Neuronal responses to visual stimuli that are correlated on a millisecond time scale are well documented in several areas of the mammalian visual cortex. This coherent activity often takes the form of synchronous rhythmic discharges ranging in frequency from 20 to 70 Hz. We performed experiments to determine the incidence and properties of this rhythmic activity in the striate cortex of alert cats and to compare this activity to similar data collected in the striate cortex of anesthetized cats. The results demonstrate that optimal visual stimuli evoke robust, locally synchronous, 20–70 Hz oscillatory responses in the striate cortex of cats that are fully alert and performing a visual fixation task. The oscillatory activity is stimulus dependent, largely absent during periods of spontaneous activity, and shows a systematic increase in frequency with increasing stimulus velocity. Thus, the synchronous oscillatory activity observed in this and earlier studies cannot be explained as an artifact of anesthesia nor as an phenomenon that occurs independent of visual stimulation. Rather, it is a robust process that is present in the alert state and is dependent on the presence and specific properties of visual stimuli.

Key words: striate cortex; physiology; synchronization; oscillation; autocorrelation; gamma-band

In the visual cortex of cats and monkeys, a significant fraction of neurons in striate and prestriate areas exhibit pronounced rhythmicity in their firing patterns in response to optimal visual stimuli presented within their receptive fields (Gray and Singer, 1987, 1989; Eckhorn et al., 1988, 1993; Gray et al., 1990, 1995b; Engel et al., 1991a, b, 1991b; Kreiter and Singer, 1992; Frien et al., 1994; Livingstone, 1996; see also Young et al., 1992; Bair et al., 1994). These stimulus-dependent neuronal oscillations occur mainly in the gamma frequency band (20–70 Hz), are largely absent during spontaneous firing, and are synchronized on a millisecond time scale (i.e., with time lags of ± 5 msec) over a range of spatial scales (Eckhorn et al., 1988, 1993; Gray et al., 1989; Engel et al., 1990, 1991a, b; Frien et al., 1994; Friedman-Hill et al., 1995; König et al., 1995a; Livingstone, 1996). Recent evidence indicates that this type of correlated firing between cells recorded in spatially separate columns occurs most often when one or both of the cells fire rhythmically in the gamma frequency band (Frien et al., 1994; König et al., 1995b; Friedman-Hill et al., 1995; Livingstone, 1996). This common finding suggests a functional link between gamma-band oscillations and response synchronization.

Until recently (Frien et al., 1994; Friedman-Hill et al., 1995), the bulk of the evidence for gamma-band oscillations and response synchronization has come from studies in the anesthetized/paralyzed cat. This has raised the issue that these phenomena may reflect an artifact of anesthesia, or may have different properties in the alert animal. In an earlier report, Raether et al. (1989) demonstrated synchronous oscillatory activity in area 17 of the alert cat. The properties of the activity, including frequency, phase and dependence on orientation, and ocular conditions of stimulation, were found to be similar to those observed in the anesthetized/paralyzed cat (Gray and Singer, 1989; Gray et al., 1990; Engel et al., 1990). The results, however, were taken from a very small number of recordings, and no provision was made to monitor or control the position of the eyes of the animal during visual stimulation. For these reasons, a systematic evaluation of the gamma-band activity in the alert animal was not possible, and the contribution of eye movements to the activity patterns could not be determined.

In the present study, we have recorded the activity of neurons in area 17 of three alert cats trained to fixate a central spot for up to 3 sec. We find oscillatory firing patterns present in a significant fraction of the recorded cells. The properties of this activity, including its frequency distribution and amplitude, are similar to those observed in the anesthetized/paralyzed cat, and contrary to an earlier report (Ghose and Freeman, 1992), we find that gamma-band activity is clearly stimulus dependent.

Parts of this paper have been published previously in abstract form (Gray and Viana Di Prisco, 1993).

MATERIALS AND METHODS

Surgery: alert cats. Three adult cats (two spayed females and one neutered male) were used in the present study. Before behavioral training, the animals were given surgical implants, using sterile technique, of a headpost for restraint, and a scleral search coil for eye position monitoring. The animals were given a preanesthetic cocktail of ketamine (15 mg/kg) and xylazine (1 mg/kg), and maintained under surgical anesthesia throughout the procedure using halothane (1–2%) and oxygen. Body temperature was maintained at 38°C, and the EKG and breathing rate were monitored continuously. The eye coil was implanted first using the technique of Judge et al. (1980). After this procedure, the animal was mounted stenally in a stereotaxic frame, and an incision was made along the midline over the skull. The periosteum was removed, and 6–10

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Stimulus-Dependent Neuronal Oscillations and Local Synchronization in Striate Cortex of the Alert Cat

In the visual cortex of cats and monkeys, a significant fraction of neurons in striate and prestriate areas exhibit pronounced rhythmicity in their firing patterns in response to optimal visual stimuli presented within their receptive fields (Gray and Singer, 1987, 1989; Eckhorn et al., 1988, 1993; Gray et al., 1990, 1995b; Engel et al., 1991a, b; Kreiter and Singer, 1992; Frien et al., 1994; Livingstone, 1996; see also Young et al., 1992; Bair et al., 1994). These stimulus-dependent neuronal oscillations occur mainly in the gamma frequency band (20–70 Hz), are largely absent during spontaneous firing, and are synchronized on a millisecond time scale (i.e., with time lags of ± 5 msec) over a range of spatial scales (Eckhorn et al., 1988; Gray et al., 1989; Engel et al., 1990, 1991a, b; Frien et al., 1994; Friedman-Hill et al., 1995; König et al., 1995a; Livingstone, 1996). Recent evidence indicates that this type of correlated firing between cells recorded in spatially separate columns occurs most often when one or both of the cells fire rhythmically in the gamma frequency band (Frien et al., 1994; König et al., 1995b; Friedman-Hill et al., 1995; Livingstone, 1996). This common finding suggests a functional link between gamma-band oscillations and response synchronization.

