



A defined network of fast-spiking interneurons in orbitofrontal cortex: responses to behavioral contingencies and ketamine administration

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Orbitofrontal cortex (OFC) is a region of prefrontal cortex implicated in the motivational control of behavior and in related abnormalities seen in psychosis and depression. It has been hypothesized that a critical mechanism in these disorders is the dysfunction of GABAergic interneurons that normally regulate prefrontal information processing. Here, we studied a subclass of interneurons isolated in rat OFC using extracellular waveform and spike train analysis. During performance of a goal-directed behavioral task, the firing of this class of putative fast-spiking (FS) interneurons showed robust temporal correlations indicative of a functionally coherent network. FS cell activity also co-varied with behavioral response latency, a key indicator of motivational state. Systemic administration of ketamine, a drug that can mimic psychosis, preferentially inhibited this cell class. Together, these results support the idea that OFC–FS interneurons form a critical link in the regulation of motivation by prefrontal circuits during normal and abnormal brain and behavioral states.

Keywords: schizophrenia, depression, fast-spiking interneuron, motivation, parvalbumin, psychosis

INTRODUCTION

Ketamine is an NMDA antagonist with potent psychoactive properties (Berman et al., 2000; Zarate et al., 2006). Ketamine's effects may be mediated in part through the orbitofrontal cortex (OFC), a region of prefrontal cortex (PFC) crucial to motivational aspects of goal-directed behavior and affective decision-making (Hollerman et al., 2000; Schoenbaum and Roesch, 2005). Electrophysiological studies have demonstrated that the activity of single OFC neurons is correlated with the predicted outcomes of behavioral decisions (Schoenbaum et al., 1998; Tremblay and Schultz, 1999, 2000; Padoa-Schioppa and Assad, 2006). OFC lesions also disrupt the expression of differences in latency of response for differences in the value of expected outcomes (Lauwereyns et al., 2002; Schoenbaum et al., 2003). However, although individual OFC neurons show firing rate

correlations with response latency, no overall correlation of OFC and response latency has been seen (Feierstein et al., 2006). Indeed, in addition to prominent correlation with expected outcomes, OFC neurons, like other cortical neurons, show a highly diverse set of behavioral correlates (Feierstein et al., 2006).

Cortical microcircuits are composed of distinct cell types with diverse intrinsic and synaptic properties (Somogyi et al., 1998; Markram et al., 2004; Ascoli et al., 2008). Seminal studies by Goldman-Rakic and colleagues dissected the functional organization of primate dorsal lateral PFC in terms of cell types (Constantinidis et al., 2002; Wang et al., 2004). This work supports the notion that the firing properties of prefrontal neurons may be attributed to their underlying role in the function of the local microcircuit, with the apparent diversity of responses reflecting the

heterogeneity of circuit elements and a failure to differentiate cell types. The significance of understanding the functional diversity of OFC neurons is highlighted by the fact that different cell types within PFC are differentially impacted by neurological diseases and psychiatric disorders (Freund, 2004; Lewis et al., 2005).

Fast-spiking (FS) GABAergic-interneurons are of particular importance to cortical function. FS cells comprise approximately 50% of interneurons in the neocortex, generate narrow action potentials, include both basket and chandelier cells, and are often parvalbumin containing (Kawaguchi and Kubota, 1993). Given their cellular and anatomical properties, FS cells are well positioned to coordinate the spiking activity of cortical pyramidal cells, and several psychiatric disorders including both schizophrenia and bipolar disorder are associated with FS cell dysfunction (Haldane and Frangou, 2004; Lewis et al., 2005). Furthermore, it has been hypothesized that FS interneurons are particularly sensitive to the inhibiting effects of NMDA antagonists such as ketamine and MK-801 (Coyle, 2004). This hypothesis could explain why these inhibitors of excitatory synaptic transmission produce overall disinhibition of cortical activity, including increases in cellular excitability and enhanced glutamate release (Lewis and Moghaddam, 2006).

