CORTICAL MECHANISMS OF SACCADIC SUPPRESSION

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ABSTRACT

Saccades are a ubiquitous and crucial component of our visual system, allowing for the efficient deployment of the fovea and its accompanying neural resources. Initiation of a saccade is known to cause saccadic suppression, a temporary reduction in visual sensitivity[1,2] and visual cortical firing rates[3-6]. While saccadic suppression has been well characterized at the level of perception and single neurons, relatively little is known about the visual cortical networks governing this phenomenon. Here we examine the effects of saccadic suppression on distinct neural subpopulations within area V4. We find cortical layer- and cell type-specific differences in the magnitude and timing of peri-saccadic modulation. Neurons in the input layer show changes in firing rate and inter-neuronal correlations prior to saccade onset, suggesting that this layer receives information about impending saccades. Putative inhibitory interneurons in the input layer elevate their firing rate during saccades and may play a role in suppressing the activity of other cortical subpopulations. A computational model of this circuit recapitulates our empirical observations and

demonstrates that an input layer-targeting pathway could initiate saccadic suppression by enhancing local inhibitory activity. Together, our results provide a mechanistic understanding of how eye movement signaling interacts with cortical circuitry to enforce visual stability.

INTRODUCTION

As we observe the world around us, our eyes dart from point to point. Each of these shifts in gaze is a saccade – a ballistic movement of both eyes. Saccades significantly improve the efficiency of the primate visual system by allowing us to flexibly deploy our fovea, the dedicated high acuity zone at the center of the retina, towards objects of interest in our environment. However, saccades also pose a substantial challenge for ongoing visual processing, as each saccade produces abrupt and rapid motion across the retina. Therefore, saccades are also accompanied by a temporary reduction in visual sensitivity, termed saccadic suppression, that serves to blunt our perception of this motion[1,2]. Perhaps reflecting the diminished sensory processing during saccades, the firing rates of neurons in many regions of the visual cortex are also transiently suppressed[3-6]**.** Yet, despite extensive psychophysical characterization and single neuron electrophysiological analysis, relatively little is known about the circuit level mechanisms underlying saccadic suppression in the visual cortex. One prevailing hypothesis suggests that saccadic suppression is mediated by a corollary discharge signal originating in the brain regions responsible for initiating eye movements[7-9]. Given the synaptic and motor delays accompanying the execution of the saccadic motor command, a coincident corollary discharge to the visual cortex could trigger compensatory mechanisms prior to the start of the eye movement. Consistent with this idea, changes in neural activity before saccade onset have been observed in several visual cortical areas[5]. However, in spite of clear neural evidence that the visual cortex receives information about upcoming saccades,

47 the pathway by which a saccadic signal arrives in the visual cortex is unknown, and how that signal
48 interacts with local circuitry to mediate saccadic suppression remains unclear.

49 We examined these questions within area V4, a critical hub in the visual processing stream
50 responsible for object recognition[10]. V4 is known to exhibit saccadic suppression in both humans[4]
51 and macaques[6], and undergoes dynamic shifts in stimulus sensitivity during the preparation of
52 saccades[11]. To investigate the underlying neural mechanisms, we studied saccadic suppression in
53 the context of the laminar cortical circuit, which is composed of six layers with highly stereotyped
54 patterns of intra- and inter-laminar connectivity[12,13]. In higher-order visual areas, such as V4, layer
55 IV (the *input* layer) is the primary target of projections carrying visual information from lower-
56 order areas, such as V1, V2, and V3[14,15]. After arriving in the input layer, visual information is
57 then processed by local neural subpopulations as it is sent to layers II/III (the *superficial* layer) and
58 layers V/VI (the *deep* layer), which serve as output nodes in the laminar circuit[12,16]. In V4, the
59 superficial and deep layers feed information forward to upstream visual areas, such as ST[17], TEO[18],
60 and TE[19]. Alongside feedforward input from lower-order visual areas, V4 also receives projections
61 from other cortical regions, such as the prefrontal cortex[15], as well as from subcortical structures[20].
62 Accordingly, several pathways targeting V4 could be responsible for relaying information about
63 upcoming saccades: 1) projections from lower visual areas (V1/V2/V3), which predominantly
64 target the input layer[14,15], 2) a projection from the frontal eye fields (FEF), which predominantly
65 targets the superficial and deep layers[21,22], and 3) a projection from the pulvinar, which
66 predominantly targets the superficial and input layers[23-25]. Neurons in V1[26,27], FEF[28-30], and
67 pulvinar[31-33] are known to respond to saccades, and the corresponding pathways are therefore
68 strong candidates for initiating saccadic suppression in V4.

69 Given the organization of cortical circuitry, we reasoned that a saccadic signaling pathway
70 should target the site of entry for visual information, the input layer, to most efficiently suppress
71 visual processing. Furthermore, in light of the similarities between peri-saccadic stimulus
72 processing and general reductions in visual gain[6,26], we considered whether saccadic suppression
73 might be mediated in a manner comparable to visual gain modulation. Specifically, we
74 hypothesized that an elevation in local inhibitory activity, a common mechanism for gain control[34-
75 36], functionally suppresses neural processing in V4 during saccades. We sought to confirm our
76 predictions by utilizing a cortical layer-specific recording approach in combination with
77 electrophysiological cell type classification to simultaneously record from six well-defined neural
78 subpopulations in V4.
79
80 **RESULTS**
81 **Dissociating the Effects of Vision and Eye Movements**
82 Past approaches for studying saccadic modulation have typically involved cued saccade
83 tasks[4,6,26,37], which risk conflating the neural effects of visual stimulus processing with the neural
84 effects of saccades. To dissociate saccadic signaling from visually induced activity in area V4, we
85 performed a series of spontaneous recordings while two rhesus macaques were seated in the dark.
86 We found that despite their inability to observe specific objects in the environment, both animals
87 continued to make saccades freely under these conditions. We identified individual saccades by
88 estimating the onset and offset times of ballistic eye movements (see Methods). In order to remove
89 fixational eye movements (microsaccades), we discarded saccades with amplitudes less than 0.7
90 degrees of visual angle. Additionally, to prevent the effects of one eye movement from overlapping
91 with those of the next, we only included saccades that were separated from one another by at least

92 500 ms. This approach provided us with a set of 4420 saccades that were used for all subsequent
93 analyses. We evaluated the quality of our eye movement dataset by examining amplitude and
94 velocity traces from individual saccades, as well as by characterizing the main sequence. The main
95 sequence is the approximately linear amplitude-velocity relationship observed across all
96 saccades[38], which we found held true for our data (Figure 1A).

97 While the monkeys were making spontaneous saccades, we simultaneously recorded
98 neural activity from well-isolated single-units (n = 211), multi-units (n = 110), and local field
99 potentials (LFPs) using linear array probes in area V4. The use of recording chambers containing
100 an optically-transparent artificial dura allowed us to make electrode penetrations perpendicular to
101 the cortical surface, and therefore in good alignment with individual cortical columns (see
102 Methods). The alignment of each penetration was confirmed by mapping receptive fields along
103 the depth of cortex, with overlapping receptive fields indicating proper alignment (Figure 1B).
104 Laminar boundaries were identified with current source density (CSD) analysis[39,40]. The CSD,
105 defined as the second spatial derivative of the LFP, maps the location of current sources and sinks
106 along the depth of the probe. Each layer of the visual cortex produces a characteristic current
107 source-sink pattern in response to visual stimulation, with the input layer producing a current sink
108 followed by a current source, and the superficial and deep layers producing the opposite pattern of
109 activity (Figure 1C). By identifying transitions between current sources and sinks, we assigned
110 laminar identities to each of our recording sites (Figure 1D – LFP traces colored by layer). We
111 also separated well-isolated single-units into two populations on the basis of their waveform
112 duration: narrow-spiking units, corresponding to putative inhibitory interneurons, and broad-
113 spiking units, corresponding to putative excitatory neurons (Figure 1D – spikes colored by unit

type)[41-43]. The three cortical layers and two unit types thus gave us six distinct neural subpopulations to consider in subsequent analyses.

Narrow-Spiking Input Layer Units Show Positive Peri-Saccadic Modulation

To evaluate neural subpopulation-specific responses, we first estimated the peri-saccadic firing rate of well-isolated single-units with a kernel-based approach, in which each spike is convolved with a gaussian kernel[44] (Figure S1). We selected the kernel bandwidth on a unit-by-unit basis at each point in time using an optimization algorithm that minimizes mean integrated squared error[45]. When looking at units grouped by cell type, narrow-spiking units exhibit less peri-saccadic suppression than broad-spiking units (Figure 2A). The cause of this difference became clear when the units were further separated by layer. All laminar subpopulations of broad-spiking units displayed peri-saccadic suppression, as did narrow-spiking units in the superficial and deep layers. In contrast, narrow-spiking input layer units elevated their firing rate in response to a saccade (Figure 2B-C; see Figure S2 for additional single-unit examples).

To quantify these observations, we calculated a modulation index that represents the degree to which the firing rate of each unit deviated from its pre-saccadic baseline activity (see Methods). We found that narrow-spiking input layer units were the only group with a positive modulation index, which was significantly different from most other subpopulations (Figure 2D). We also noted that at the single-unit level, each of the six subpopulations had some proportion of positively modulated units (Figure S3). Therefore, the difference in subpopulation level firing rates could be caused by narrow-spiking input layer neurons having a greater average magnitude of positive modulation, or by having a greater proportion of positively modulated units. We determined both to be true. Narrow-spiking input layer neurons displayed a larger magnitude of positive modulation

137 (Figure 2E), and were also more likely to be positively modulated (Figure 2F). We found it
138 particularly striking that, in accordance with our hypothesis, only narrow-spiking input layer units
139 (i.e. putative inhibitory interneurons) showed an enhancement in firing. This led us to believe that
140 an elevation in input layer inhibitory activity could be responsible for the peri-saccadic suppression
141 of firing within the visual cortex reported previously[3-6].
142
143 **Input Layer Units Respond Prior to Saccades**
144 In order to gate out visual signals received during a saccade, neurons in the cortical layer initiating
145 suppression would need to be activated prior to the start of the eye movement. To confirm that this
146 was true for the input layer, we performed a timing analysis to determine whether any of the
147 recorded subpopulations show significant changes in firing rate prior to saccade onset. For each
148 single-unit, we found the time at which their activity first deviated outside a 95% confidence
149 interval calculated from their pre-saccadic baseline activity, an event that was marked as the time
150 of first significant response. The results of this approach for six example units (the same example
151 units as in Figure S1) are illustrated in Figure 3A. While there was response variability at the level
152 of single-units, the input layer subpopulations (both narrow- and broad-spiking) were the only ones
153 that responded prior to saccade onset (Figure 3B). This provides strong evidence that the input
154 layer receives information about upcoming saccades with sufficient time to enact changes in
155 activity throughout the cortical column.
156
157 **Saccades Increase Correlated Variability in the Input Layer**
158 If input layer neurons receive pre-saccadic excitation from a common source, that signaling should
159 also manifest itself as an increase in correlated variability among pairs of input layer units prior to

saccade onset. To test this, we computed spike count correlations (SCC) between pairs of simultaneously recorded single- and multi-units as a function of time, and pooled these results into six groups on the basis of their pairwise laminar locations. All pair types showed a substantial increase in SCC around the time of saccades (Figure S4A). We quantified the precise time at which correlations first began to rise with a bootstrapping analysis (see Methods). SCC began increasing prior to saccade onset in input-input pairs (Figure 4A), but after saccade onset in all other pairs (Figure 4B). This pattern of activity is consistent with the pre-saccadic arrival of a signal that is shared among input layer units. The delayed response of all other pair types suggests that they inherit their activity from the input layer. Additionally, the precise time at which input-input correlations began to rise (~40 ms before saccade onset) is remarkably similar to the time at which input layer units began to show firing rate modulation (also ~40 ms before saccade onset), further indicating that these two observations have a common cause: the activation of a saccadic signaling pathway.