Until recently (Frien et al., 1994; Friedman-Hill et al., 1995), the bulk of the evidence for gamma-band oscillations and response synchronization has come from studies in the anesthetized/paralyzed cat. This has raised the issue that these phenomena may reflect an artifact of anesthesia, or may have different properties in the alert animal. In an earlier report, Raether et al. (1989) demonstrated synchronous oscillatory activity in area 17 of the alert cat. The properties of the activity, including frequency, phase and dependence on orientation, and ocular conditions of stimulation, were found to be similar to those observed in the anesthetized/paralyzed cat (Gray and Singer, 1989; Gray et al., 1990; Engel et al., 1990). The results, however, were taken from a very small number of recordings, and no provision was made to monitor or control the position of the eyes of the animal during visual stimulation. For these reasons, a systematic evaluation of the gamma-band activity in the alert animal was not possible, and the contribution of eye movements to the activity patterns could not be determined.

In the present study, we have recorded the activity of neurons in area 17 of three alert cats trained to fixate a central spot for up to 3 sec. We find oscillatory firing patterns present in a significant fraction of the recorded cells. The properties of this activity, including its frequency distribution and amplitude, are similar to those observed in the anesthetized/paralyzed cat, and contrary to an earlier report (Ghose and Freeman, 1992), we find that gamma-band activity is clearly stimulus dependent.

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stainless steel skull screws (4-40, 3⁄16"
3240
containing 2.5% dextrose was given throughout the experiment (4 ml /
(1 mg / kg) and given atropine (0.05 mg / kg s.c.) to reduce salivation. The
anesthetized with an i.m. injection of ketamine (12 mg / kg) and xylazine
was begun. 
recover from the effects of surgery before electrophysiological recording
the first surgical procedure. The animals were again given 10 –14 d to
the skull with 4 – 6 additional screws and dental acrylic. Induction and
of the area centralis of area 17. A recording chamber, fashioned out of
omtory (5
3
screws using a manipulator mounted on the stereotaxic frame, and the
posterior to the frontal sinus. The head bolt was positioned over the skull
frame, and a small craniotomy was made in one hemisphere over the

behavioral training. Once the training was
1.0 –1.5° and 3 sec, respectively. The performance of the animals on this
task varied widely, and it was difficult to tell at the outset which animals
would be the quickest to learn. One animal learned the basic fixation task
within 2 weeks, whereas another took up to 3 months.

The third phase of training consisted of teaching the animals to maintain their gaze within the window even when the bars moved across the fixation point. Once the animals were performing at approximately 80% correct, the second surgery was conducted to implant the recording chamber. Overall behavioral performance varied between animals, but we generally found that a given animal would work for 100–250 correct trials over a period of 3hr.

Electrophysiological recording and visual stimulation. The techniques for unit recording and visual stimulation using anesthetized animals have been published previously (Gray et al., 1990; Maldonado and Gray, 1996) and were similar to those used with the alert animals. We therefore limit our description of the recording and stimulation methods to those used for the behavioral experiments.

Once the animals were fully trained to perform the fixation task, electrophysiological recordings were carried out on a daily basis. We used two slightly different techniques for the manipulation of our recording electrodes. For both techniques, the electrodes (1–2 MΩ, tungsten) were advanced using miniature micromanipulators available from Biela Engineering (Malpeli et al., 1992). These manipulators, approximately 5 mm in diameter and 20–30 mm in length, were small enough to easily enable the insertion of the recording chamber within the area centralis. We therefore limit our description of the recording and stimulation methods to those used for the behavioral experiments.

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stable unit recordings obtained, and the receptive fields mapped before data could be collected. This sequence often required 30–40 min and 30–40 behavioral trials, thereby reducing the available time for collecting data. In an attempt to improve the yield, we resorted to leaving the electrodes in place at the end of an initial recording session. This proved to be a rather simple procedure, and the manipulators were protected by a hard cap made out of Delrin that we fixed to the recording chamber with screws. On subsequent days, the setup procedure was now reduced to 5–10 min. The protective cap was removed and the electrode connector plugged in. A quick check could be made to see whether unit activity was still present at the recording site, and if not, the manipulator was advanced slowly until new units were isolated. Because of the columnar and retinotopic organization of the striate cortex, the time needed to map the receptive fields of the new cells was dramatically reduced. Overall, this technique greatly improved the yield of the experiments and seems to be essential when using cats as subjects because of the limited time available to record.

In addition to the simple sequences of optimal bar stimulation, we also investigated the influence of changes in stimulus velocity on the properties of oscillatory responses. In a subset of sessions, we performed velocity tuning curves using 3–6 different velocities ranging from 0.9/sec to 12/sec presented in a pseudorandom order. The range of velocities chosen covered the preferred velocity for any given recording. All stimuli ranged in luminance from 0.1 to 20 cd/m² and were presented on a dark background.

Data analysis. The amplitude and time-dependent properties of the recorded spike trains were evaluated using several quantitative measures. First, we computed a perstimulus time histogram (PSTH) for each recording using a bin width of 50 msec. The mean firing rate was computed from each recording during two epochs, one before stimulus presentation during spontaneous activity, and the second during the period of stimulus presentation. We refer to these two epochs as window 1 (win1) and window 2 (win2), respectively. The duration of these windows ranged from 0.6–2 sec. Tuning curves were constructed by computing both the mean and peak firing rate within each of the two windows. To plot the results as a function of stimulus number.