Recent studies have provided support for the disinhibition hypothesis of NMDA antagonism by showing that systematic MK-801 administration decreases the firing rate of putative interneurons in the PFC of awake rats (Homayoun and Moghaddam, 2007a,b). Furthermore, studies by Goldman-Rakic and colleagues have described the activity of putative interneurons during behavior in both normal and during pharmacological manipulations (Constantinidis et al., 2002; Williams et al., 2002). In these studies, narrow-spiking (NS) interneurons were identified based on a combination of waveform and firing rate criteria. Here, we use a similar approach based on extracellular waveform analysis to identify two physiologically distinct NS OFC cell populations, termed NS1 and NS2. These two populations differ in their local and global network properties and have distinct behavioral correlates in a sensory decision task. Importantly, NS1 cells are selectively suppressed by subanesthetic doses of ketamine. Together, these data provide insight into GABAergic processing within OFC and into how manipulations of GABAergic processing may contribute to the “psychotomimetic” and antidepressant properties of acute ketamine administration.

MATERIALS AND METHODS

All procedures involving animals were carried out in accordance with NIH standards and approved by the Cold Spring Harbor Institutional Animal Care and Use Committee. Fifteen male Long-Evans rats (300–500 g) were implanted with a microdrive array consisting of six independently adjustable tetrodes (Wilson and McNaughton, 1993; Feierstein et al., 2006) (Bregma coordinates: 3.5 mm AP; 2.5 mm L). Following 5–7 days of post-operative recovery, tetrodes were advanced into OFC and physiological recordings were obtained from implanted animals as they performed a two-alternative olfactory discrimination task for water reward (Uchida and Mainen, 2003; Feierstein et al., 2006) or explored a holding box adjacent to the testing environment. During holding box exploration, the animal was allowed to freely wander a clean Plexiglas cage; over the course of exploration, animals would most

often walk around the cage, sniffing and probing the borders of the cage with its whiskers, with occasional bouts of motionlessness and quiescence near the sides of the cage. During the two-alternative choice olfactory discrimination task, animals were trained to make a nose poke into one of two spatially separated (“left” and “right”) response ports based upon the identity of an odor presented from a centrally located odor delivery port. Over the course of behavioral training, animals would learn to associate a monomolecular odorant (e.g., hexanol) with a movement to the left or right response port; a correct response resulted in the receipt of a small water reward (approximately 0.03 ml). Animals were trained on this task until they were able to perform at or above criterion (85% correct) in a single session (approximately 200–300 individual trials), after which animals were implanted with a tetrode drive for electrophysiological recording during behavioral performance. While the OFC is not known to be necessary for the performance of this task, the OFC has been implicated in the process by which stimuli become associated with reward, and neurons in OFC have been shown to encode a number of variables involved in the performance of the two-alternative olfactory task (e.g., movements to choice ports, reward, odor identity) (Feierstein et al., 2006). Extracellular signals were amplified (5,000–50,000×), band-pass filtered (0.6–6 kHz) and sampled at 32.5 kHz/channel using a Cheetah acquisition system (Neuralynx, Tucson, AZ, USA). Individual threshold values were set for each tetrode channel such that any time a recorded voltage crossed threshold on at least one channel of a tetrode, 32 points were recorded for each of the four tetrode leads. Sampled waveforms, along with event flags generated by a behavioral control system, were subsequently saved to disk for further off-line analysis.