Alongside the expected rise in correlations, we also found a reduction in correlations that began ~250 ms prior to saccade onset among input-input pairs. Because the timing of this drop precedes saccadic initiation, which begins ~ 150 ms before saccade onset[46,47], it could not be the result of a corollary discharge signal. Instead, it may reflect the shift in internal state that occurs before saccadic initiation as the subject shifts their attention towards a potential target location[48]. In accordance with this possibility, pupil diameter, a well-established correlate of internal state[49,50], begins to ramp several hundred milliseconds prior to saccade onset (Figure S4B)[51]. Similar shifts in internal state are known to be accompanied by comparable reductions in correlated variability[50,52-54].

To further characterize changes in coordinated peri-saccadic activity, we also examined differences in spike-spike coherence (SSC) before and after saccade onset. Spike-spike coherence is a measure of the degree to which the activity of two signals, in this case a pair of simultaneously recorded single- and/or multi-units, fluctuates together across a range of frequencies. We found that saccade onset increases wide-band coherence most prominently in the input layer, with weaker, but still significant, effects in the superficial and deep layers (Figure 4C). When quantified as a modulation index (Figure 4D), we observed significant modulation in all layers at all frequencies, with the exception of the 30-100 Hz band in the superficial layer. The input layer exhibits larger increases in coherence, particularly at higher frequencies, while the deep layer experiences smaller increases, particularly at lower frequencies. Collectively, these data suggest that saccade onset leads to a temporary elevation of correlated activity, which manifests itself most prominently in the input layer. The timing and magnitude of changes in correlated variability, measured with both SCC and SSC, further suggests that an external signal arrives prior to saccade onset within the input layer before then propagating to the other layers.

Saccades Increase the Strength of Low-Frequency Wide-Band Activity

Increases in coordinated activity at the level of single neurons can result in an elevation of population-level wide-band activity. To determine whether this held true for saccades, we calculated the spectral power, a frequency-resolved measure of signal strength, of the LFP around the time of saccade onset. Saccade onset was associated with a large elevation in low frequency power, and a slight reduction in high frequency power (Figure 5A). Comparing the power spectrum before and after saccade onset, we found that these differences were significant in all three frequency bands tested, with an elevation in the 0-12 Hz and 15-25 Hz bands and a reduction in

205 the 30-80 Hz band after saccade onset (Figure 5B). Next, we sought to investigate whether the
206 increased strength of low-frequency wide-band activity entrains spiking units, as reported
207 previously in other contexts[55-57]. We found that saccade onset significantly increases low-
208 frequency spike-LFP phase locking (Figure 5C), as measured by pairwise phase consistency
209 (PPC). These results indicate that saccade onset increases the strength of population-level low-
210 frequency wide-band activity, which could in turn entrain the activity of local neurons.
211

212 **A Computational Model Demonstrates that Activation of an Input Layer-Targeting**
213 **Projection is Sufficient for Inducing Saccadic Suppression**

214 Given that input layer units exhibit significant changes in firing properties prior to saccade onset,
215 we hypothesized that a projection targeting this layer could serve as the initiator of saccadic
216 suppression within area V4. To determine whether such a projection would be sufficient for
217 initiating suppression across all cortical layers, as well as to gain further insight into the associated
218 circuit-level mechanisms, we developed a computational model of peri-saccadic activity changes
219 within a cortical column. Our model draws inspiration from previous cortical circuit models that
220 describe population-specific firing rate changes in response to the activation of an external input[58].
221 To more faithfully simulate cortical circuit-level dynamics, our model consists of six
222 interconnected populations across three layers of the cortex (Figure 6A, left). Each cortical layer
223 contains an excitatory (E) and inhibitory (I) population, which project to each other as well as to
224 themselves. Given evidence of multiple operating regimes for a local E-I network in the visual
225 cortex, connections within a layer were tuned to allow the local E-I network to switch in and out
226 of an inhibition-stabilized network (ISN) regime[59,60]: the baseline regime of the network is non-
227 ISN, but direct external input to the E population switches it to ISN (see Methods; Figure S5A-C).

228 Connections between layers exist in the form of projections from E neurons in the source layer to
229 both E and I neurons in the target layer, with E-E and E-I connections tuned to affect a net increase
230 in E activity in the target layer in response to an increase in E activity in the source layer. In our
231 model, the input layer projects to the superficial layer, which then projects to the deep layer; this
232 connectivity motif highlights the primary pathway for information flow within a visual cortical
233 column[13,61]. In light of our empirical firing rate analysis, which found that both broad- and narrow-
234 spiking units in the input layer show responses prior to saccade onset (Figure 3B), we chose to
235 model the input layer as the primary recipient of the 'saccadic signal' input. We tuned the strength
236 of the excitatory 'saccadic signal' pathway to the local E and I populations in this layer such that
237 its activation affected a net suppression of the E activity.

238 In response to the activation of a saccadic signal of sufficient strength slightly before the
239 time of saccade onset, the E input layer population rate was suppressed. In addition, our model
240 simultaneously recapitulated our experimental findings in the input layer I population
241 (enhancement), as well as the E/I populations in other layers (suppression) by virtue of the inter-
242 and intra-layer connectivity (Figure 6B, S5B). Since local E activity is a major source of excitation
243 to the I population within a layer, we next explored the role of the saccadic signal in sustaining
244 input layer I activity despite a concurrent decline in local E activity. We explored three scenarios
245 with saccadic signals of varying magnitudes ramping up and down over a fixed period (Figure 6C,
246 top). We found that the increase in I activity in the input layer indeed depended on the strength of
247 the saccadic signal and was sustained at sufficiently high magnitudes. Examination of the
248 contribution of different excitatory pathways to the I population showed that sufficiently high
249 saccadic signal magnitude led to an increase in net excitation onto the I population that exceeded
250 the loss in excitation resulting from the reduction of local E firing rates (Figure 6C, middle and

251 bottom). Our model thus demonstrates that an input layer-specific saccadic signal of adequate
252 strength can increase local I activity, which is sufficient for inducing suppression across the depth
253 of the cortex. These findings correspond closely to our empirical observations. The model further
254 predicted that the time course of the increase in input layer I activity was dependent on the baseline
255 operating regime of the E-I network. When the network was robustly within the non-ISN regime,
256 I activity increased immediately following saccadic signal onset. However, when the network was
257 close to the switching point between the two regimes, the I activity showed little change below a
258 saccadic signal of sufficient magnitude, even when the local E activity showed significant
259 suppression (Figure S5E). This latter scenario corresponds closely to our empirical observations
260 in the input layer (Figure 2B-C), suggesting that the baseline state of the cortical network lies close
261 to the boundary between regimes.

262 Our experimental paradigm examined peri-saccadic cortical dynamics in the absence of
263 visual input. While this allowed us to investigate the cortical dynamics that are purely a result of
264 the saccadic signal and avoid potential stimulus-induced artifacts in our recordings, it leaves open
265 the question of how activity patterns might change in the presence of visual stimulation. To explore
266 this further, we added a second external excitatory input ('visual signal') to the input layer E
267 population (Figure 6A, right), which is the major target of feedforward visual input into V4[14]. The
268 addition of this second input shifted the local E-I network into an ISN regime[59], in which E and I
269 activity increase or decrease concurrently. As a result, simultaneous activation of the visual signal
270 led to a net reduction in the firing rate of both E and I input layer populations (Figure 6D, S5C).
271 This produced a similar reduction in excitatory drive to the superficial and deep layers, where
272 firing was again suppressed. These results are consistent with prior reports of suppression from
273 uncategorized neurons in the visual cortex in the presence of visual stimulation[3,5,6].

DISCUSSION

Our results provide evidence that the input layer of area V4 receives the earliest information about upcoming saccades. Units in the input layer showed changes in firing rate prior to saccade onset, while units in other layers did not. In addition, elevated spike-count correlations, a common byproduct of activating a shared input to a neural population[62-65], were observed earliest among pairs of input layer neurons. Consistent with a single input source (i.e. projection) producing both phenomena, changes in the firing rate and correlational structure of input layer units were initiated with similar timing, approximately 40 ms before saccade onset. These convergent results indicate that the saccadic signaling pathway in question predominantly targets the input layer, which thereby allows us to infer its source. Given the lack of pre-saccadic modulation in the superficial and deep layers, it is unlikely that a projection from the FEF is responsible for the observed suppression. Likewise, it is doubtful that suppression is fed forward by projections from lower visual areas (V1/V2/V3), as such a mechanism would not explain the enhancement of narrow-spiking input layer firing rates in V4. Furthermore, to our knowledge, there have been no clear reports of saccadic suppression in V2 or V3, and while V1 is known to be suppressed during saccades, it does not exhibit pre-saccadic modulation[26], and could therefore not be responsible for the changes in activity described here. Thus, we propose that saccadic suppression in V4 is likely initiated via thalamo-cortical projections from the pulvinar.

Several lines of research are consistent with this view. Peri-saccadic modulation has been observed in the pulvinar of macaques, where neurons respond to saccades made in the dark without visual stimulation[32,33]. Neurons in the pulvinar of cats are able to distinguish between internally generated saccades and simulated saccades produced with image motion, demonstrating that they encode the underlying motor command[31]. Furthermore, the pulvinar is known to play a crucial role

297 in spatial attention[66-68], which is necessary for the proper execution of saccades towards target
298 stimuli; indeed, temporary inactivation of the pulvinar results in significant deficits in saccadic
299 target selection and execution[69,70]. These findings indicate that neurons in the pulvinar would be
300 able to provide a meaningful saccadic signal to their efferent targets.

