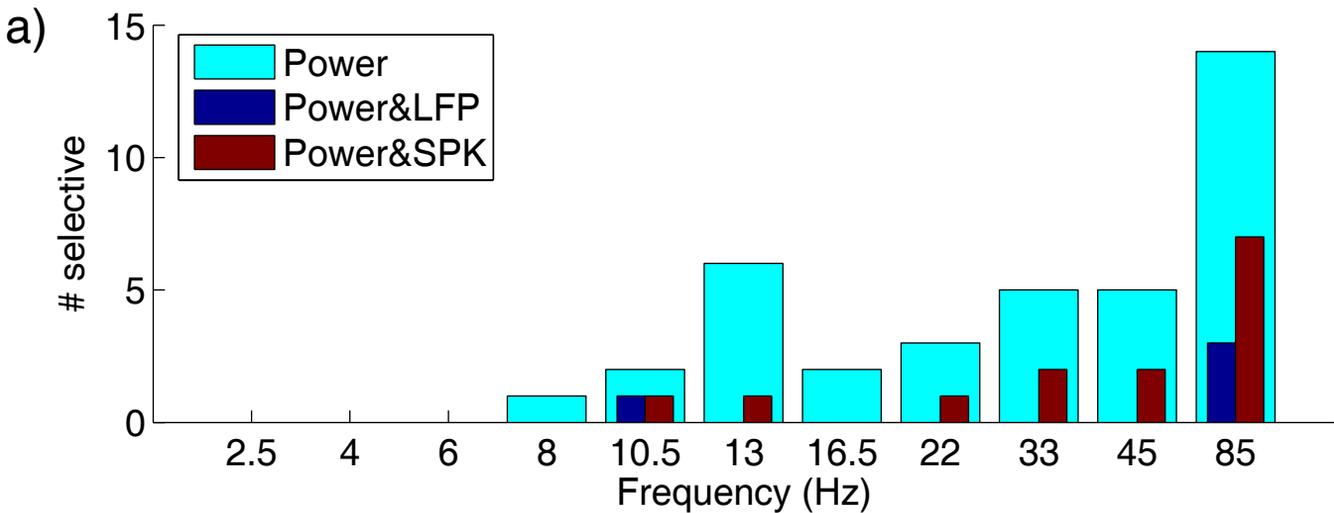


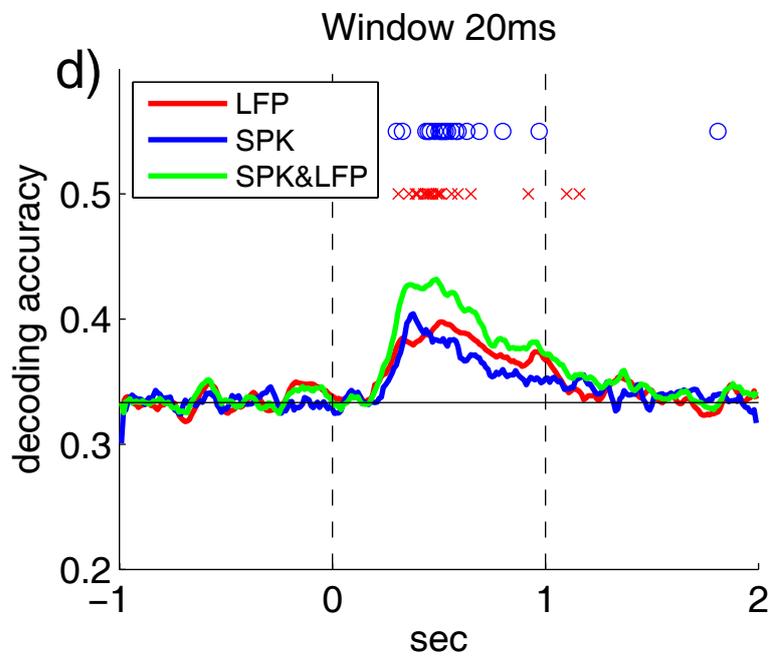
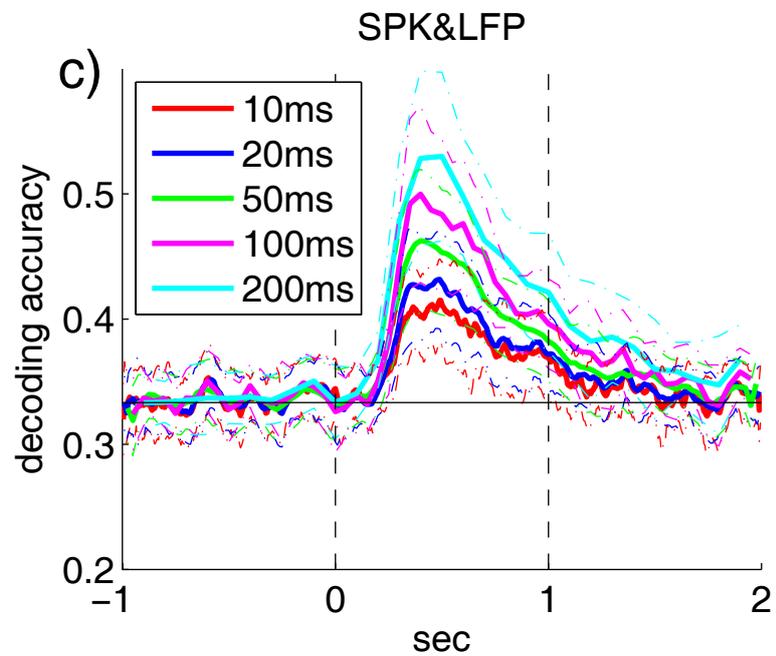
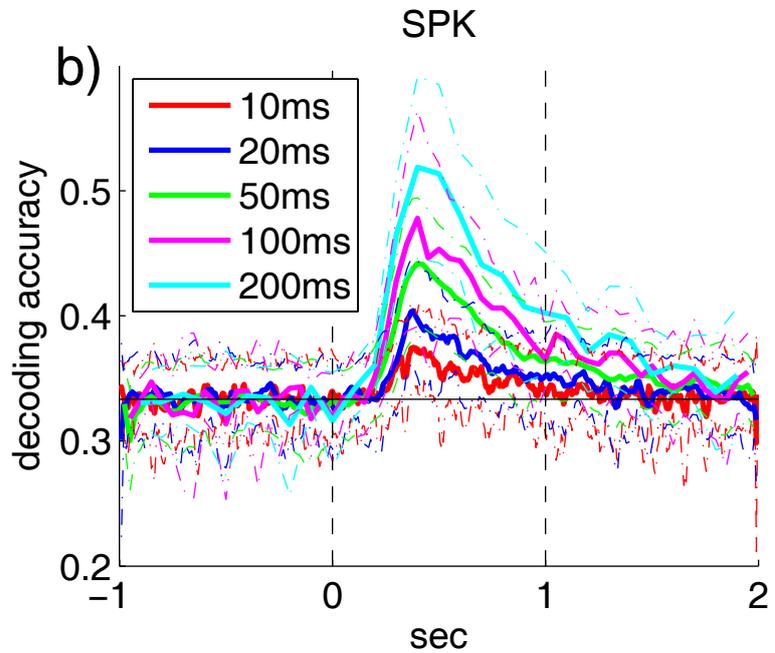
