Dynamics of Eye-Position Signals in the Dorsal Visual System

Adam P. Morris,1,* Michael Kubischik,2 Klaus-Peter Hoffmann,3 Bart Krekelberg,1,4 and Frank Bremmer3,4
1Center for Molecular and Behavioral Neuroscience, Rutgers University, Newark, NJ 07102, USA
2Allgemeine Zoologie und Neurobiologie, Ruhr-Universität, Bochum 44780, Germany
3Department of Neurophysics, Philips-Universität, Marburg 35032, Germany
4These authors contributed equally to this work

Summary

Background: Many visual areas of the primate brain contain signals related to the current position of the eyes in the orbit. These cortical eye-position signals are thought to underlie the transformation of retinal input—which changes with every eye movement—into a stable representation of visual space. For this coding scheme to work, such signals would need to be updated fast enough to keep up with the eye during normal exploratory behavior. We examined the dynamics of cortical eye-position signals in four dorsal visual areas of the macaque brain: the lateral and ventral intraparietal areas (LIP; VIP), the middle temporal area (MT), and the medial-superior temporal area (MST). We recorded extracellular activity of single neurons while the animal performed sequences of fixations and saccades in darkness.

Results: The data show that eye-position signals are updated predictively, such that the representation shifts in the direction of a saccade prior to (<100 ms) the actual eye movement. Despite this early start, eye-position signals remain inaccurate until shortly after (10–150 ms) the eye movement. By using simulated behavioral experiments, we show that this brief misrepresentation of eye position provides a neural explanation for the psychophysical phenomenon of perisaccadic mislocalization, in which observers misperceive the positions of visual targets flashed around the time of saccadic eye movements. Such experiments have shown that targets flashed just before saccades are mislocalized in the direction of eye movement, whereas targets flashed just after saccades are mislocalized in the opposite direction (see [14] for a review). This pattern of perceptual error has been attributed to a damped (i.e., temporally smoothed) internal representation of eye position [10, 12, 13, 15]. If cortical eye-position signals contribute to visuospatial perception, as suggested by theoretical work, one would predict a match between their dynamics and the eye-position signal inferred from psychophysical data.

To examine the dynamics of cortical eye-position signals, we recorded extracellular spiking activity of single neurons in two macaque monkeys (four hemispheres) as they performed a combination of fixations and saccadic eye movements in near darkness. Recordings were performed in four visual cortical areas: the lateral intraparietal area (LIP), the ventral intraparietal area (VIP), the middle temporal area (MT), and the medial superior temporal area (MST). These areas contain neurons that exhibit systematic changes in firing rate across eye positions, even in darkness [2, 4, 5, 16]. Our experiments show that the updating of cortical eye-position signals begins before the onset of eye movement but remains incomplete until shortly after saccade offset. We show that these dynamics provide a striking match to those predicted from psychophysical studies of perisaccadic mislocalization. This correspondence between neural data and a well-established perceptual illusion provides, to our knowledge, the first physiological evidence for a direct role of cortical eye-position signals in perception.

Conclusions: Together, these results suggest that eye-position signals in the dorsal visual system are updated rapidly across eye movements and play a direct role in perceptual localization, even when they are erroneous.

Introduction

To localize a visual object, an observer must take into account not only its position on the retina, but also the positions of the eyes in the orbit, the angle of the head relative to the body, and many other postural variables. A candidate mechanism for this type of integration is embodied in cortical neurons that modulate their response to stimuli as a function of gaze direction [1–5]. Such neurons represent simultaneously the contents of a visual scene and the current positions of the eyes in the orbit, and thus contain the crucial ingredients for the construction of a stable spatial code in the context of ongoing eye movements. Indeed, since their discovery in parietal cortex, a wealth of theoretical work has linked these eye-position signals to many fundamental spatial abilities of the primate nervous system, including navigation, multisensory integration, sensorimotor transformations, and perceptual localization [6–8].

Despite the maturity of the theoretical framework, it remains controversial whether eye-position signals in visual areas of the brain are actually used for online spatial coding (see [9] for a recent discussion). Further, many characteristics of these signals that are fundamental to the theory have not been investigated experimentally. In particular, it is unknown whether they are updated fast enough to keep up with the eyes during exploratory behavior—an obvious prerequisite for their purported roles.