301 We also found that narrow-spiking units in the input layer showed a peri-saccadic
302 enhancement in firing. Narrow-spiking units are thought to correspond to parvalbumin-expressing
303 inhibitory interneurons[71,72], and are known to receive strong thalamocortical excitation.[73-75] This
304 led us to speculate that an elevated level of inhibition within the input layer could suppress local
305 excitatory neurons, as well as neurons in other cortical layers (Figure 6A). To explore this
306 hypothesis, we developed a six-population firing rate model of the visual cortex and its modulation
307 during saccades. The model demonstrated that pre-saccadic activation of an input layer-targeting
308 projection is sufficient for suppressing signaling within a cortical column. A compelling aspect of
309 our model is its simplicity – with straightforward inter-population connectivity rules, we were able
310 to both replicate our empirical findings as well as reconcile the opposing patterns of activity
311 observed in the input layer subpopulations. The model also reproduces trends in our experimental
312 observations on the relative time course of broad and narrow spiking activity under a set of
313 conditions wherein the input layer operates on the cusp of ISN and non-ISN operating regimes.
314 Given these results, we propose that inhibitory input layer neurons initiate saccadic suppression
315 and gate the processing of visual information entering the local cortical network.

316 We also observed increases in correlated activity across all cortical layers during saccades,
317 in the form of elevated spike-count correlations, spike-spike coherence, low-frequency power, and
318 low frequency spike-phase locking. Comparable increases in neural correlation have been
319 associated with diminished stimulus sensitivity in multiple model organisms across a variety of

320 sensory modalities[50,76-79]. This may be a result of heightened correlations functionally limiting the
321 ability of a neural population to encode information, as suggested by previous experimental and
322 theoretical work[80-83]. Similarly, saccadic suppression at the level of neurons and circuits is also
323 accompanied by a behaviorally significant reduction in stimulus detection capabilities[1,2],
324 indicating that the increases in correlated variability reported here may reflect or contribute to that
325 deficit.

326 In summary, we conclude that an input layer-targeting projection, likely originating in the
327 pulvinar, initiates saccadic suppression within V4 by enhancing the firing of local inhibitory
328 interneurons. Future work could extend this line of subpopulation-specific, circuit-level analysis
329 to examine activity changes during saccades in the presence of visual stimulation. Our model
330 provides a prediction for such experiments, suggesting that the peri-saccadic activity of inhibitory
331 input layer neurons would also be suppressed when the cortex is in a state of active information
332 processing. The absence of any reports of subpopulation-specific differences in peri-saccadic
333 modulation during previous studies employing cued saccade tasks is consistent with this view.

334 **METHODS**

335 **Surgical Procedures**

336 Surgical procedures have been described in detail previously[84,85]. In brief, an MRI-
337 compatible, low-profile titanium chamber was placed over the pre-lunate gyrus on the basis of
338 preoperative MRI imaging in two rhesus macaques (right hemisphere in monkey A, left
339 hemisphere in monkey C). The native dura mater was then removed, and a silicone-based, optically
340 clear artificial dura (AD) was inserted, resulting in an optical window over dorsal V4. All
341 procedures were approved by the Institutional Animal Care and Use Committee and conformed to
342 NIH guidelines.

343 **Electrophysiology**

344 At the beginning of each recording session, a plastic insert with an opening for targeting
345 electrodes was lowered into the chamber and secured. This served to stabilize the recording site
346 against cardiac pulsations. Neurons were recorded from cortical columns in dorsal V4 using 16-
347 channel linear array electrodes (‘‘laminar probes’’; Plexon, Plexon V-probe). The laminar probes
348 were mounted on adjustable x-y stages attached to the recording chamber and positioned over the
349 center of the pre-lunate gyrus under visual guidance through a microscope (Zeiss). This ensured
350 that the probes were maximally perpendicular to the surface of the cortex and thus had the best
351 possible trajectory to make a normal penetration down a cortical column. Across recording
352 sessions, the probes were positioned over different sites along the center of the gyrus in the
353 parafoveal region of V4 with receptive field eccentricities between 2 and 7 degrees of visual angle.
354 Care was taken to target cortical sites with no surface micro-vasculature and, in fact, the surface
355 micro-vasculature was used as reference so that the same cortical site was not targeted across
356 recording sessions. The probes were advanced using a hydraulic microdrive (Narishige) to first

357 penetrate the AD and then through the cortex under microscopic visual guidance. Probes were
358 advanced until the point that the top-most electrode (toward the pial surface) registered LFP
359 signals. At this point, the probe was retracted by about 100–200 mm to ease the dimpling of the
360 cortex due to the penetration. This procedure greatly increased the stability of the recordings and
361 also increased the neuronal yield in the superficial electrodes.

362 The distance from the tip of the probes to the first electrode contact was either 300 mm or
363 700 mm. The inter-electrode distance was 150 mm, thus negating the possibility of recording the
364 same neural spikes in adjacent recording channels. Neuronal signals were recorded extra-
365 cellularly, filtered, and stored using the Multi-channel Acquisition Processor system (Plexon).
366 Neuronal signals were classified as either multi-unit clusters or isolated single-units using the
367 Plexon Offline Sorter program. Single-units were identified based on two criteria: (1) if they
368 formed an identifiable cluster, separate from noise and other units, when projected into the
369 principal components of waveforms recorded on that electrode and (2) if the inter-spike interval
370 (ISI) distribution had a well-defined refractory period. Single-units were classified as either
371 narrow-spiking (putative interneurons) or broad-spiking (putative pyramidal cells) based on
372 methods described in detail previously[41]. Specifically, only units with waveforms having a clearly
373 defined peak preceded by a trough were potential candidates. Units with trough-to-peak duration
374 less than 225 ms were classified as narrow-spiking units; units with trough- to-peak duration
375 greater than 225 ms were classified as broad-spiking units.

376 Data were collected over 17 sessions (8 sessions in monkey A and 9 in monkey C), yielding
377 a total of 211 single-units (99 narrow-spiking, 112 broad-spiking) and 110 multi-unit clusters.

378 **Recording**

379 Stimuli were presented on a computer monitor placed 57 cm from the eye. Eye position
380 was continuously monitored with an infrared eye tracking system (ISCAN ETL-200). Trials were
381 aborted if eye position deviated more than 1 degree of visual angle from fixation. Experimental
382 control was handled by NIMH Cortex software (http://www.cortex.salk.edu/).

383 ***Receptive Field Mapping***

384 At the beginning of each recording session, neuronal RFs were mapped using subspace
385 reverse correlation in which Gabor (eight orientations, 80% luminance contrast, spatial frequency
386 1.2 cycles/degree, Gaussian half-width 2 degrees) or ring (80% luminance contrast) stimuli
387 appeared at 60 Hz while monkeys fixated. Each stimulus appeared at a random location selected
388 from an 11 x 11 grid with 1 degree spacing in the appropriate visual quadrant. Spatial receptive
389 maps were obtained by applying reverse correlation to the evoked LFP signal at each recording
390 site. For each spatial location in the 11 x 11 grid, we calculated the time-averaged power in the
391 stimulus-evoked LFP (0–200 ms after each stimulus flash) at each recording site. The resulting
392 spatial map of LFP power was taken as the spatial RF at the recording site. For the purpose of
393 visualization, the spatial RF maps were smoothed using spline interpolation and displayed as
394 stacked contours plots of the smoothed maps (Figure 1B). All RFs were in the lower visual
395 quadrant (lower left in monkey A and lower right in monkey C) and with eccentricities between 2
396 and 7 degrees of visual angle.

397 ***Current Source Density Mapping***

398 In order to estimate the laminar identity of each recording channel, we used a CSD mapping
399 procedure[39]. Monkeys maintained fixation while 100% luminance contrast ring stimuli were
400 flashed (30 ms), centered at the estimated RF overlap region across all channels. The size of the
401 ring was scaled to about three-quarters of the estimated diameter of the RF. CSD was calculated

402 as the second spatial derivative of the flash-triggered LFPs. The resulting time-varying traces of
403 current across the cortical layers can be visualized as CSD maps (Figure 1C). Red regions depict
404 current sinks in the corresponding region of the cortical laminae; blue regions depict current
405 sources. The input layer (Layer 4) was identified as the first current sink followed by a reversal to
406 current source. The superficial (Layers 1–3) and deep (Layers 5 and 6) layers had opposite sink-
407 source patterns. LFPs and spikes from the corresponding recording channels were then assigned
408 to one of three layers: superficial, input, or deep.

409 ***Spontaneous Recordings***

410 In the main experiment, all monitors and lights were turned off in the recording room and
411 adjacent control room to ensure that the environment was as dark as experimentally feasible. Eye
412 tracker and electrophysiological data were recorded while the animals executed eye-movements
413 freely in the dark.

414 **Data Analysis**

415 ***Saccade Identification***

416 To identify saccade onset and offset times from the raw eye-tracker data, we used the
417 Cluster Fix algorithm[86]. Cluster Fix performs k-means clustering on four parameters (distance,
418 velocity, acceleration, and angular velocity) to find natural partitions in the eye-tracker data and
419 identify saccades. To ensure that our set of identified eye movements was 'clean,' we imposed
420 several additional criteria: 1) to avoid adjacent saccades affecting our analysis, we only considered
421 saccades separated from neighboring saccades by at least 500 ms, 2) to exclude microsaccades, we
422 only considered saccades that were greater in amplitude than 0.7 degrees of visual angle, 3) to
423 exclude slower eye movements that Cluster Fix had mislabeled as saccades, we only considered
424 eye movements less than 200 ms in duration, and 4) to exclude periods of data erroneously labeled

425 as saccades due to blinking or drowsiness, we identified those periods using pupil diameter
426 measurements and excluded nearby saccades.

427 Saccadic eye movements are known to have an approximately linear relationship between
428 amplitude and peak velocity that is referred to as the main sequence[38]. To ensure that our approach
429 identified a set of eye movements that quantitatively resembled the properties of saccades, we
430 calculated the amplitude and peak velocity for each of the identified eye movements and found an
431 approximately linear relationship (Figure 1A).

432 ***Firing Rate Estimation and Modulation Index***

433 To estimate the peri-saccadic firing rate, spikes were extracted for each saccade beginning
434 500 ms prior to saccade onset and ending 500 ms after saccade onset. To obtain a continuous
435 estimate of the firing rate, each spike was convolved with a gaussian kernel and the average firing
436 rate across saccades was calculated for each unit (Figure S1; Figure S2). The kernel bandwidth
437 was selected on a unit-by-unit basis at each point in time using a bandwidth optimization algorithm
438 that minimized the mean integrated squared error[45]. In order to obtain reliable firing rate estimates,
439 only units that fired at least 75 spikes within 500 ms of saccade onset (before or after) were
440 included.

441 A modulation index was calculated for each unit to estimate the degree to which their
442 activity deviated from baseline following saccade onset (Figure 2D, E; Figure S3). The modulation
443 index was defined as $FR_s/FR_b - FR_b$, where FR_s is the maximum firing rate in the 200 ms after
444 saccade onset and FR_b is the baseline firing rate of the unit, calculated by averaging the firing rate
445 for each unit in a time window 350 to 250 ms before saccade onset.