CELL IDENTIFICATION

Multiple single units were isolated by clustering spike features derived from the sampled waveforms (Mclust 3.4; A.D. Redish). Initial cluster identification and discrimination was based on the relative ratios of each spike’s peak amplitudes recorded across the four tetrode leads. Further refinement of spike clusters was primarily obtained through the use of additional features derived from the tetrode channel(s) yielding the strongest amplitude signal. Typical features included: spike width, spike valley, the fast Fourier transform (FFT) of the spike waveform, energy of the waveform (L1-norm) and the first principal component of the extracted waveforms (PCA1). Only cells that produced a minimum of 100 spikes with less than 1% refractory period violations (refractory period <1 ms) were used for subsequent analysis. In addition, to insure that we used well-isolated cells, cells with an isolation distance (ID, Mahalanobis distance for clusters using the following features: peak, energy, FFT, PCA1; Schmitzer-Tobert et al., 2005) of less than 20 were excluded (Harris et al., 2001).

SPIKE SHAPE ANALYSIS

To analyze the average spike shape of each isolated cell, only the tetrode channel yielding the largest voltage signal was considered. Each spike waveform within a cluster was normalized by the peak amplitude of the largest spike within a cluster and the peaks of all spikes were temporally aligned prior to averaging. Following averaging, the derived waveform was smoothed using a spline interpolation function (MATLAB: SPLINE) in which 10 points were

160 estimated for every data point in the original waveform and the fol-
 161 lowing features were extracted: (1) Pre-valley (V1): defined as the
 162 minimum value of the waveform prior to the peak; (2) Post-valley
 163 (V2): defined as the minimum value of the waveform following the
 164 peak; (3) Spike width: defined as the time between the occurrence of
 165 the peak and V2; (4) the ratio of V1 to V2 (V1:V2); and (5) the ratio
 166 of V2 to the spike's peak (V2:P).

167 ACTIVITY MEASUREMENTS

168 Cortical interneurons exhibit different excitability levels and
 169 regularity in firing in response to constant current injection. As sur-
 170 rogates for these *in vitro* measurements, we used mean firing rate over
 171 the whole recording session and a measurement of interspike inter-
 172 val (ISI) variability (local coefficient of variability, CV_2) (Holt et al.,
 173 1996). For a spike sequence, $CV_2 = 2 \cdot |\Delta t_{i+1} - \Delta t_i| / (\Delta t_{i+1} + \Delta t_i)$
 174 where Δt_i and Δt_{i+1} are adjacent ISIs. CV_2 was used as a measure
 175 of spike train variability rather than conventional measurements
 176 such as the coefficient of variation of ISIs and Fano Factor, because
 177 CV_2 is less susceptible to non-stationarities in firing rate typically
 178 associated with phasically driven spike trains (Holt et al., 1996).

179 SPIKE BROADENING

180 Activity dependent spike broadening (SB) was calculated as the
 181 average fractional change in spike width for pairs of adjacent spikes.
 182 Because the magnitude of SB depends on the time between spikes
 183 (Fee et al., 1996; Harris et al., 2001), only spike pairs in which the
 184 time between the two spikes was 1–10 ms were considered.

185 HIERARCHICAL CLUSTERING

186 To distinguish NS from wide-spiking (WS) cells, we performed
 187 unsupervised clustering on spike width and firing rate. Because the
 188 distribution of firing rates across the population was approximately
 189 log-normal, log firing rates rather than raw rate values were used so
 190 that outliers in the rate distribution did not distort variance esti-
 191 mates for the population. Following normalization to z -scores, and
 192 rescaling (for log-normal distributions), unsupervised clustering
 193 was performed using Ward's method (MATLAB Statistics Toolbox;
 194 Mathworks, Natick, MA, USA). We first calculated the Euclidean
 195 distance between all cell pairs based on the two-dimensional space
 196 defined by each cell's average spike width and firing rate. Based on this
 197 matrix of distances, an iterative agglomerative procedure was used to
 198 combine cells into groups such that at each stage the total number
 199 of groups was reduced by merging those groups whose combina-
 200 tion gave the smallest possible increase in the within group sum of
 201 squared deviation (Cauli et al., 2000). A similar clustering procedure
 202 was used to cluster populations of NS interneurons. For interneuron
 203 sub-classification, rather than spike width and firing rate, for cluster
 204 features we used CV_2 , fractional SB, V2:P and log(V1:V2).