To evaluate the time-dependent properties of the spike trains, we computed the autocorrelation histogram (ACH) on the activity present within both time windows. This was done for individual trials, and the sum of the activity recorded across all trials. As a control for periodicities in the spike trains introduced by the computer monitor (80 Hz noninterlaced refresh rate), we also computed the cumulative autocorrelation histogram across trials after shuffling the trial sequence by one stimulus period (i.e., the shift predictor). Any stimulus-locked frequencies present in the spike train would be present in the shift predictor correlogram. We used an additional measure to compare the resulting correlograms with those that are computed from an equivalent random process. For the data collected on each trial within each of the two windows, we generated a spike train having a pseudorandom distribution of intervals with an equivalent number of spikes at the same mean firing rate. These intervals, serving as a control, were subjected to autocorrelation analysis in exactly the same manner as the experimental data. From the resulting correlograms, we computed the mean and SD across all bins and used these values to estimate the confidence limits for significant peaks and/or troughs in the experimental correlograms. For all the correlogram calculations, we used positive and negative time lags of both 128 msec and 256 msec at a bin width of 1 msec.

To evaluate the temporal structure of each ACH (single trial, across trials, shift predictor, random control, win1 and win2), we computed the power spectrum of the ACH using the fast fourier transform (Press et al., 1988). The amplitude and time-dependent properties of the power spectrum were also estimated. The peak value and the ratio of the peak value to the DC component (peak/DC) provide the first two measures of the spectra. These values were considered significant if the peak between 20–70 Hz exceeded the mean plus three SDs of the corresponding control power spectrum (excluding the DC component) computed from the ACH of the pseudorandom spike train. As a third measure of the data, we computed the signal to noise ratio (SNR) of the power spectra (Ghose and Freeman, 1992). The signal was computed by taking the average over an 8 Hz window centered on the peak in the spectrum between 20–70 Hz. The noise for each spectrum was computed by taking the average of all the values between 250 and 500 Hz.

For each of these three measures, we examined the cumulative distributions for correlation histograms computed across trials and on individual trials. Because single trials often did not yield a sufficient number of spikes to reliably employ autocorrelation analysis, we devised a simple criterion for the number of spikes needed, below which the correlograms were excluded from the analysis. For each correlogram (across trials and individual trials), the mean of the corresponding pseudorandom control correlogram had to be equal to or greater than twice its SD. Although this criterion was arbitrary, it served to provide a discrete cutoff for acceptance that was directly related to the criterion for statistical significance.

RESULTS
Incidence, magnitude, and stimulus dependence of oscillatory activity

The main focus of this report is the analysis of oscillatory activity in the striate cortex of the alert cat. Because the properties of this activity in the anesthetized cat have been documented extensively (Gray and Singer, 1989; Gray et al., 1989, 1990; Engel et al., 1990), the data collected from anesthetized animals in this study are used primarily for the purpose of comparison. Our aim in this regard is to compare the incidence of occurrence, the magnitude, and the frequency distribution of oscillatory activity under the two conditions. Our sample consisted of 128 multunit and 17 single unit recordings from 3 alert cats, and 100 multunit and 48 single unit recordings from 11 anesthetized cats.

An example of multunit activity recorded from an alert cat during the presentation of an optimally oriented, drifting light bar is shown in Figure 1. The animal maintained fixation for 2.4 sec and the stimulus was presented at a latency of 0.6 sec after the start of fixation. Initially, the light bar appeared outside the receptive field and then passed through it. Therefore, the peak of the response occurred roughly between 1 and 2 sec into the trial. The timing of the behavioral task, shown in the middle portion of the figure, allowed us to select two epochs of data for further analysis. These epochs, termed win1 and win2, correspond to periods of spontaneous and stimulus-evoked activity. The quantitative analysis of these data are shown in Figure 2. The ACH was computed across all 10 trials in the session for win1 (Fig. 2A), win2 (Fig. 2C), and the shift predictor control of win2 (Fig. 2E). The horizontal lines in Figure 2, A, C and E, represent the mean and the mean ± 2 SDs (i.e., 95% confidence limits) of the autocorrelation histogram computed from an equivalent pseudorandom spike train. In all 3 histograms, the mean is greater than twice the SD, although in Figure 2C by only a slight margin. Thus, each of these data sets met our criteria for sufficient numbers of spikes. Any data set not meeting this criteria was excluded from the cumulative results. The power spectra of the corresponding ACHs are shown in Figure 2, B, D, and F. The thick and thin lines in these graphs represent the mean and the mean + 3 SDs (99% confidence limit) of the power spectra computed from the equivalent pseudorandom spike trains. These data illustrate several points; statistically significant rhythmic firing is present during the response to the stimulus and absent during spontaneous firing. The absence of any peaks in the shift predictor control demonstrates that the oscillations are not time-locked to the stimulus. This latter result excludes the possibility that the rhythmic firing is driven by the video refresh (80 Hz) of the computer monitor.

In the same recording, we found that oscillatory activity was easily detectable on single trials. Figure 3 shows three trials of activity, recorded during a separate trial block, at the same site as the data shown in Figures 1 and 2. The spike trains, along with the vertical and horizontal eye position, are shown in Figure 3A. The receptive field of the cells was located at an eccentricity of 1.8°,
and the bar passed over the center of gaze while moving across the receptive field. Thus, even though small saccades and slow drifts in eye position of roughly 0.5° are evident, the animal was able to suppress a smooth pursuit eye movement during the stimulus presentation. The ACHs and associated power spectra computed for each trial during win2 are shown in Figure 3, respectively. The top plot is a PSTH of multiunit activity recorded across 10 consecutive trials. win1 and win2 illustrate the epochs selected for correlation analysis and reflect periods of spontaneous and stimulus-evoked activity, respectively. The bottom traces are raster plots of the 10 spike trains.