446 ***Spike Count Correlations***

To obtain a continuous estimate of spike count correlations, we calculated the Pearson correlation of spike counts across saccades at each time bin for every pair of simultaneously recorded single- and multi-units (Figure 4A, Figure S4A). The time window was 201 ms in width, and was shifted by 1 ms steps. Spike counts were z-scored to control for firing rate differences across time. Traces were smoothed with a 101 ms Gaussian weighted moving average. To perform the timing analysis, we took the local minimum in correlation as the time at which the rise begins. Bootstrapping was performed for each pair type to obtain confidence intervals on the estimate (Figure 4B).

Spike-Spike Coherence

We computed the coherence between simultaneously recorded, multi-unit pairs within a cortical layer using multi-taper methods[87] for two different windows: the 200 ms before and after saccade onset (Figure 4C). Spike trains were tapered with a single Slepian taper (TW = 3, K = 5). Magnitude of coherence estimates depend on the number of spikes used to create the estimates[88]. To control for differences in firing rate across the two windows, we adopted a rate-matching procedure[53]. In order to obtain a baseline for the coherence expected solely due to trends in firing time-locked to saccade onset, we also computed coherence in which saccade identities were randomly shuffled. Only units that elicited at least 50 spikes from each neuron cumulatively across all saccades in both time windows were considered. A modulation index, defined as $(SSC_{after} - SSC_{before})/(SSC_{after} + SSC_{before})$, was calculated for each pair of units at each frequency band (Figure 4D).

Power

Local field potential power was calculated with multi-tapered methods[87]. Signals were tapered with a single Slepian taper (TW = 3, K = 5). Spectrograms (Figure 5A) were generated by

470 sliding a 201 ms window by increments of 1 ms. Power spectra before (-200 to 0 ms) and after (0
471 to 200 ms) saccade onset were calculated with static variants of the same methods (Figure 5B). No
472 layer-specific differences were found, so results shown here are averaged across all recording
473 channels.

474 ***Pairwise Phase Consistency***

475 The pairwise phase consistency (PPC) is an estimate of spike-field locking that is not
476 biased by spike count or firing rate[89]. Two segments of data were analyzed: the 200 ms before and
477 after saccade onset. In order to obtain reliable estimates, only units that fired at least 50 spikes
478 cumulatively across all saccades during both time windows were considered. The spectro-temporal
479 representation of the local field potential signal was generated through a continuous wavelet
480 transform with a family of complex Morlet wavelets (spanning frequencies from 2 to 100 Hz, with
481 6 cycles at each frequency). Phase information was extracted at the time of each spike, and the
482 PPC was then calculated for each unit with each LFP channel in the laminar probes (Figure 5C).
483 We observed minimal change in the PPC as a function of LFP layer, so the results presented here
484 were averaged across all recording channels.

485 ***Computational Model***

486 The firing rate or mean field model consists of a network of six populations representing
487 excitatory (E) and inhibitory (I) populations in the superficial (L1), input (L2), and deep (L3) layers
488 of the cortex. The firing rate model describes temporal changes in the firing rates (r) of each
489 population as a function of the activity of two external inputs (i_{sacc}, the saccadic signal, and i_{stim},
490 the visual signal) and the activity of other populations in the network. The connectivity between
491 populations is depicted in Figure 6A. The network consists of a coupled E-I network in each layer
492 and captures the core aspect of connectivity between cortical layers: information flow from input

to superficial to deep layer. The saccadic signal is sent to both E and I populations in the input layer, while the visual signal is sent only to the E population in the input layer. The saccadic signal is described by a ramping function (Figure 6C, top). The visual signal is only activated in the second set of simulations, where it is held constant to replicate sustained visual input. The six-population rate model is described by the following equations:

$$\tau_E \frac{dr_{E_{L1}}}{dt} = -r_{E_{L1}} + G_E\left(W_{EE} \cdot r_{E_{L1}} - W_{EI} \cdot r_{I_{L1}} + W_{EE_{L12}} \cdot r_{E_{L2}}\right) \quad (1)$$

$$\tau_I \frac{dr_{I_{L1}}}{dt} = -r_{I_{L1}} + G_I\left(W_{IE} \cdot r_{E_{L1}} - W_{II} \cdot r_{I_{L1}} + W_{IE_{L12}} \cdot r_{E_{L2}}\right) \quad (2)$$

$$\tau_E \frac{dr_{E_{L2}}}{dt} = -r_{E_{L2}} + G_E\left(W_{EE} \cdot r_{E_{L2}} - W_{EI} \cdot r_{I_{L2}} + W_{EE_{SS}} \cdot i_{sacc} + i_{stim}\right) \quad (3)$$

$$\tau_I \frac{dr_{I_{L2}}}{dt} = -r_{I_{L2}} + G_I\left(W_{IE} \cdot r_{E_{L2}} - W_{II} \cdot r_{I_{L2}} + W_{IE_{SS}} \cdot i_{sacc}\right) \quad (4)$$

$$\tau_E \frac{dr_{E_{L3}}}{dt} = -r_{E_{L3}} + G_E\left(W_{EE} \cdot r_{E_{L3}} - W_{EI} \cdot r_{I_{L3}} + W_{EE_{L32}} \cdot r_{E_{L1}}\right) \quad (5)$$

$$\tau_I \frac{dr_{I_{L3}}}{dt} = -r_{I_{L3}} + G_I\left(W_{IE} \cdot r_{E_{L3}} - W_{II} \cdot r_{I_{L3}} + W_{IE_{L32}} \cdot r_{E_{L1}}\right) \quad (6)$$

$$G(x) = \begin{cases} 0 & \text{for } x < \theta \\ m(x-\theta)^{2.5} & \text{for } \theta < \mathrm{x} < \theta + 1/m \\ 1 & \text{for } x > \theta + 1/m \end{cases} \quad (7)$$

τ_E and τ_I are the rates at which the excitatory and inhibitory populations approach their steady states. G_E and G_I are the population response functions, described by equation (7), for the excitatory and inhibitory populations that transform the given inputs into a firing response. θ is the threshold input and m is the rate of the response function. Parameters tuned in the model to recapitulate the population level observations in experimental data: $W_{EE} = 25,\ W_{II} = 4,\ W_{EI} = 25, W_{IE} = 20, W_{EE_{L12}} = 20, W_{IE_{L12}} = 2, W_{EE_{SS}} = 0.5, W_{EI_{SS}} = 3, W_{EE_{L32}} = 30, W_{IE_{L32}} = 4, \tau_E =$

$0.008, \tau_I = 0.004, m_E = 0.001, m_I = 0.005, \theta_{E_{L1}} = -3, \theta_{E_{L2}} = -5, \theta_{E_{L3}} = -5, \theta_{I_{L1}} = 0,$ $\theta_{I_{L2}} = 8, \theta_{I_{L3}} = 8, i_{stim} = 10.$

A further tuning was applied to the model to switch each E-I population between an inhibitory stabilized network (ISN) regime and a non-ISN one[60]. Experimental data from visual cortex suggest that a cortical network switches to an ISN regime during stimulus processing[59], hence we included this feature to generate model predictions for saccadic suppression-related laminar activity during stimulus presentation. Previous modeling work has shown that a firing-rate E-I network model can switch between these regimes either by modification of connection weights (W), change in slope or rate of response functions (m) or a combination of both[60]. We implemented the simplest mechanism for the network to switch between ISN and non-ISN regime: nonlinear response functions. When visual stimulation (modeled as an excitatory input to the input layer E population) is applied to the model cortical column, the network moves to a higher slope region of its response functions, and switches to the ISN operating regime. A phase-plane analysis of the E-I network model illustrates how this results in a distinct response of the network to exactly the same saccadic signal input (Figure S5).

ACKNOWLEDGEMENTS

This research was supported by NIH R01 EY021827 to JHR and ASN, NARSAD Young Investigator Grant, Ziegler Foundation Grant and Yale Orthwein Scholar Funds to ASN, NIH R00 EY025026 to MPJ, and by NEI core grants for vision research P30 EY019005 to the Salk Institute and P30 EY026878 to Yale University. We would like to thank Catherine Williams and Mat LeBlanc for excellent animal care.

AUTHOR CONTRIBUTIONS

SD & ASN conceptualized the project. ASN collected the data and supervised the project. SD analyzed the data, with assistance from MPM and ASN. MPJ developed the computational model. SD, ASN, MPJ, and JHR wrote the manuscript.

COMPETING INTERESTS

The authors declare no competing interests.

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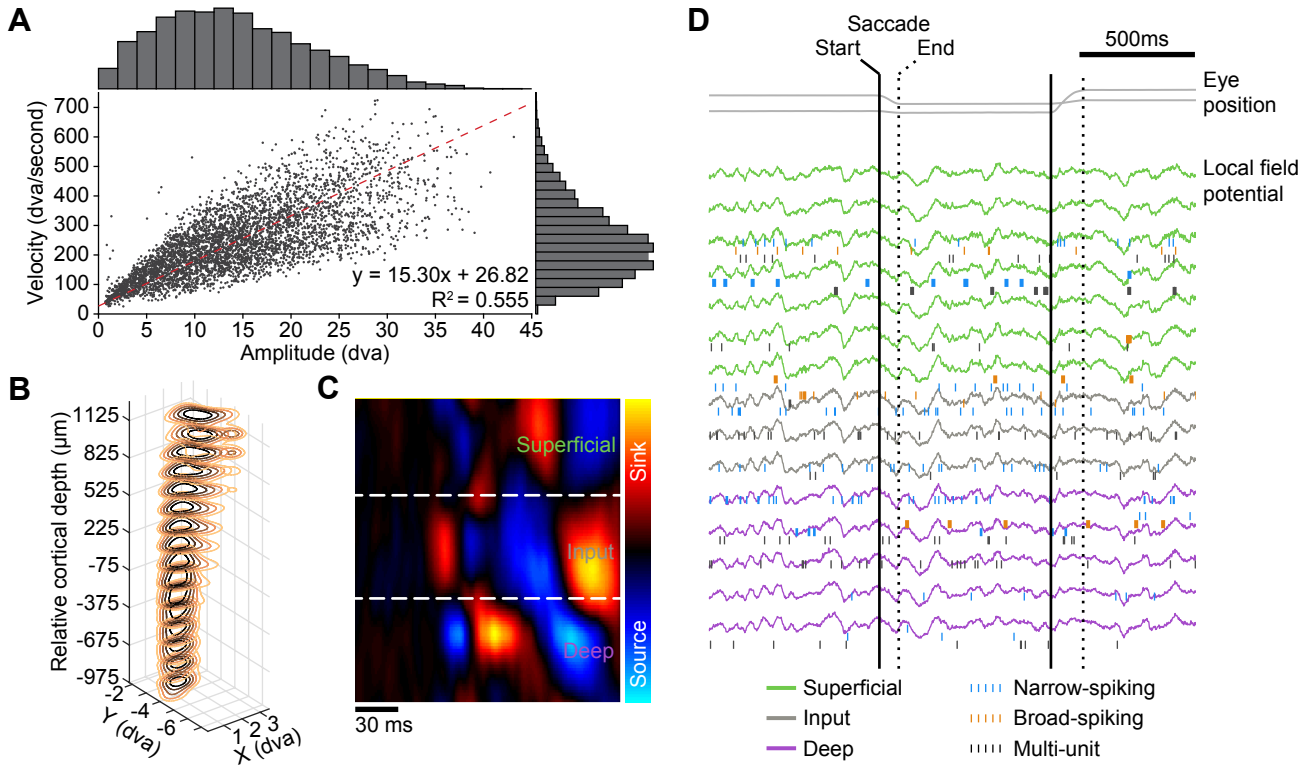

Figure 1. Laminar Recordings and Saccade Identification

(A) The amplitude-velocity relationship ('main sequence') for all saccades (n = 4420) in the dataset. An approximately linear trend is expected for saccades, as is found here (R^2 = 0.555). dva = degrees of visual angle.