205 CORRELATION ANALYSIS

206 Cross-correlation histograms (CCHs) for all pairs of simultaneously
 207 recorded cells were computed using techniques proposed by
 208 Brillinger and employed by Frank et al. (2001). Briefly, individual
 209 CCHs were constructed for each cell pair (A, B) with spikes from
 210 one cell (B) serving as the reference point for the CCH. The CCH
 211 was then subjected to a variance normalizing transformation:
 212 $\hat{x}_i = \sqrt{x_i / (dt \cdot N_B)}$ where x_i is the original value for time bin i , \hat{x}_i

213 is the new value, dt is the size of the time bin (2 ms), and N_B is
 214 the number of spikes fired by the reference cell. After this transforma-
 215 tion, the expected correlation value for the two independent spike
 216 trains over an entire recording session is $\langle \hat{x} \rangle = \sqrt{N_A / T}$, where T
 217 is the duration of the recording session and the 95% confidence
 218 intervals for the correlation are $\langle \hat{x} \rangle \pm \sqrt{1 / (dt \cdot N_B)}$. To determine the
 219 coupling probabilities for pairs of cortical neurons independent of
 220 slow drifts or co-variations in firing rate, a local mean value for
 221 each CCH was calculated based on the average of all bins within
 222 $\pm 1,000$ ms. For averaging purposes, each bin within a CCH was then
 223 transformed into a z -score relative to this local mean. Because of
 224 sampling issues, we were unable to distinguish the occurrence of
 225 two simultaneous spikes on the same tetrode; as a result, within-
 226 tetrode cross-correlations contained an artificial trough at time 0. To
 227 account for this trough, we replaced zero bin values with the mean
 228 value of the two adjacent bins prior to z -score normalization and
 229 averaging. In some analyses, high-pass filtering was used to remove
 230 slow time scale components of each CCH. For high-pass filtering,
 231 we constructed an ideal high-pass filter using the REMEZ function
 232 in MATLAB with the following parameters: Stop band freq = 0.01;
 233 Pass band frequency = 0.1; Stop band attenuation = 0.0001; Pass
 234 band ripple = 0.058, and Density factor = 16. A cutoff period of
 235 10 bins, corresponding to 20 ms, was used. Filtered CCHs were
 236 renormalized prior to averaging.

237 To quantify CCH interactions at the population level, significant
 238 increases in firing probability were detected for each CCH using a
 239 threshold of three standard deviations above the mean. To test for
 240 excitatory interactions, the CCH peak was required to be asym-
 241 metric and to occur 2–5 ms before the reference (postsynaptic)
 242 spike train. For correlations between and within NS groups, four
 243 different categories of correlations were identified: positive, nega-
 244 tive, mixed, and no-correlation. Positively (negatively) correlated
 245 cell pairs exhibited one or more bin within ± 5 ms of the refer-
 246 ence event that was greater (less) than three standard deviations
 247 above (below) the mean. Whereas positive/negative cell pairs had
 248 both a significant peak and trough within the ± 5 ms window, non-
 249 correlated cell pairs had no significant deviation from 0.

250 KETAMINE EXPERIMENTS

251 Each day, two baseline recording sessions and one post-injection
 252 recording session were performed. Baseline recording sessions
 253 consisted of a 15-min "resting" baseline, during which the rat was
 254 awake but resting in a clean plastic cage, and a 15-min "behaving"
 255 baseline, during which the rat was performing the two-alternative
 256 behavioral task. The order in which the baseline recording sessions
 257 were performed was pseudo-randomly varied from day to day.