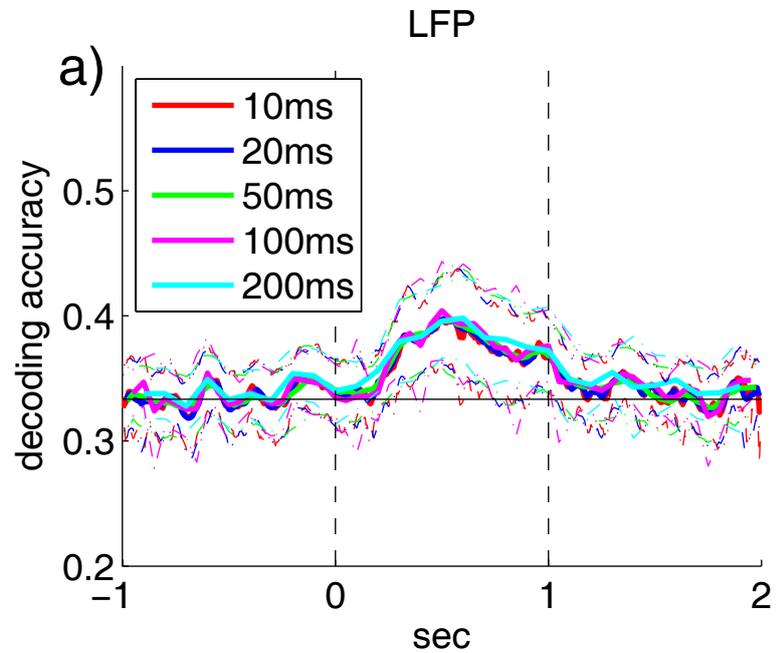
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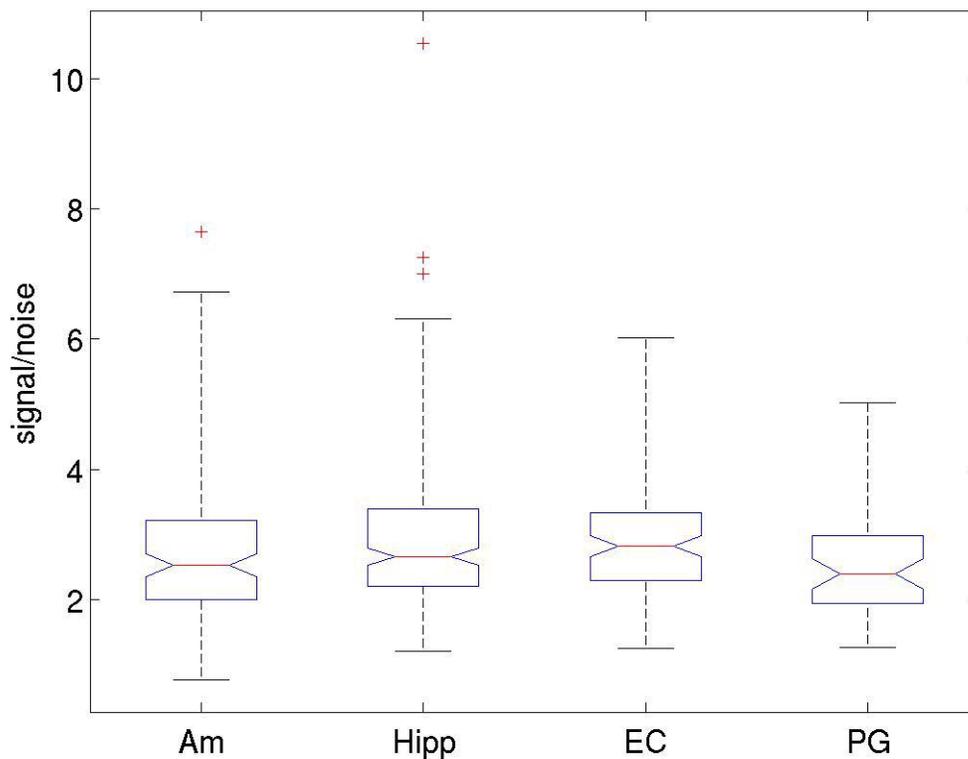