To quantify the properties of the rhythmic firing, we computed the peak/DC value, the frequency at the peak, and the S/N (see Materials and Methods). These values are shown to the right of each spectrum in Figures 2 and 3, respectively. Strong oscillatory activity near 40 Hz is present on each trial. This indicates that oscillatory firing among local groups of cells is highly synchronous and can be analyzed on a trial by trial basis from short epochs of data.

We applied these calculations to the cumulative ACHs for both the alert and anesthetized data sets. The results for the peak/DC and the peak frequency values computed during win2 are shown than twice the SD. Second, the peak in the power spectrum had to exceed the mean + 3 SDs of the power spectrum computed from the pseudorandom control. Third, the peak/DC value had to exceed the value at which 90% of all the peak/DC values, computed from the trial-shuffled correlograms, fell below. For the S/N values, we used the cutoff value of 1.5 as proposed previously by Ghose and Freeman (1992). Thus, spectra exhibiting significant oscillatory firing during the stimulus presentation. For the S/N values, we used the cutoff value of 1.5 as proposed previously by Ghose and Freeman (1992). Thus, spectra having an S/N value greater than 1.5 were considered significant. In contrast to the peak/DC measure, the S/N measure indicated that significant oscillatory firing was present in all three conditions of Figure 2 and for all three trials of Figure 3.

Figure 1. The behavioral paradigm used for recording unit activity from an alert cat trained to visually fixate a central target. The middle four traces schematically illustrate the temporal sequence of events that occur during a single trial. s, t, f, and r represent the time of occurrence of the stimulus, the period of data acquisition, the fixation spot, and the food reward, respectively. The top plot is a PSTH of multiunit activity recorded across 10 consecutive trials. win1 and win2 illustrate the epochs selected for correlation analysis and reflect periods of spontaneous and stimulus-evoked activity, respectively. The bottom traces are raster plots of the 10 spike trains.

Figure 2. Autocorrelation histograms (A, C, E) and corresponding power spectra (B, D, F) computed from the multiunit recording shown in Figure 1 during periods of spontaneous (A, B) and stimulus-evoked (C, D) activity. E and F were computed from the data collected during win2 after shuffling the spike trains by one stimulus period. Note the complete absence of significant peaks in the histogram in E. This indicates that the oscillatory activity is not phase-locked to the stimulus. The three horizontal lines passing through the autocorrelation histograms represent the mean and 95% confidence limits computed from the trial shuffled control (i.e., shift predictor). The thick and thin lines shown in the power spectra represent the mean and 99% confidence limit, respectively, for an equivalent random spike train. In this and subsequent figures, the three values shown to the right of each spectrum are the peak/DC value, the frequency of the peak, and the S/N, respectively.
in Figure 4 and Table 1. The S/N values for win2 are shown in Figure 5. Using the criteria described above for the peak/DC values, we found that 38 of 144 (26.4%) recordings in alert animals and 25 of 148 (16.9%) recordings in anesthetized animals showed significant oscillations. Analysis of the distributions in Figure 4, A and C, revealed that the magnitude of the oscillatory activity was greater in the alert (0.127 ± 0.057) than in the anesthetized (0.089 ± 0.022) state (p < 0.001, Mann–Whitney U test). Although this difference is significant, it is clear from Table 1 that differences in the sample size from each animal as well as individual variation in the incidence and magnitude of oscillatory activity could account for this result. The mean frequencies in the alert and anesthetized states were 41.4 ± 9.4 Hz and 39.7 ± 11.8 Hz, respectively. Neither the mean values nor the variance in peak frequency differed between the anesthetized and alert conditions.

In contrast to these results, we found that the criterion value of 1.5 for the S/N measure was much more sensitive than the peak/DC measure (Fig. 5). This analysis detected significant oscillations in 122 of 144 (85%) and 120 of 148 (81%) recordings in the alert and anesthetized animals, respectively. We suspected that this high percentage of significant values might result from the fact that our autocorrelation calculations used a time lag of 6128 msec. We therefore recomputed the S/N values using a correlation time lag of 6256 msec (Ghose and Freeman, 1992). With this calculation, we found that 104 of 144 (72%) recordings showed significant oscillations in the alert state. The greater sensitivity of the S/N compared to the peak/DC calculations seems to be a result of the arbitrary selection of 1.5 as a criterion for statistical significance. We examined the distribution of S/N values for both data sets obtained from those recordings that met the criteria for significant peak/DC values and found that many of the spectra having low (<8), as well as high (>8), S/N values did...
not meet the criteria applied to the peak/DC values. This indicates that many spectra having S/N values $> 1.5$ contained peaks that were no greater than those that would be expected to occur by chance. Thus, the S/N measure is subject to a large percentage of false positive detections of oscillatory activity.

After applying a similar analysis to the data collected during spontaneous firing (win1), we found that the incidence of significant oscillations was much lower than that recorded in response to visual stimulation. This was, in large part, a result of the low spontaneous firing rate of most cells in striate cortex. In the alert and anesthetized states, respectively, 4 of 57 and 4 of 28 recordings, meeting the criteria for sufficient numbers of spikes, displayed significant oscillations. The frequency of this activity was slightly lower than that observed during visual stimulation (alert = 30 ± 9.5 Hz; anesthetized = 29 ± 2 Hz anesthetized).

We also applied the peak/DC analysis to the correlation histograms computed on individual trials. The results of these calculations are shown in Figure 6A,C. In the alert animals, 977 of 2115 trials met the criteria for sufficient numbers of spikes, whereas in the anesthetized animals, this value was 861 of 1975 total trials. Of these, significant peak/DC values were detected on 393 trials in the alert animals ($40.2\%$; mean = 0.144 ± 0.081) and 245 trials in the anesthetized animals ($28.5\%$; mean = 0.117 ± 0.045). As with the cumulative correlograms, the magnitude of the oscillatory activity was greater in the alert state ($p < 0.0001$, Mann–Whitney U test). The distribution of peak frequencies (Fig. 6B,D) revealed that oscillation frequency was similar for the alert $[41.4 ± 10.1 \text{ Hz} (n = 393)]$ and anesthetized $[40.9 ± 12.3 \text{ Hz} (n = 245)]$ states, although the frequency variance was greater in the data obtained from the anesthetized animals ($p < 0.0006$, F test).