(B) Stacked contour plot showing spatial receptive fields (RFs) along the laminar probe from an example session. The RFs are well aligned, indicating perpendicular penetration down a cortical column. Zero depth represents the center of the input layer as estimated with current source density (CSD) analysis.

(C) CSD displayed as a colored map. The x-axis represents time from stimulus onset; the y-axis represents cortical depth. The CSD map has been spatially smoothed for visualization.

(D) Example of continuous eye-tracker and electrophysiological recordings. Eye traces are plotted at the top in grey. LFP traces are plotted by depth below, and spikes are overlaid on the corresponding channel. LFP traces have been color coded by layer identity; spikes have been color coded by cell type.

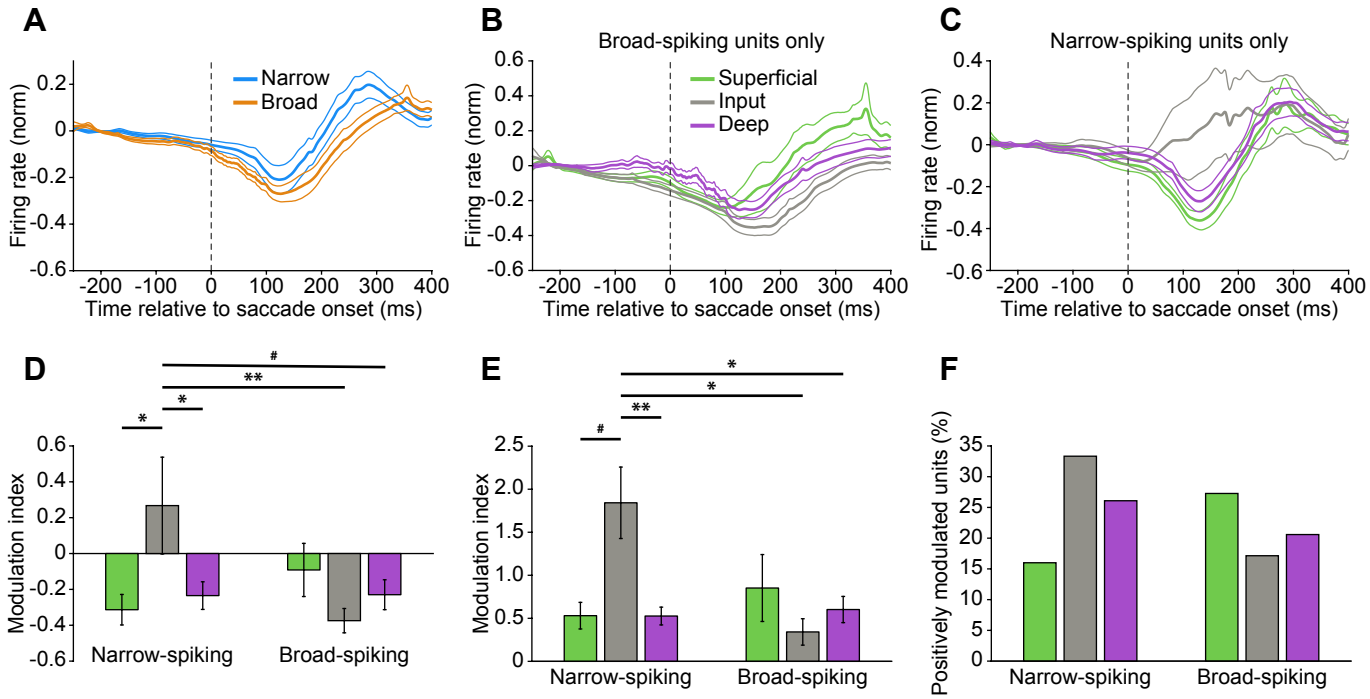

Figure 2. Narrow-Spiking Input Layer Units Display Peri-Saccadic Enhancement of Firing

(A) Average peri-saccadic normalized firing rate for broad- (n = 102 units) and narrow-spiking (n = 95 units) populations. Thin lines indicate standard error of the mean.

(B) As in (A), but including only broad-spiking units from the superficial (n = 33), input (n = 35), and deep (n = 34) layer populations.

(C) As in (A), but including only narrow-spiking units from the superficial (n = 25 units), input (n = 24 units), and deep (n = 46 units) layer populations.

(D) Average peri-saccadic modulation index for the six neural subpopulations. Two-way ANOVA (F_{layer} = 1.16, P = 0.3151; $F_{cell\ type}$ = 1.84, P = 0.1765; $F_{interaction}$ = 5.98, P = 0.0030) with Tukey's multiple comparison tests (narrow-spiking superficial vs narrow-spiking input, $^{*}P$ = 0.0423; narrow-spiking input vs narrow-spiking deep, $^{*}P$ = 0.0495; narrow-spiking input vs broad-spiking input, $^{**}P$ = 0.0070; narrow-spiking input vs broad-spiking deep, $^{\#}P$ = 0.0817). Error bars indicate standard error of the mean.

(E) Average peri-saccadic modulation index for positively modulated units from the six neural subpopulations. Two-way ANOVA (F_{layer} = 1.87, P = 0.1666; $F_{cell\ type}$ = 2.32, P = 0.1355; $F_{interaction}$ = 5.51, P = 0.0077) with Tukey's multiple comparison tests (narrow-spiking superficial vs narrow-spiking input, $^{\#}P$ = 0.0833; narrow-spiking input vs narrow-spiking deep, $^{**}P$ = 0.0073; narrow-spiking input vs broad-spiking input, $^{*}P$ = 0.0106; narrow-spiking input vs broad-spiking deep, $^{*}P$ = 0.0384). Error bars indicate standard error of the mean.

(F) Proportion of units in each neural subpopulation with a positive peri-saccadic modulation index.

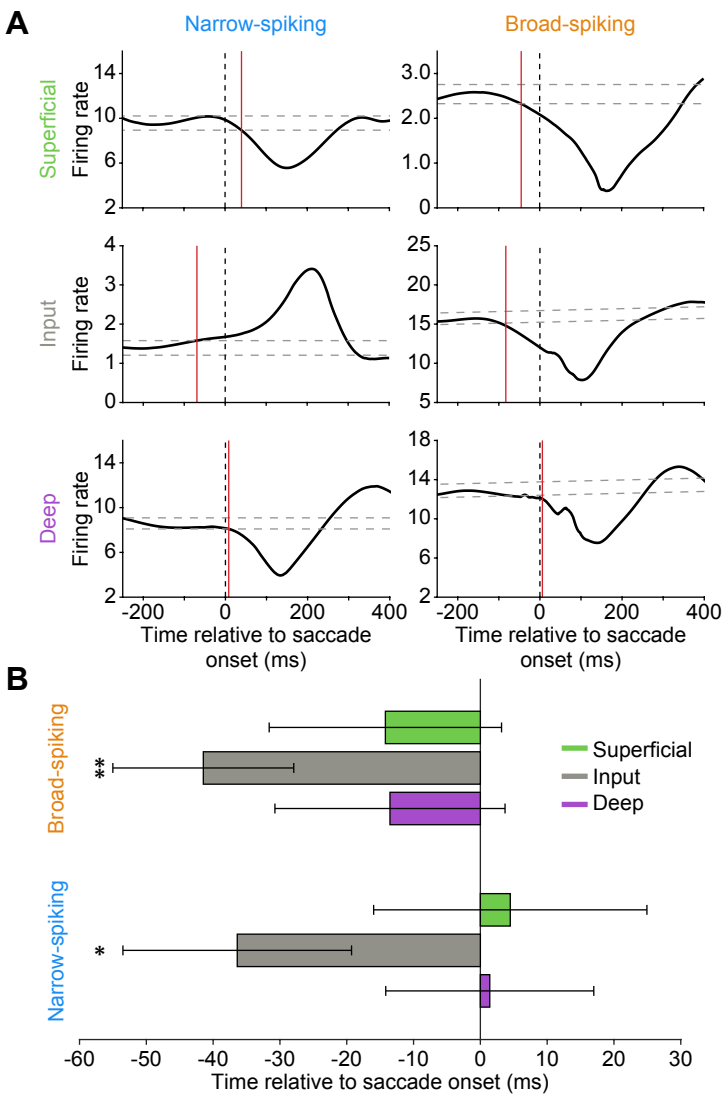

Figure 3. Input Layer Units Display Modulation Prior to Saccade Onset

(A) Six example single-units demonstrating the approach for identifying the time of significant peri-saccadic firing rate modulation. 95% confidence intervals were calculated for the baseline activity of each unit (dashed grey lines) and the first deviation outside of those bounds was marked as the time of first significant response (red line). The example units shown are the same as those from Figure S1.

(B) Average timing of first significant peri-saccadic modulation for the six neural subpopulations. Input layer units, both broad- and narrow-spiking, show modulation prior to saccade onset. Two-tailed one-sample t-test (broad-spiking input layer, $^{**}P = 0.0049$; narrow-spiking input layer, $^{*}P = 0.0492$). Error bars indicate standard error of the mean.