Supplementary material

Signal-to-noise ratio in different brain areas

One possible explanation for the fact that 30 out of 36 LFPs selective microelectrodes were found in the hippocampus could be that because of its anatomical structure, the signal to noise ratio in the hippocampus is better than in the amygdala. We statistically compared signal to noise ratios for evoked potentials in different brain areas. Signal-to-noise ratio was estimated for each microelectrode as the ratio of the maximum amplitude of the evoked potential to its standard error of the mean. Mean values of signal-to-noise ratios in the hippocampus and entorhinal cortices were found to be higher than in the amygdala and parahippocampal gyrus. But the difference was not statistically significant (t-test, $p > 0.05$ for hippocampus vs amygdala, and hippocampus vs entorhinal cortex, and was only slightly significant for hippocampus vs parahippocampal gyrus ($0.01 < p < 0.05$). Therefore, the difference between signal to noise ratios in different brain areas cannot explain the larger selectivity of LFPs found in the hippocampus than in any other area.



Boxplot of the signal to noise ratios in different brain areas. Horizontal red lines represent medians of distributions of signal to noise ratios.

Numbers of responsive LFPs and units to different categories.

For spiking activity a t-test comparison with the firing rate defined during the baseline interval [-1000:-300] msec was used as the test to define the responsive category. For LFPs we applied a sample-by-sample t-test comparing the distribution of LFPs values for

each category and each time point during stimulus presentation, with the distribution of all LFPs values during the baseline interval. The LFPs was defined to be responsive to a category if the t-test trace crossed a significance level of $p=0.001$ for at least 15 consecutive points (see Fig. S1). The following tables provides information about the relative number of responsive units and LFPs across different categories as well as different brain areas.

LFP	animals	Faces	places
All	24 (67%)	35 (97%)	14 (39%)
AM	3	2	1
Hipp	21	30	13
EC	0	3	0
PG	0	0	0

SPIKES	animals	Faces	places
All	32 (48%)	31 (47%)	26 (39%)
AM	17	15	12
Hipp	10	7	8
EC	1	7	1
PG	4	2	5

Note that the same unit might show a significant increase in firing rate for one category and a significant decrease for another one (see, e.g. Fig. 2). For the LFP we use a *time resolved* measure, therefore, the LFP from the same microelectrode can respond to several categories but can be different at different time points. For example, the LFPs in Fig. S1 are responsive for all three image categories. Yet at the same time, an ANOVA test shows that the LFP differs among the three image categories at somewhat different times. Thus, in this example the LFP can discriminate among the three categories but at different times.