During spontaneous firing, only 10 of 57 trials (alert data) and 4 of 20 trials (anesthetized data) having sufficient numbers of spikes displayed significant oscillations. Again, these low numbers stemmed largely from the very low firing rates of cells during spontaneous activity.

Because the oscillation frequency varies within as well as across trials (Gray et al., 1990, 1992), and this variance would be expected to average out the oscillatory components in the cumulative autocorrelograms, we suspected that our measure of the incidence of oscillatory activity (Fig. 4) underestimated the number of recordings that displayed oscillatory firing. To test this conjecture, we determined the percentage of trials at each recording site that displayed significant oscillatory modulation. The results of these calculations, shown in Figure 7, revealed that greater than 50% of the cells, in both the alert and anesthetized animals, exhibited significant oscillatory responses to visual stimuli on at least 10% of the trials. This finding has several implications. First, it indicates that the cumulative ACH underestimates the incidence of oscillatory activity. Second, it demonstrates that neuronal oscillations are episodic, nonstationary events. They may vary in frequency or in probability of occurrence or both. Thus, the failure to find oscillatory modulation in a cumulative ACH may result from too few trials displaying oscillations, or from a variation in oscillation frequency that smoothes the cumulative histogram.

The example of both of these effects recorded in the alert animals are shown in Figures 8 and 9, respectively. Figure 8 shows an example where the cumulative ACH (Fig. 8A,B), but none of the 20 individual trial histograms (Fig. 8C,D), met the criteria for significant oscillations. In those trials where it is possible to see oscillatory modulation (i.e., 3, 7, 10, 11, 13, and 16), the

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Table 1. Distributions of the incidence, magnitude, and frequency of statistically significant oscillatory responses for each animal

<table>
<thead>
<tr>
<th></th>
<th>No. of cells</th>
<th>Mean peak/DC</th>
<th>Mean frequency (Hz)</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Behavioral</strong></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Cat 1</td>
<td>2 of 17 (11.7%)</td>
<td>0.095 ± 0.03</td>
<td>36 ± 0</td>
</tr>
<tr>
<td>Cat 2</td>
<td>4 of 44 (9.1%)</td>
<td>0.0829 ± 0.0124</td>
<td>47 ± 6.8</td>
</tr>
<tr>
<td>Cat 3</td>
<td>32 of 83 (38.5%)</td>
<td>0.1345 ± 0.0592</td>
<td>41 ± 9.8</td>
</tr>
<tr>
<td>Total</td>
<td>38 of 144 (26.4%)</td>
<td>0.1270 ± 0.0573</td>
<td>41.4 ± 9.4</td>
</tr>
<tr>
<td><strong>Anesthetized</strong></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Cat 1</td>
<td>1 of 15 (6.7%)</td>
<td>0.0958</td>
<td>32</td>
</tr>
<tr>
<td>Cat 2</td>
<td>1 of 7 (14.3%)</td>
<td>0.0783</td>
<td>28</td>
</tr>
<tr>
<td>Cat 3</td>
<td>3 of 10 (30%)</td>
<td>0.0775 ± 0.01</td>
<td>40 ± 13.8</td>
</tr>
<tr>
<td>Cat 4</td>
<td>7 of 11 (63.6%)</td>
<td>0.0905 ± 0.0235</td>
<td>39.4 ± 5.8</td>
</tr>
<tr>
<td>Cat 5</td>
<td>3 of 12 (25%)</td>
<td>0.0882 ± 0.0362</td>
<td>42.7 ± 18.9</td>
</tr>
<tr>
<td>Cat 6</td>
<td>0 of 15 (0%)</td>
<td>0.0942 ± 0.0189</td>
<td>47.2 ± 13.7</td>
</tr>
<tr>
<td>Cat 7</td>
<td>5 of 25 (20%)</td>
<td>0.0836 ± 0.0142</td>
<td>28 ± 4</td>
</tr>
<tr>
<td>Cat 8</td>
<td>2 of 15 (13.3%)</td>
<td>0.0896 ± 0.0218</td>
<td>39.7 ± 11.8</td>
</tr>
<tr>
<td>Cat 9</td>
<td>0 of 6 (0%)</td>
<td>0.0989</td>
<td>32</td>
</tr>
<tr>
<td>Cat 10</td>
<td>1 of 20 (5%)</td>
<td>0.1168 ± 0.0412</td>
<td>32 ± 16.9</td>
</tr>
<tr>
<td>Cat 11</td>
<td>2 of 12 (16.7%)</td>
<td>0.1289 ± 0.0218</td>
<td>39.7 ± 11.8</td>
</tr>
<tr>
<td>Total</td>
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</tbody>
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number of spikes was insufficient to meet the criterion. The remaining trials have either too few spikes or no rhythmic modulation. Despite these sources of variance, the cumulative ACH was highly significant. This is because the oscillatory activity, when it did occur, was relatively constant in frequency. Thus, summing the individual trials did not lead to an averaging out of the rhythmic modulation. Figure 9 shows a different example in which 8 of 10 single trial histograms (Fig. 9C,D), but not the cumulative ACH (Fig. 9A,B), met the criteria for statistical significance. Here it is clear that the variation of oscillation frequency accounts for the low modulation amplitude of the cumulative ACH. An example of this can be clearly seen by comparing trials 4 and 8 in Figure 9C,D. Thus, frequency variance, in addition to probability of occurrence, can account for the failure to observe significant oscillations.