In contrast, the dynamics of eye-position representation have been investigated extensively in the human psychophysical literature [10–13]. There, inferences about eye-position signals have been made on the basis of perceptual performance around the time of saccadic eye movements. Such experiments have shown that targets flashed just before saccades are mislocalized in the direction of eye movement, whereas targets flashed just after saccades are mislocalized in the opposite direction (see [14] for a review). This pattern of perceptual error has been attributed to a damped (i.e., temporally smoothed) internal representation of eye position [10, 12, 13, 15]. If cortical eye-position signals contribute to visuospatial perception, as suggested by theoretical work, one would predict a match between their dynamics and the eye-position signal inferred from psychophysical data.

*These authors contributed equally to this work
Correspondence: adam@vision.rutgers.edu
spiking activity while the eyes were stationary at 13 unique positions (“fixation” epochs), as well as during eye movements (“perisaccadic” epochs). We report our findings in two sections. First, we present physiological data that reveal the dynamics of cortical eye-position signals. Second, we use a simulation approach to determine the patterns of perisaccadic localization that would be expected if these signals were used to localize visual objects.

Modulation of Firing Rates by Eye Position during Fixation

Figure 1B shows mean firing rates during fixation at each of the unique eye positions for a representative LIP neuron. The rate varied systematically with the position of the eye. We refer to this relationship between firing rate and eye position as an eye-position field. To quantify these effects, we fitted the firing rates observed during fixation for each neuron with a two-dimensional polynomial (via stepwise regression). The surface in Figure 1B shows the fitted function for the example neuron. On average, eye-position fields consisted of a near-doubling of firing rate across the oculomotor region we examined. The modulations associated with the change in eye position for a given saccade condition were smaller (approximately 20% on average; Figure S1A available online). These results were comparable across all cortical regions we examined. Significant regression surfaces were observed in approximately half (VIP, MT/MST) to three-quarters (LIP) of the recorded neurons (see Figure S1B for results and statistical analyses). These results are consistent with those reported previously [4, 5, 16].

Dynamics of Cortical Eye-Position Signals

Our aim was to determine the time course of the transition from the firing rate during fixation before a saccade to the new firing rate after the saccade. Individual neurons, however, do not provide information about a change in eye position for all saccade directions. For instance, the neuron in Figure 1B would not change its response after a vertical saccade and should not be included when studying the population dynamics around vertical saccades. For this reason, we included a neuron in the analysis of a particular saccade condition only if there was a statistically significant step in activity from the pre- to
Saccade dynamics are also plotted (dotted line, scaled to match shown in Figure 2B. For comparison, the monkey's actual ence in activity between the neurons in the two groups, as way the brain might extract such a signal is to take the differ-

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activity before the onset of the eye movement and stabilized shortly after (<50 ms) the offset of the saccade. The POS group, in contrast, showed a transient increase in activity prior to (~150 ms) the eye movement, followed by stabilization after (~150 ms) the eye landed. These changes in firing rate occurred earlier than could be accounted for by the temporal averaging associated with the calculation of firing rates (up to 25 ms of shift). These dynamics were comparable across the ten saccade conditions, as shown implicitly by the small standard errors in Figure 2A (shaded regions).

The dynamics observed for POS and NEG groups suggest that they provide incongruent information about eye position during saccades. For object localization, however, the brain must compute a singular estimate of gaze direction. One simple way the brain might extract such a signal is to take the difference in activity between the neurons in the two groups, as shown in Figure 2B. For comparison, the monkey's actual saccade dynamics are also plotted (dotted line, scaled to match the overall change in firing rate). The derived eye-position signal resembled a damped (i.e., low-pass filtered) version of the saccade.

Figure 3 shows the same analyses when neurons from each cortical region were considered separately. Although there was variation in the specific patterns for POS and NEG groups across areas, such as drifts during fixation for VIP and MT/MST, the time courses shared many common characteristics (Figure 3A). As can be seen in the derived eye-position signals (Figure 3B), the primary difference across areas was the strength of modulation rather than the dynamics. To confirm the predictive nature of these signals, as well as to compare their magnitudes across areas, we compared their values during the initial fixation epoch with those just before the onset of the saccade (t = ~25 ms). The data were analyzed with a mixed-design ANOVA with factors of time (2 levels) and area (3 levels), and the ten independent observations across task conditions as the dependent variable. There was a significant main effect of time ($F(1,27) = 5.97, p < .05$), confirming the predictive effect, and no significant interaction ($F(2,27) = 0.122, p = .89$), suggesting that predictiveness did not vary across areas.