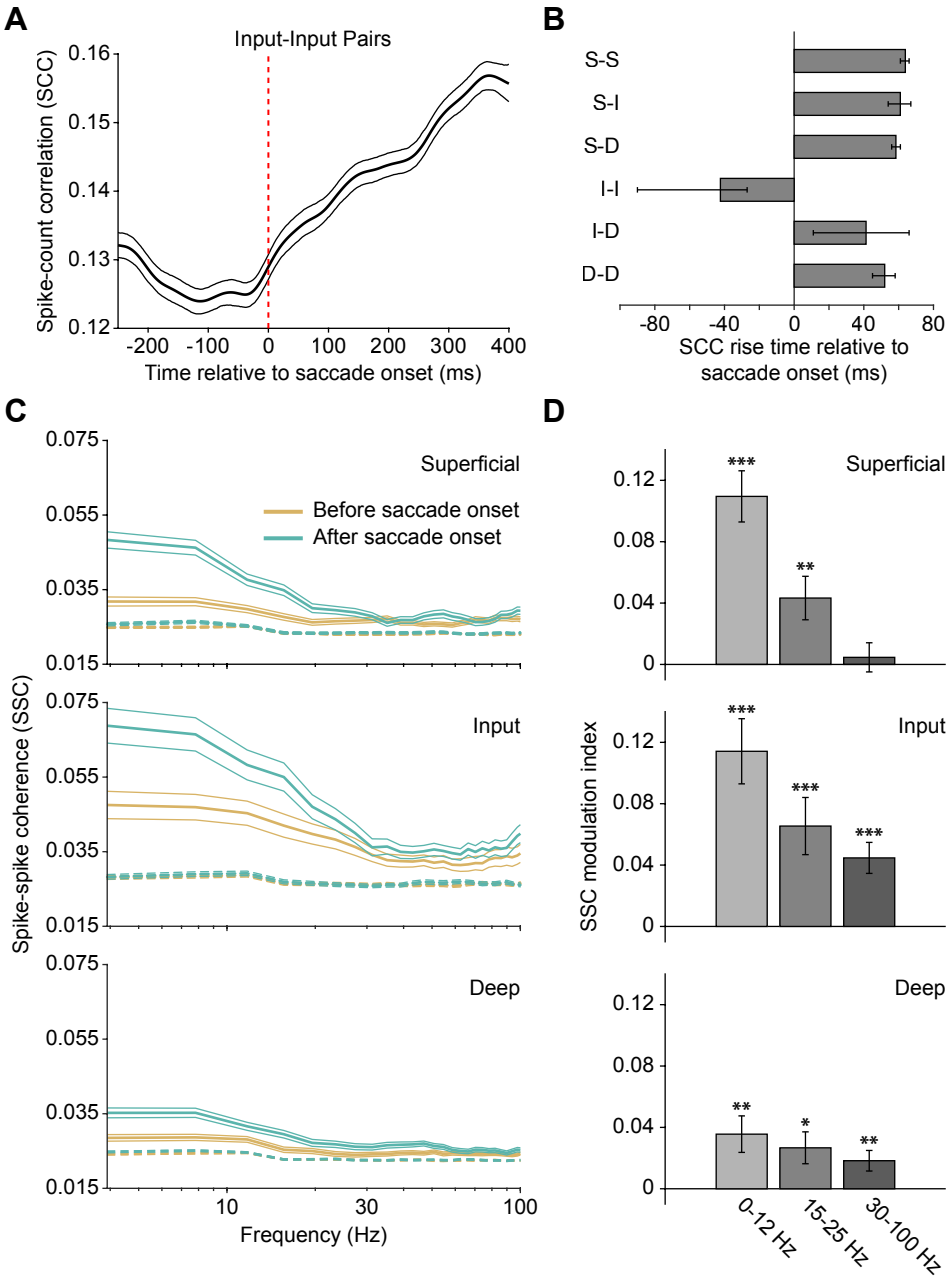

Figure 4. Saccade Onset Increases Correlated Variability Most Prominently in the Input Layer

(A) Spike count correlations as a function of time relative to saccade onset for input-input unit pairs (n = 240 pairs). Correlations were calculated with a sliding 201 ms window. Thin lines indicate bootstrapped 95% confidence intervals.

(B) Average start time of spike count correlation rise relative to saccade onset. Y-axis labels indicate layer identity of unit pairs (S - superficial; I - input; D - deep). Error bars represent bootstrapped 95% confidence intervals. Input-input pairs begin to rise before saccade onset and before all other pair combinations. All other pairs respond after saccade onset.

(C) Average spike-spike coherence between pairs of simultaneously recorded units (single-units and multi-units) before (-200 to 0 ms) and after (0 to 200 ms) saccade onset as a function of frequency. Superficial layer, n = 299 pairs; input layer, n = 191 pairs; deep layer, n = 604 pairs. Thin lines indicate standard error of the mean. Dashed lines indicate spike-spike coherence for shuffled data.

(D) Same data as in (C), but averaged across three frequency bands and plotted as a modulation index: $(SSC_{after} - SSC_{before})/(SSC_{after} + SSC_{before})$. Two-tailed one-sample t-tests (Superficial 0-12 Hz, ***$P = 1.64\times10^{-10}$; Superficial 15-25 Hz, **$P = 0.0027$; Superficial 30-100 Hz, $P = 0.5468$; Input 0-12 Hz, ***$P = 2.2131\times10^{-7}$; Input 15-25 Hz, ***$P = 5.5458\times10^{-4}$; Input 30-100 Hz, ***$P = 1.6503\times10^{-5}$; Deep 0-12 Hz, **$P = 0.0013$; Deep 15-25 Hz, *$P = 0.0116$; Deep 30-100 Hz, **$P = 0.0050$). Error bars indicate standard error of the mean.

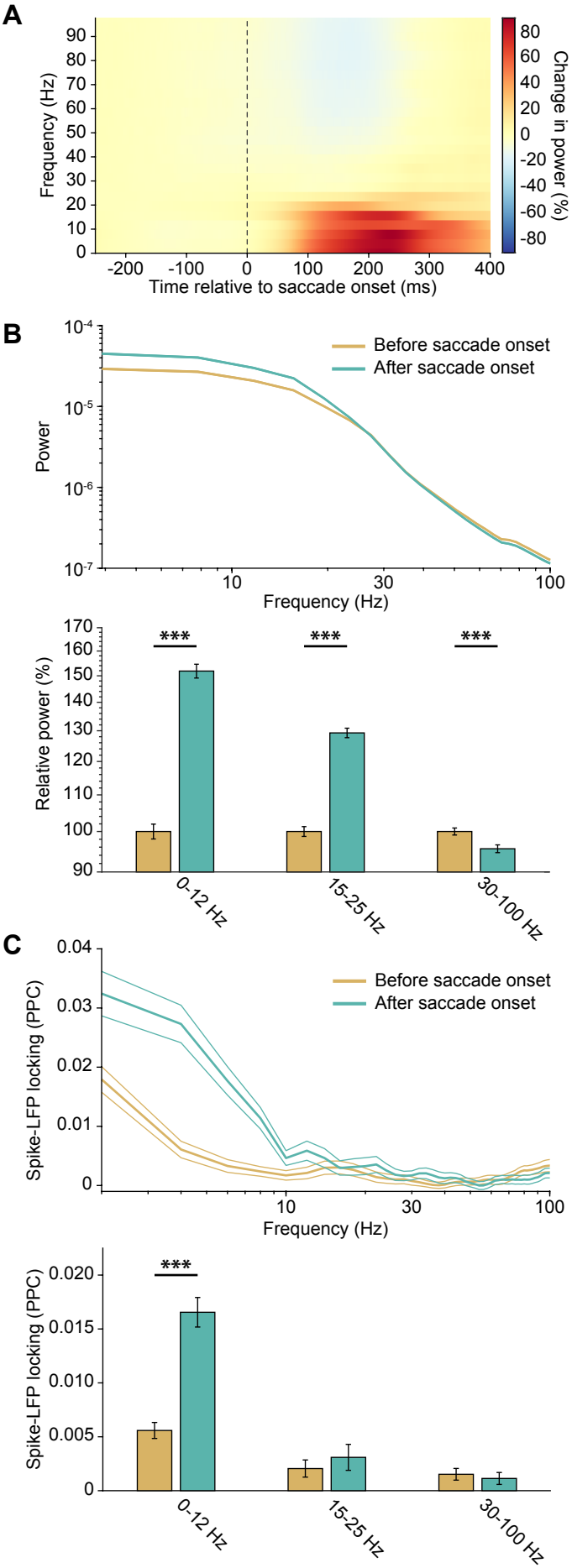

Figure 5. Saccade Onset Increases Low Frequency Power and Spike-LFP Locking

(A) Time-frequency representation (spectrogram) of LFP power around saccade onset. Signal strength is represented as percent change in power (i.e. the raw power at each timepoint is divided by the average power in the baseline period, defined here as 250 to 150 ms before saccade onset).

(B) Top: Average power spectra before (-200 to 0 ms) and after (0 to 200 ms) saccade onset. Thin lines indicate standard error of the mean. Bottom: Same data as in top, but averaged across three frequency bands and normalized to data before saccade onset for visualization. Two tailed paired-sample t-tests (0-12 Hz, ***$P = 3.4669\times10^{-92}$; 15-25 Hz, ***$P = 9.3690\times10^{-74}$; 30-100 Hz, ***$P = 2.7322\times10^{-9}$). Error bars indicate standard error of the mean.

(C) Top: Average spike-LFP locking spectra before (-200 to 0 ms) and after (0 to 200 ms) saccade onset. Thin lines indicate standard error of the mean. Bottom: Same data as in top, but averaged across three frequency bands. Two tailed paired-sample t-tests (0-12 Hz, ***$P = 1.7147\times10^{-19}$; 15-25 Hz, $P = 0.1598$; 30-100 Hz, $P = 0.1091$). Error bars indicate standard error of the mean.