Our effort to evaluate the occurrence of oscillatory firing during spontaneous activity was limited by the low firing rates in the absence of visual stimulation and, hence, a failure to collect sufficient numbers of spikes at many of the recording sites. We therefore changed our strategy during the experiments and ran sessions in which we interleaved stimulus trials with trials containing only a fixation spot, or in which we varied the stimulus luminance. This allowed us to sample spontaneous activity for longer periods of time and to compare its temporal properties to that evoked during visual stimulation. Figure 10 shows one example in which oscillatory activity evoked during visual stimulation (Fig. 10B,D,F) is absent during a comparable period of spontaneous firing while the animal is fixating (Fig. 10A,C,E). Similarly, Figure 11 illustrates a recording in which significant oscillations were absent at low levels of stimulus luminance but became highly significant as the response magnitude approached saturation. These examples were consistent with our sample of 14 multiunit recordings that displayed significant oscillatory responses to an optimal visual stimulus (mean = 541 Hz), whereas in the absence of visual stimulation these cells showed significant oscillatory modulation in only 2 of the 14 cases (mean = 34 Hz). Thus, in the absence of visual stimulation, rhythmic firing occurs less often and at a lower frequency among cells that display pronounced oscillations during visual stimulation.

We obtained further evidence for the stimulus dependence of oscillatory activity by comparing the frequency of oscillations at different velocities of stimulus movement. It has been shown previously in anesthetized animals (Eckhorn et al., 1988; Gray et al., 1990) that oscillation frequency increases with stimulus velocity. Our results in the alert cats were consistent with these earlier reports. In 11 of 12 multiunit recordings, we found that the frequency of significant oscillations evoked by a moving light bar increased monotonically with the velocity of stimulus movement. An example of this result, from a multiunit recording in area 18, is shown in Figure 12. A scatter plot of the cumulative data is shown in Figure 13. In this sample, significant oscillations spanned a frequency range of 36–60 Hz and occurred at stimulus velocities ranging from 1.4 to 11.8°/sec. Linear regression analysis revealed a significant correlation between stimulus velocity and oscillation frequency variance, in addition to probability of occurrence, can account for the failure to observe significant oscillations.
frequency \( r = 0.71, p \ll 0.0001 \) with frequency increasing by 1.5 Hz for each 1°/sec increase in stimulus velocity. These results closely resemble those obtained in the anesthetized cat (Gray et al., 1990) and further demonstrate that the properties of oscillatory responses in cat striate cortex are stimulus dependent.

**Local synchronization**

In a subset of our multiunit recordings that displayed significant oscillations, we utilized spike sorting techniques (Abeles and Goldstein, 1977; Gray et al., 1995a) to enable us to analyze
the local temporal correlation among pairs or small groups of neurons. In some recordings, we were able to identify a pair of single units using principal components analysis, whereas in others, we extracted a single unit and a multunit background signal using amplitude discrimination. We then performed cross-correlation analysis on these pairs of signals and applied the same statistical test for significance as that applied to the autocorrelation histograms. Figure 14 illustrates four examples of the raw data collected on single trials from single and/or multunit recordings that were typical of the data base as a whole. Figure 14A illustrates a single unit that exhibits a pattern of repetitive burst discharges that we refer to as “chattering” (Gray and McCormick, 1996). This pattern, which is also prevalent in area 17 and 18 of the anesthetized cat and area 17 of the alert monkey (Hubel and Wiesel, 1965; Gray et al., 1990, 1995b), consists of sequences of burst discharges that repeat at intervals of 15–50 msec during the response to a visual stimulus. Within each burst, the firing rate ranges from roughly 300 Hz to as high as 800 Hz. This pattern gives rise to the pronounced oscillations observed in the autocorrelation histograms and is often locally synchronized as shown in Figure 14, B and D. In Figure 14B, a chattering cell can be seen to fire with a slight but consistent phase lead over the lower amplitude multunit background activity. In Figure 14D, there are several units firing in close temporal correlation. Such local synchronization does not, however, occur exclusively among cells that fire in repetitive bursts. Figure 14C illustrates an example in which a single unit that displays no evidence of bursting or oscillation is tightly synchronized with the oscillatory multunit activity recorded on the same electrode. This demonstrates that nonoscillating cells do participate among populations of neurons that display synchronous rhythmic firing.

Figure 15 illustrates an example of the local synchronization of oscillatory firing displayed by two cells recorded on the same electrode. These data were taken from the same recording shown in Figures 1–3 and demonstrate that oscillations in multunit activity reflect the temporal synchronization of the underlying single units. The oscillations observed in each single unit are of the same frequency as the multunit recording in Figure 3, and the two cells have an average time lag of 0 msec. Thus, the prevalence of significant rhythmicity in the discharges of multunit recordings is indicative of a high degree of local temporal synchronization.

Figure 9. Example of a multunit recording in which the cumulative autocorrelation histogram did not meet the criteria for significance (i.e., the peak/DC value is 0.06). A, Cumulative autocorrelation histogram. B, Corresponding power spectrum. The autocorrelation histograms and associated power spectra for each of the 10 trials recorded on this session are shown in C and D, respectively. Eight of the 10 trials in this session met the criteria for significance. These are indicated by an asterisk to the right of the power spectra in D.
DISCUSSION

Methodological considerations

The occurrence and properties of stimulus-dependent oscillations in the visual cortex has been the subject of considerable debate. Much of this debate has focused on the probability of occurrence, the magnitude and statistical significance of neuronal oscillations. However, each of these parameters depends on the presence or absence of a sampling bias, the statistical test used to classify the data, and the criteria for what constitutes an oscillation. Concerning sampling bias, there are at least three possible sources of error, variation in the properties of oscillatory activity between animals, and inadequate or experimenter-biased sampling from individual animals. This study is subject to the first two of these sources of variation, which could account for the differences in the incidence and magnitude of oscillations between the alert and anesthetized states. However, we believe that the present data represent our best effort to avoid subjective biases in sampling the neuronal activity during data collection. Because of the constraints of recording from alert cats, we made a deliberate effort to collect every high quality recording we could obtain. This precluded us, for example, from searching for activity that gave more vigorous responses and were easier to map. In the anesthetized animals, we made electrode penetrations perpendicular to the cortical surface and sampled any multiunit or single unit activity that could be visually driven with a drifting light bar or square wave grating.