To this point, we have considered only the dynamics of the average population activity. To determine whether individual neurons also carried predictive information about future eye positions, we compared two quantities for each neuron in each condition. The first was the overall change in mean firing rate across successive fixations. By definition, positive and negative values on this measure correspond to POS and NEG neurons, respectively. The second was the difference in activity between the baseline rate during the initial fixation epoch and that observed just before the saccade (~75 ms to ~25 ms). A positive correlation between these two measures across the population would indicate that neurons anticipate the impending change in eye position. For NEG neurons, this was indeed the case for all of the cortical regions we examined (Figure S3). That is, neurons that showed larger decreases in activity across successive fixations also tended to show larger drops in activity just prior to the saccade. POS neurons, in contrast, showed no such effect in any area, suggesting that their predictive behavior is limited to a presaccadic bump in the population response.

**Optimized Saccade Directions**

In Experiment 2, we first performed a preliminary assessment of the eye-position field to assess the axis along which the neuron modulated its firing rate maximally, and then performed an experiment with large (40° amplitude) saccades back and forth along this axis (Figure 4A). This tailored design...
NEG directions, pooled across cortical areas. The data provide a striking replication of the key findings of the first experiment (cf. Figure 2A). In particular, it is clear that the transition from a low to high firing rate across the saccade was delayed relative to that of a high-to-low transition. Moreover, the derived eye position signal (POS – NEG) again resembled a damped version of the actual saccade. Figure S5 shows the time courses for example neurons from areas LIP, VIP, and MST.

For completeness, we also report the population dynamics separately for each cortical region (Figure 5), though these results should be interpreted with caution because of relatively small sample sizes. The results largely replicated the findings of Experiment 1. One exception was the dynamics for NEG saccades in area LIP, which did not show the otherwise ubiquitous decline in activity just before and during the saccade. Instead, a presaccadic increase in firing rate was observed, which resulted in a delay of the derived eye-position signal. This suggests that saccade-related responses—known to be prevalent in LIP—can mask the predictive eye-position signal when they predominate in a given sample of neurons.

Perceptual Localization and the Psychophysical “Eye-Position Signal”

We next considered the pattern of perceptual localization that would be expected if the cortical eye-position signals reported here were used for online spatial coding. We simulated an experiment in which a visual target was flashed at a fixed position in space at different times relative to the onset of the saccade (Figure S6). The perceived position of the target was computed as the sum of the position of the target on the retina (which changes depending on the actual eye position at the time of the flash) and the position of the eye encoded by the neural eye-position signal (from Figure 2B). A latency parameter, \( \tau \), was included to take into account the delay between the actual time of the stimulus event and the time at which the relevant visual signals are combined with eye-position information. The specific pattern of mislocalization depends on the assumed value of \( \tau \) (a free parameter).

The results of these simulations show that the neural eye-position signal can account for both the spatial and temporal aspects of perisaccadic mislocalization in humans (Figure 6) [10, 12, 13]. Specifically, simulated visual targets were localized accurately when flashed well before or well after the saccades, but mislocalized in the direction of the eye movement when flashed just before saccade onset. This effect peaked at saccade onset—at which time the error was approximately half the saccade amplitude—and declined rapidly thereafter. After the offset of the saccade, visual targets were mislocalized in the direction opposite to that of the saccade.

Discussion

Our experiments reveal that the dorsal visual processing stream of the macaque brain is furnished with a surprisingly nimble representation of eye position. We found that eye-position signals were updated predictively, such that just before the onset of a saccade, neurons behaved as if the eyes had already begun their journey toward the new fixation position. The updating occurred more slowly than the saccade itself, but was completed well within the duration of a typical fixation (∼300 ms) [17]. Thus, although imperfect, eye-position signals in cortex are updated sufficiently fast to keep steady with the eyes during normal exploratory behavior. By using simulations, we showed that the imperfections of this neural signal predict a pattern of spatial (mis)localization that matches increased the magnitude of eye-position-dependent modulations across successive fixations, and also ensured that we recorded the same neuron during both POS- and NEG-direction saccades over the same region of oculomotor space.

Most neurons (62 out of 78) showed significant and large changes in mean firing rate across the two fixation positions (paired t tests), as expected for this tailored design. Figure 4B shows the population dynamics for saccades in the POS and NEG directions, pooled across cortical areas. The data provide...
that observed in humans [10, 12, 13]. To our knowledge, these findings provide the first physiological evidence that cortical eye-position signals contribute to the localization of visual objects.

Before discussing the implications of these findings, we first consider whether our results could be explained by incidental visual stimulation from the visual target or ambient illumination. Such effects are unlikely to account for our findings. First, we minimized visibility by performing experiments in near darkness and by preventing dark adaptation. Second, visual responses cannot explain the presaccadic reduction of activity for saccades in the NEG direction. Finally, presaccadic increases in activity for saccades in the POS direction peaked just before the saccade (<75 ms; see Figure 2A), later than the expected onset for visually evoked responses (50–100 ms after target onset; mean saccadic latency was 215 ms) [18]. Therefore, the dynamics we observed are probably related to intrinsic signals around the time of an eye movement.