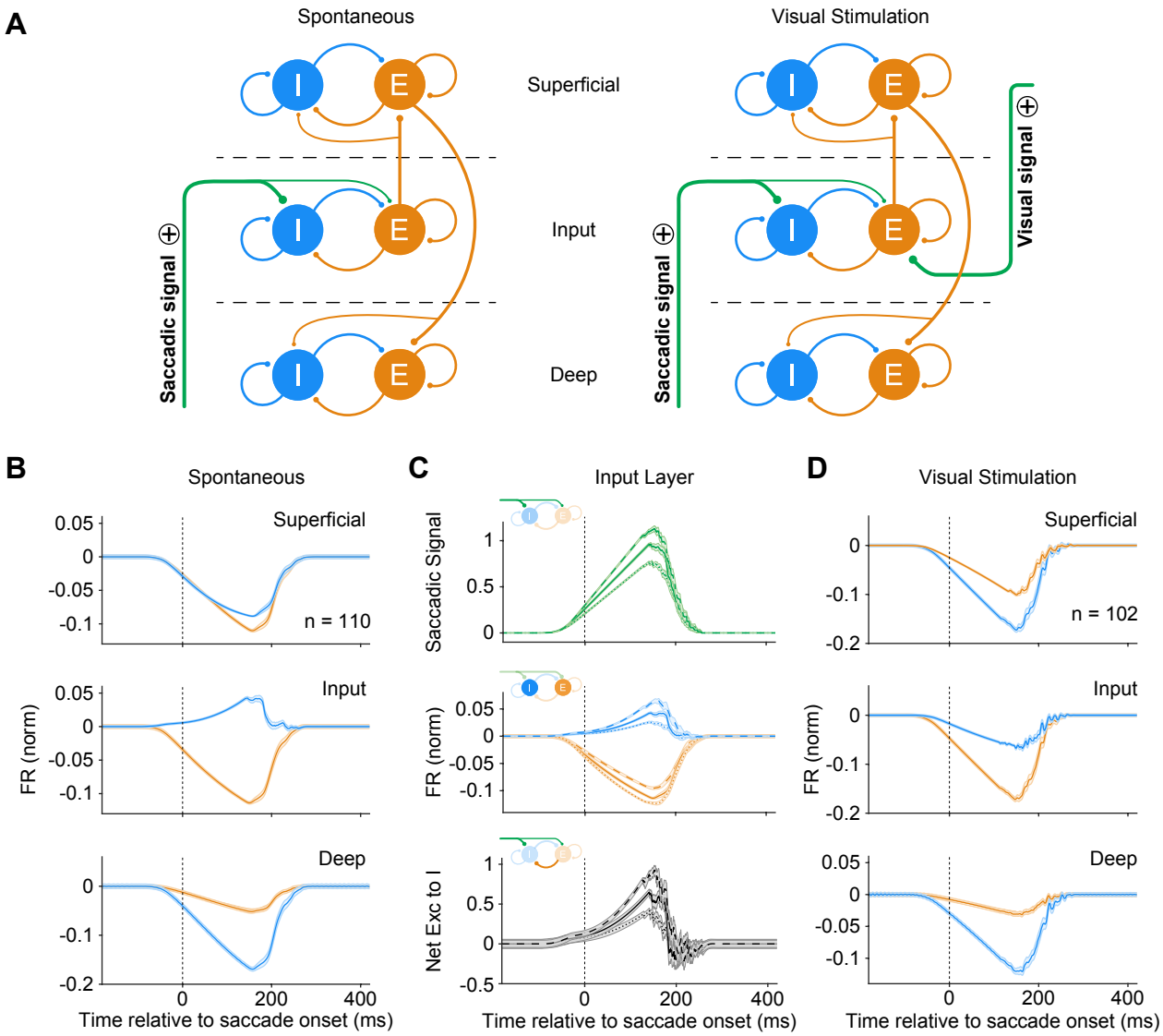

Figure 6. A Computational Model Confirms that Activation of an Input Layer-Targeting Projection Induces Saccadic Suppression

(A) Connectivity between populations in our computational model. Excitatory (E) and inhibitory (I) populations within each layer project to each other as well as themselves. The input layer excitatory population projects to the superficial layer, while the superficial excitatory population projects to the deep layer. In our model of spontaneous saccades (left), we activate an external excitatory projection that selectively targets the input layer. In our model of saccades executed in the presence of visual stimuli (right), we add an additional input that selectively targets input layer excitatory neurons.

(B) Normalized peri-saccadic firing rate of simulated neural subpopulations during spontaneously executed saccades (n = 110 simulated saccades). The input layer inhibitory subpopulation shows enhancement of firing, while all other subpopulations show suppression.

(C) Further dissection of input layer activity. (Top) The saccadic signal arriving in the input layer was simulated with a ramping function. Represented here are saccadic signals of three different strengths. From weakest to strongest, these are represented by dotted, continuous, and dashed lines, both here and in the following subpanels. (Middle) The excitatory and inhibitory input layer subpopulation firing rates in response to three saccadic signals of varying strength. (Bottom) Net excitatory drive onto the inhibitory input layer subpopulation in response to three saccadic signals of varying strength. The excitatory drive is the sum of drive from the external saccadic input and drive from the local excitatory population.

(D) Normalized peri-saccadic firing rate of simulated neural subpopulations during saccades executed in the presence of visual stimulation (n = 102 simulated saccades). Visual stimulation is represented by the activation of a second input, which selectively targets the input layer excitatory subpopulation. Here, all subpopulations within the cortical column show peri-saccadic suppression.

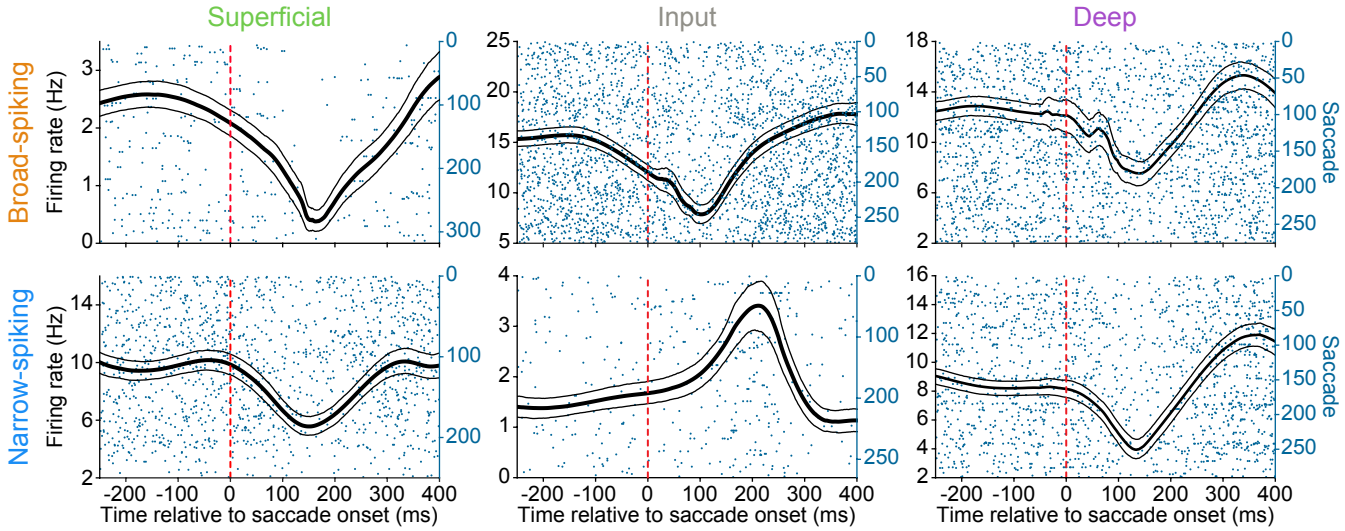

Figure S1. Peri-Saccadic Single-Unit Firing Rate Estimation

Peri-saccadic single-unit firing rates were estimated by smoothing individual spikes with gaussian kernels. The kernel bandwidth was selected for each unit at each point in time using an optimization algorithm that minimizes the mean integrated squared error[45]. The firing rate estimates produced by this approach are shown for six representative example units, one from each recorded neural subpopulation. The smoothed firing rates (black) are overlaid on the raster plots showing raw spikes (blue). In the raster plot, each row represents a single saccade. Thin lines are bootstrapped 95% confidence intervals of the firing rate estimate.

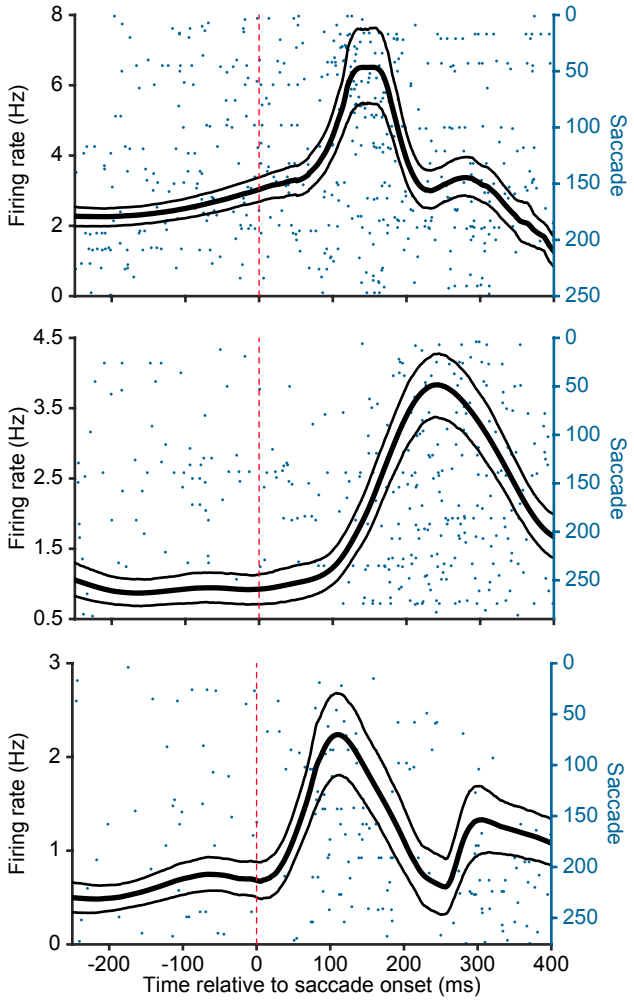

Figure S2. Additional Narrow-Spiking Input Layer Single-Unit Examples

Additional example narrow-spiking input layer single-units that display peri-saccadic enhancement of firing. Firing rates were estimated using a kernel smoothing approach and bandwidth optimization algorithm[45]. The smoothed firing rates (black) are overlaid on the spike raster plots (blue). Thin lines are bootstrapped 95% confidence intervals of the firing rate estimate.

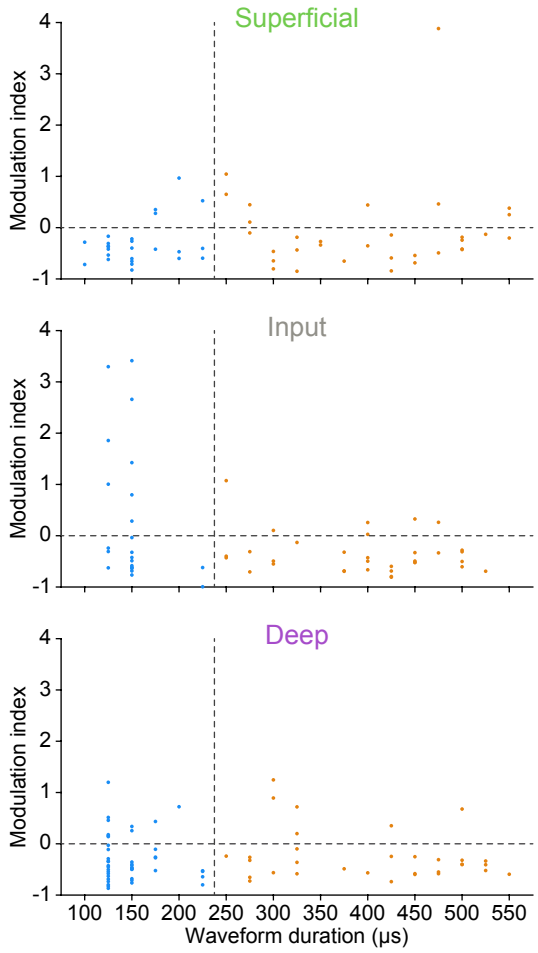

Figure S3. Single-Unit Modulation Indices
Single-unit modulation indices as a function of spike waveform duration. Neural populations from each layer are displayed in different subplots. The dashed horizontal line represents a modulation index of zero (no firing rate change), while the dashed vertical line represents the threshold separating narrow- and broad- spiking units.