The statistical tests used to measure and classify neuronal oscillations are another source of variability in results from different studies. The majority of investigations have used the autocorrelation histogram as the initial step in the quantification of rhythmicity (Gray and Singer, 1989; Gray et al., 1990; Engel et al., 1990; Ghose and Freeman, 1992; Young et al., 1992). The methods used to quantify the correlograms have differed, however. Our previous method of fitting a damped sine wave function to the autocorrelation histogram and then extracting the parameters of that function (Engel et al., 1990; König et al., 1994) have been criticized as being subject to error (Ghose and Freeman, 1992; Young et al., 1992). In an effort to address these criticisms and to make our results comparable to those of others, we applied the method of Ghose and Freeman (1992) of calculating the autocorrelation histogram, computing the power spectrum, and calculating a S/N of the peak in the spectrum between 20–70 Hz relative to the average spectral values between 250–500 Hz. Ghose and Freeman (1992) arbitrarily chose a S/N value of 1.5 as indicative of significant periodicity in the ACH. We applied this measure to our data and found that it classified a much greater percentage of

Figure 10. Oscillatory neuronal activity is present during visual stimulation but absent during periods of spontaneous firing. The left column illustrates the data obtained during fixation trials, whereas the right column illustrates the activity recorded during the presentation of an optimally oriented light bar. A, B, PSTHs. C, D, Autocorrelation histograms computed during the period of stimulus presentation. In C, this period ranged from 0.2 to 1.9 sec. E, F, Power spectra of the autocorrelation histograms shown in C and D, respectively. Note the presence of a clear peak at 44 Hz in F that is not present in E.

Figure 11. The magnitude of stimulus-evoked oscillatory firing increases with stimulus luminance. An optimally oriented light bar was passed over the receptive field of the recorded cells at five different levels of luminance. A, PSTHs computed from the spike trains recorded at each level of luminance. B, Tuning curve computed from the mean firing rate during the period of stimulus presentation. The plot is normalized to the maximum response. C, D, Autocorrelation histograms and corresponding power spectra computed from each of the five sets of spike trains, respectively.
correlograms as oscillatory (Fig. 5) than obtained by previous estimates (Gray et al., 1990; Engel et al., 1990). In addition, we found that the S/N measure did not take into account the total numbers of spikes used to compute the ACH. To avoid this problem, we used a measure of the ratio of the peak in the spectrum to the DC component. To be classified as significant, the peak spectral values had to exceed the 99% confidence limit, computed from the ACH of an equivalent random spike train, and their peak/DC ratios had to be greater than 90% of the peak/DC values computed from the trial-shuffled histograms of the entire sample.

When we applied these relatively stringent criteria for statistical significance, we found a much lower incidence of oscillatory activity than revealed by the S/N measure. This may account for the wide differences in results, and hence interpretations, reported by Ghose and Freeman (1992) and earlier studies (Gray and Singer, 1989; Gray et al., 1990; Engel et al., 1990). In support of this conclusion, we found that S/N values of the recordings classified by the peak/DC measure as significant displayed a wide

Figure 12. The frequency of oscillatory firing in the visual cortex of the alert cat increases with increasing stimulus velocity. Multiunit activity was recorded from a single electrode in area 18, and moving light bars of high contrast were passed over the receptive field of the cell at four different velocities (—4, 8, 12, and 16°/sec). A, PSTHs of the activity recorded in response to each of the 4 stimuli, velocity increasing from stimulus 0 to 3. B, Tuning curve of the oscillation frequency computed from the power spectra shown in D. C, Autocorrelation histograms computed from the activity recorded during the response to each visual stimulus. D, Power spectra of the autocorrelation histograms shown in C.

Figure 13. Scatter plot of the oscillation frequency, obtained from the peaks in the power spectra, versus the stimulus velocity for 11 multiunit recordings in area 17. Three to five different velocities were presented for each recording (n = 34). The linear correlation coefficient is 0.71 (p << 0.0001) and the slope of the fitted line is 1.49 Hz/deg/sec.
Figure 14. Four examples of oscillatory single and multiunit activity recorded on single trials in area 17 of the alert cat in response to an optimally oriented, drifting light bar. In each plot, the activity is displayed at a slow (top) and fast (bottom) time scale. The lines connecting the top and bottom plots indicate the epoch in the top plot that is being displayed in the bottom plot. A, Activity recorded from a single cell that exhibits a pattern of repetitive bursting during the visual response. B, A multiunit recording, consisting of a single isolated unit and lower amplitude background activity, illustrates that the repetitive burst discharges of the single unit are correlated with the multiunit activity. C, Another multiunit recording illustrating that a nonrhythmically firing single unit is synchronous with the local oscillatory multiunit activity. D, A further example of synchronous, rhythmic multiunit activity.
range, and that many S/N values greater than 1.5 did not meet our criteria for statistical significance. By utilizing an unbiased and stringent measure, it is also likely that the present results reflect a more accurate estimate of the incidence and magnitude of oscillatory activity than earlier studies (Eckhorn et al., 1988; Gray and Singer, 1989; Gray et al., 1989, 1990; Engel et al., 1990).