The data suggest a role for eye-position signals in the computations that underlie visual stability across saccades. However, the nature of this involvement remains unclear. In particular, it is unknown whether they participate directly in an implicit representation of visual space [7] or indirectly through their involvement in the construction of explicit head- or world-centered receptive fields downstream [8]. Head-centered receptive fields have been reported in multiple visual areas, including VIP [19] and MT [20–22], but see [23–26], but are generally considered to be rare in cortex. Consistent with this observation, our data show that explicit representations are not necessary to explain visual stability and its imperfections, and thus lend support to theories that incorporate distributed spatial codes.

One caveat is that these conclusions rely on a comparison of physiological data with previously published psychophysical data and not with simultaneous behavioral measurements from our animals. A second caveat is that localization is more complex than the combination of a perfect representation of retinal position with a possibly erroneous eye-position signal. This is supported by data showing a dependence of perisaccadic mislocalization on such features as visual references [27] and stimulus contrast [28], as well as physiological data showing remapping or distortions of retinal codes during eye movements [29, 30]. Additionally, under certain experimental conditions, measurements of perisaccadic localization also reveal a compression component in which targets are misperceived toward the saccade target [31]. This latter effect could reflect differences in eye-position signal dynamics for neurons whose receptive fields cover different parts of the retina [32, 33].

We next consider the kinds of signals that could give rise to the dynamics we observed. The predictive behavior is inconsistent with proprioceptive input from the stretch receptors of the extraocular muscles, because such signals necessarily lag behind the eye [9]. Moreover, the timing does not correspond well to the output of the neural integrator of the brain stem oculomotor plant [34], which pre-empts eye movement by less than 20 ms [35]. In contrast, the dynamics provide a good match to those reported in studies of eye-position signals in the central thalamus [36, 37]. There, eye-position signals are updated as early as 100 ms before saccade onset and as late as 200 ms after an eye movement, consistent with our results. Moreover, such neurons exhibit a comparable hysteresis for saccades in the POS and NEG directions (for specific examples, compare our data with Figures 5b and 8 of [37]).

Although these subcortical effects bolster our findings, they nevertheless leave unanswered the question of how such eye-position signals arise. We speculate that they are computed in cortex from corollary signals that predict the onset and metrics of an impending saccadic eye movement. There are many such signals, including those related to saccade planning [38, 39], shifts of attention [40], spatial remapping [29, 41], and explicit corollary discharge [42]. LIP, for example, carries both a representation of current eye position (via eye-position fields [2]) and a representation of the impending saccade vector (via eye-centered “motor” fields [39]), and thus has all the ingredients needed to compute future eye positions.

An interesting possibility is that eye-position signals are linked to mechanisms of saccadic suppression. Saccadic suppression is characterized by a reduction of visually evoked and spontaneous activity in cortical neurons around the time of a saccade [43]. For a neuron whose eye-position field...
imposes a lower firing rate after the saccade (i.e., a NEG neuron), this suppression would also act as a predictive eye-position signal. In contrast, suppression is antagonistic for a POS neuron, and would cause the eye-position signal to be lower after the saccade (i.e., a NEG position signal). In contrast, suppression is antagonistic for a POS neuron, this suppression would also act as a predictive eye-position signal. In contrast, suppression is antagonistic for a POS neuron, and would cause the eye-position signal to be lower after the saccade (i.e., a NEG position signal).

Figure 6. The Dynamics of Cortical Eye-Position Signals Predict Mislocalization of Visual Targets

The plots compare patterns of mislocalization predicted from our neural data (black curves) with those observed experimentally in classical studies of human perception by Dassonville, Schlag, and Schlag-Rey [13] and Honda [10, 12] (red curves; shading = standard deviation across subjects; replotted with permission). The abscissa represents the time of a target flash relative to saccade onset. The ordinate represents the predicted or actual perceptual error, expressed as a percentage of saccade amplitude. Predicted errors were calculated as the mismatch between the actual eye position and that represented by neurons in dorsal visual cortex (i.e., Figure 2B), taking into account visual latency (Figure S6). The best-fitting latency values for the Dassonville et al. [13] and Honda [10, 12] data sets were 66 ms and -6 ms, respectively. The first of these values matches the latency of visual responses in these cortical areas [18], but the second (negative) latency is unexpected (assuming that the brain does not compensate for visual latency during perceptual localization). However, good accounts of the psychophysical data were produced across a wide range of latencies (see Figure S6 for further examples). The inclusion of visual latency as a free parameter was inspired by theoretical work that has linked temporal windows for the POS/NEG categorization. These included all factorial combinations of five different window positions before (-700 to -300 ms, 100 ms increments, 100 ms window width) and after (+300 to +700 ms, 100 ms increments, 100 ms window width) the saccade. For each configuration, a population curve was computed by taking the median firing rate across neurons at each point in time. Finally, to combine the data across window configurations and remove potential selection effects, we erased from each curve the epochs used for the categorization (taking