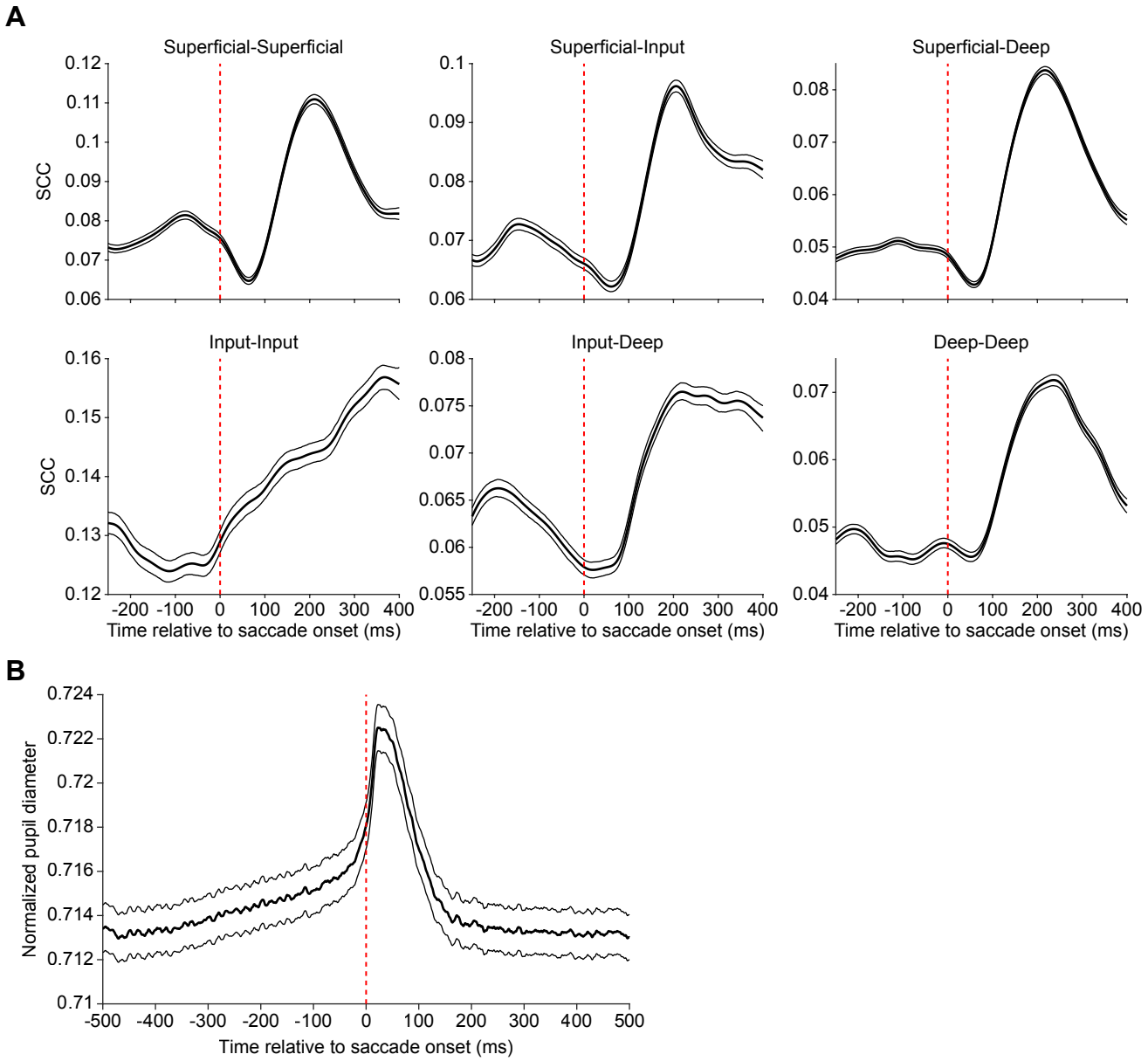

Figure S4. Input Layer Pairs Show Increases in Spike Count Correlations Prior to Saccade Onset
(A) Spike count correlations as a function of time relative to saccade onset for all combinations of layer-wise unit pairs. Superficial-superficial, *n* = 442 pairs; superficial-input, *n* = 558 pairs; superficial-deep, *n* = 755 pairs; input-input, *n* = 240 pairs; input-deep, *n* = 501 pairs; deep-deep, *n* = 680 pairs. Correlations were calculated with a sliding 201 ms window. Thin lines indicate bootstrapped 95% confidence intervals.
(B) Normalized pupil diameter around the time of saccade onset (*n* = 4420 saccades). Thin lines indicate standard error of the mean.

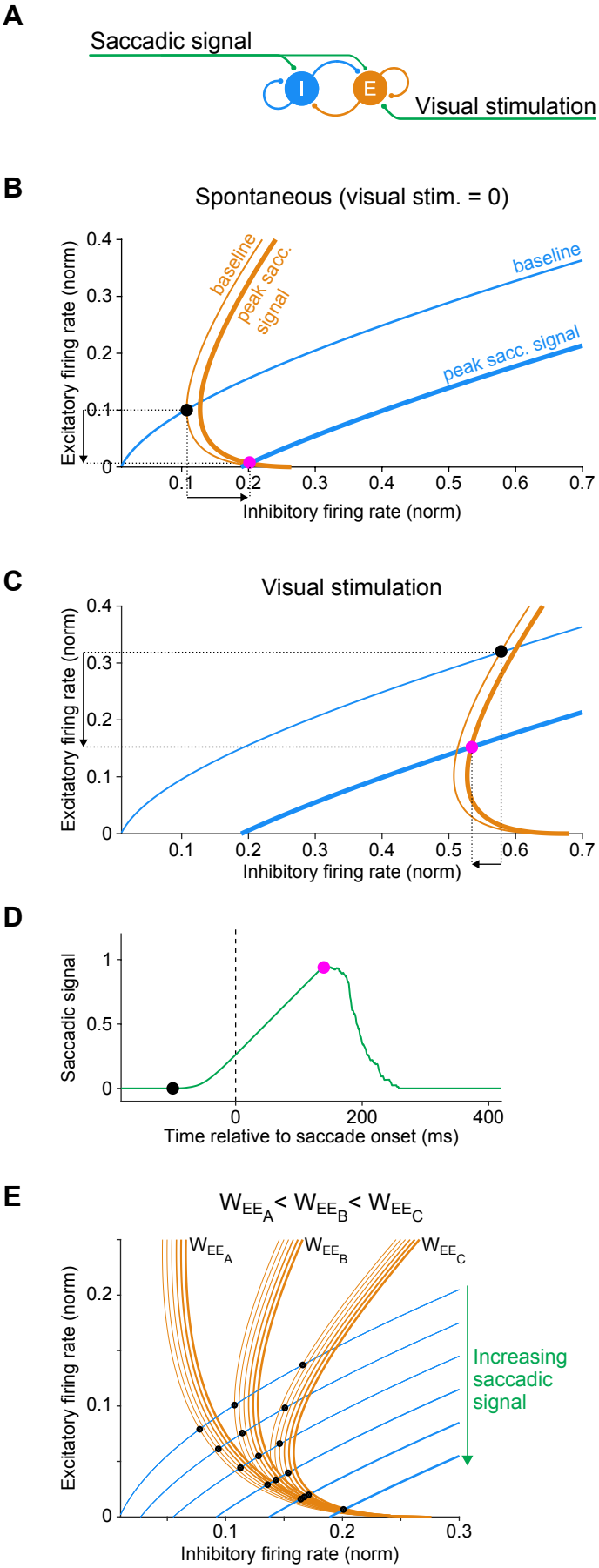

Figure S5. Phase Plane Analysis of the Input Layer E-I System

(A) The E-I network in the input layer.

(B) The nullclines and stable activity levels of the input layer E and I populations during simulation of spontaneous activity. Equations and parameters are the same as those described in the Methods and used in Figure 6. The orange and blue curves show the E and I nullclines, respectively, of the E-I network in (A). Each point on a nullcline indicates the steady state firing rate of the E or I population when the activity of the other population is fixed, i.e. the rate of change in equations (3) or (4) is zero. The ISN and non-ISN regimes of the network are demarcated by the switch between positive and negative slopes along the E nullclines. This ability to switch is a consequence of strong E-E connectivity, and is absent when the E-E connectivity strength is either weak or zero. The thin nullclines depict firing rates when saccadic input is at baseline; the thick nullclines depict firing rates when saccadic input increases by a non-zero value. The intersections of the E and I nullclines indicate the steady state firing rates at which the network stabilizes if allowed to seek equilibrium (indicated by black and pink dots at baseline and during peak saccadic suppression, respectively). As illustrated here, the nullclines for our model intersect in the non-ISN regime in the absence of visual input, causing E and I activity to shift in opposite directions in response to the saccadic signal.

(C) Same as in (B), but during simulated visual stimulation. The addition of excitatory input to the E population shifts the E nullclines to the right, causing them to intersect with the I nullclines in the ISN regime. As illustrated here, this causes the E and I populations to shift in the same direction in response to the saccadic signal.

(D) The approximate level of saccadic signal strength that corresponds to the E (orange) and I (blue) nullcline sets shown in (B) and (C). The phase plane analysis is most applicable to constant levels of saccadic signal, and is only an approximate depiction of the scenario of a ramp signal shown in (D), and as used in the model simulation in Figure 6.

(E) Effect of E-E connection strength on inhibitory firing rate changes in response to the saccadic signal. Each set of E nullclines corresponds to a given E-E connection strength (W_{EE}), and illustrates the effects of raising saccadic signal magnitude (indicated by greater line thickness), causing either increases or decreases in steady state firing rates (black dots), as determined by their intersection with the I nullclines. This phase plane analysis approximately predicts three ways in which I firing rate can shift in response to increasing saccadic input: 1) it increases quickly in response to a small increase in saccadic signal (W_{EE_A}), 2) it increases minimally until a threshold level of saccadic signal is reached (W_{EE_B}), or 3) it first decreases then increases with rising saccadic signal strength (W_{EE_C}). The operating regime shown in the simulation results (Figure 6) that recapitulate experimental observations corresponds to the E-E connection strength illustrated by W_{EE_B}. In this case, the network is maintained at the boundary between ISN (positive E nullcline slope) and non-ISN (negative E nullcline slope) regimes in the absence of the saccadic signal.