It must be pointed out, however, that the cumulative ACH is not the best method for detecting and classifying neuronal oscillations in the visual cortex. Because neuronal oscillations can and often do exhibit wide variations in their frequency (Gray and Singer, 1989; Gray et al., 1992), summing these variations across trials in the cumulative ACH will tend to average out the periodicity (for example, see Fig. 9). This will lead to an underestimate of the incidence of rhythmic activity. Moreover, the magnitude of this underestimate is likely to increase as the number of trials collected increases and if spontaneous and stimulus-evoked activity are combined in the calculation of the correlation histograms. The traditional approach to correlation analysis has been to enhance the S/N in the data by accumulating large numbers of spikes collected during periods of spontaneous and stimulus-driven activity (Perkel et al., 1967; Ts’o et al., 1986; Aiple and Krueger, 1988). The data presented here demonstrate that this practice is likely to mask rather than enhance the occurrence of rhythmicity in the spike trains.

Finally, because all visual stimulation was performed by computer, it is possible that refresh artifact (firing synchronized to the refresh frequency of the monitor) could account for some percentage of the oscillatory firing detected. Cells showing this behavior exhibit peaks in their correlation and shift predictor spectra having a frequency that is equal to, or a harmonic/subharmonic of, the refresh frequency. It was surprising that we did not find a single example of this effect in our data. It is, therefore, very unlikely that any of the significant spectral peaks observed in the correlograms reflect refresh artifact.

**Functional implications**

The results of this study demonstrate that synchronous gamma-band activity in the visual cortex is not a phenomenon limited to anesthetized animals. It is robust and easily detectable on single trials in alert cats. Although problems of sampling bias preclude a firm conclusion regarding the differences between the alert and anesthetized data, the data presented here suggest that the incidence and magnitude of oscillatory activity is greater in the alert cats. These data are consistent with the recent findings that activation of the mesencephalic reticular formation (Munk et al., 1996) or the mesopontine cholinergic nuclei (Steriade et al., 1996) increase the probability and the magnitude of synchronous gamma-band activity in cat cortex. Thus, arousal seems to be an important variable influencing the occurrence and properties of gamma-band response synchronization.

The stimulus dependence of synchronous gamma-band activity, and its relative absence during spontaneous firing, indicate that
neuronal oscillations are indeed related to visual processing, although in a manner which is not yet fully understood (Frien et al., 1994; Roelfsema et al., 1994; Livingstone, 1996; Kreiter and Singer, 1996). In their earlier study, Ghose and Freeman (1992) concluded that cortical gamma-band activity is not stimulus dependent and that oscillation strength is greatest at low firing rates. These results were complimented by the finding that a significant fraction of neurons in the LGN exhibit pronounced rhythmicity in the absence of visual stimuli, with a frequency distribution overlapping that of the cortical data (Lauffer and Verzeano, 1967). In many of these thalamic neurons, the oscillations were suppressed or uninfluenced by visual stimuli. On the basis of these data, Ghose and Freeman (1992) proposed a model where cortical gamma-band activity is generated by the convergent input of independent but rhythmically firing LGN cells onto cortical neurons. Although this model seems to account for their data, it is inconsistent with at least two general findings presented in this and earlier studies: (1) gamma-band activity has been reported to be largely absent or greatly reduced in magnitude in the absence of visual stimulation (e.g., Figs. 2, 10, 11; Eckhorn et al., 1988; Gray and Singer, 1989; Gray et al., 1990; Engel et al., 1990; Jagadeesh et al., 1992; Gray and McCormick, 1996); and (2) neuronal oscillations in the LGN of the anesthetized cat are stimulus dependent, both locally and globally synchronized (as long as they are driven by the same eye), and have a mean frequency near 80 Hz (Ito et al., 1994; NeuenSwander and Singer, 1996). These data, therefore, lead to the prediction that LGN oscillations should contribute to cortical gamma-band synchronization primarily during the response to visual stimuli. However, any model using this mechanism to generate cortical oscillations must account for the large difference in frequency distribution between the LGN and cortex.

Although the results presented here can shed little light on the underlying mechanisms of cortical gamma-band activity, one salient aspect is consistent with an alternative model for cortical gamma-band activity generation. When neuronal oscillations occur in the visual cortex, the cells contributing to these synchronous discharges often fire in a pattern of repetitive high frequency bursts (e.g., Fig. 14; Hubel and Wiesel, 1965; Gray et al., 1990, 1995b). These bursts recur at intervals of 15–50 msec and can reach firing rates within a burst as high as 800 Hz. Recently, using intracellular recording and staining in vivo, we have demonstrated that cells in striate cortex having these characteristics comprise a subpopulation of superficial pyramidal neurons (chattering cells) that, when injected with suprathreshold depolarizing current, exhibit intrinsic repetitive burst discharges at frequencies ranging from 20–70 Hz (Gray and McCormick, 1996). These cells also exhibit larger stimulus-evoked, gamma-band fluctuations in their membrane potential than any of the other cell classes. These findings suggest that a major component of visually evoked cortical gamma-band oscillations is influenced by the intrinsic membrane properties of chattering cells, and hence is a result of an intracortical mechanism. Cells that receive presynaptic input from chattering cells would, therefore, be expected to display membrane potential oscillations having properties consistent with a synaptic rather than an intrinsic mechanism (Jagadeesh et al., 1992). If chattering cells were synaptically connected to one another, the oscillatory activity in these cells would reflect both mechanisms. In view of the findings that oscillatory firing is a strong predictor of synchronization, these data suggest that chattering cells may play an important role in the generation of synchronous activity in visual cortex.

REFERENCES