Experimental Procedures

Electrophysiology

All animal procedures have been described in detail previously [18]. In brief, two male macaque monkeys (“M1” and “M2”) were implanted with recording chambers (M1: left hemisphere; LIP and VIP; right hemisphere; MT and MST; M2: opposite configuration) and scleral search coils. The animal sat in a primate chair with the head restrained and received liquid reward for each completed trial. All procedures were in accordance with published guidelines on the use of animals in research (European Council Directive 86/609/EEC and the National Institutes of Health Guide for the Care and use of Laboratory Animals) and approved by local ethics committees.

Stimuli

Target stimuli were small light-emitting diodes (0.5 diameter, 0.4 cd cm⁻²) back-projected onto a translucent screen subtending 60° × 60°. There were no other sources of ambient illumination and room lighting was used between recordings to prevent dark adaptation.

Behavioral Task

Experiment 1

At the beginning of each trial, the monkey maintained gaze (within 1° of error) on a target at one of five positions (x, y) = [0°, 0°], [−10°, 10°], [−10°, −10°], [10°, −10°], [10°, 10°]. After 1,000 ms, the target stepped 10° either rightward or downward. The animal performed a saccade to the new position within 500 ms and maintained fixation for another 1,000 ms. All trial types were interleaved in a pseudorandom order. The mean number of trials per condition across neurons was 15 (SD = 6).

Experiment 2

Experiment 2 required larger saccades back and forth along an inclined region of the neuron’s eye-position field (M1 only; LIP, N = 24; VIP, N = 29; MT/MST, N = 25). The task was identical to that of Experiment 1 except that the target stimuli were located at diametrically opposite positions on an imaginary circle around the center of the display (i.e., one of 0°/180°, 45°/225°, 90°/270°, or 135°/315°; radius = 20°). To choose this saccade axis, the neuron’s eye-position field was quantified online by fitting a regression plane to data from preliminary trials in which the animal performed the task of Experiment 1. One of the two saccade directions was thus in a “POS” direction for the neuron and the other was in a “NEG” direction. These conditions alternated across trials. The mean number of trials for each saccade direction was 41 (SD = 13).

Data Analysis

Primary saccades were detected offline with eye-velocity-based criteria, and trials in which the monkey failed to perform the task correctly were discarded.

Dynamics of Eye-Position Signals

A firing rate time course was constructed for each neuron by counting spikes that occurred within a 50 ms window stepped in 25 ms increments across each trial. These data were then aligned to the onset of the saccade and averaged across trials to provide separate rate curves for each of the saccade conditions. Finally, to facilitate the combination of data across neurons, these curves were normalized by expressing the rates as a percentage of the mean firing rate across the two fixation intervals in each condition (Figure 1A).

Population responses were computed separately for neurons that on average increased (POS) and decreased (NEG) their activity across a given saccade condition. For Experiment 1, in which the saccade vectors in the task were held constant across recordings, the categorization of neurons into POS and NEG groups was achieved posthoc by identifying neurons that had significantly different firing rates during two fixation epochs for each condition (paired t tests). To ensure that the results were not unduly influenced by our operational definition of “fixation,” a family of population curves was computed for each condition, each incorporating different temporal windows for the POS/NEG categorization. These included all factorial combinations of five different window positions before (−700 to −300 ms, 100 ms increments, 100 ms window width) and after (+300 to +700 ms, 100 ms increments, 100 ms window width) the saccade. For each configuration, a population curve was computed by taking the median firing rate across neurons at each point in time. Finally, to combine the data across window configurations and remove potential selection effects, we erased from each curve the epochs used for the categorization (taking
into account temporal smoothing) and averaged over the remaining portions of the data sets.

For Experiment 2, in which the task parameters were optimized at the time of data collection, there was less chance of false categorizations and thus a single population time course was obtained for each of the two saccade directions (with the standard fixation epochs shown in Figure 1; identical results were found with the more robust method of Experiment 1).

Supplemental Information

Supplemental Information includes Supplemental Experimental Procedures and six figures and can be found with this article online at doi:10.1016/j.cub.2011.12.032.

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