258 Under normal conditions, NS neurons make up approximately
 259 5–10% of units in extracellular recordings. In order to investigate
 260 whether or not acute ketamine injections specifically alter the firing
 261 rates of NS1 cells, it was important to obtain data from as many
 262 NS cells as possible. Therefore, in this set of recordings, prior to the
 263 first baseline session each tetrode was manually lowered through
 264 OFC in 40 μm steps every 10–15 min until a cell eliciting narrow
 265 ($< 250 \mu\text{m}$) action potentials was identified by visual inspection of
 266 signal traces. If no such neuron was observed within 200 μm of
 267 the starting location, the tetrode was left in place until the follow-
 268 ing recording. As a result of this screening procedure, within the

269 ketamine data set, 40 out of 152 neurons met the criteria necessary
 270 for inclusion NS cells, a three- to four-fold increase in the fraction
 271 of NS cells relative to random sampling.

272 The baseline recordings were immediately followed by brief
 273 nasal exposure to paper towel dabbed with isoflurane, and the sub-
 274 cutaneous injection of either drug (30 mg/kg ketamine) or control
 275 (1% saline) solution. The administration of drug and control
 276 solution was alternated each day. After receiving an injection, the
 277 rat was immediately placed back into the behavior box and neural
 278 activity was continuously recorded for 90 min. Individual rats were
 279 used for an average of five recording sessions, receiving approxi-
 280 mately two to three total injections of ketamine and two to three
 281 total injections of saline before being sacrificed. No significant effect
 282 of drug treatment across days was observed.

283 HISTOLOGY

284 At the end of the experiment, animals were deeply anesthetized with
 285 sodium pentobarbital. Recording sites were marked in 12 rats by
 286 electrolytic lesions made for each electrode. Rats were then perfused
 287 transcardially with 4% paraformaldehyde. Brains were removed and
 288 stored in 4% paraformaldehyde until sections were made. Thirty
 289 micron sections were made, Nissl-stained, and photographed at
 290 1.25 \times to determine the location of recording sites. Recordings were
 291 obtained from ventrolateral and lateral OFC. We found no difference
 292 in basic properties across OFC including more lateral regions.

293 STATISTICAL METHODS

294 Unless otherwise mentioned, statistical analysis of data was per-
 295 formed using nonparametric (Wilcoxon) test for differences in
 296 mean values between populations with non-normal distribu-
 297 tions, two-tailed *t*-tests for normally distributed populations, or a
 298 χ^2 test for proportional differences across populations. All values
 299 are reported as mean \pm SEM.

300 RESULTS

301 Electrophysiological recordings were obtained from 1,044 neu-
 302 rons in the OFC of awake rats using chronically implanted tet-
 303 rodes (Wilson and McNaughton, 1993; Feierstein et al., 2006). An
 304 unsupervised clustering algorithm applied to the spike widths and
 305 firing rates of this cell population yielded two classes consisting of
 306 925 WS and 119 NS neurons (Figure 1) (Mountcastle et al., 1969;
 307 Baeg et al., 2001; Constantinidis and Goldman-Rakic, 2002). Data
 308 from previous studies indicate that NS cells are likely to correspond
 309 to interneurons (including both FS cells and non-FS cells), and
 310 that the WS population is predominately composed of pyramidal
 311 cells (McCormick et al., 1985; Kawaguchi and Kubota, 1997; Cauli
 312 et al., 2000). In this report, we focus on subgroups within the NS
 313 cell population.

314 DISTINCT SUBPOPULATIONS OF NARROW-SPIKING CELLS

315 NS neurons were subjected to a second round of hierarchical cluster-
 316 ing. For this analysis, we used three features of the extracellular spike
 317 train suggested by *in vitro* studies (Cauli et al., 2000; Hasenstaub
 318 et al., 2005): (1) rate of spike repolarization, (2) regularity of ISIs,
 319 and (3) degree of activity-dependent SB. In addition, we included a
 320 measure of spike shape – the ratio of pre-peak valley depth (V1) to
 321 post-peak valley depth (V2) (Figure 2A) – that has been used previ-

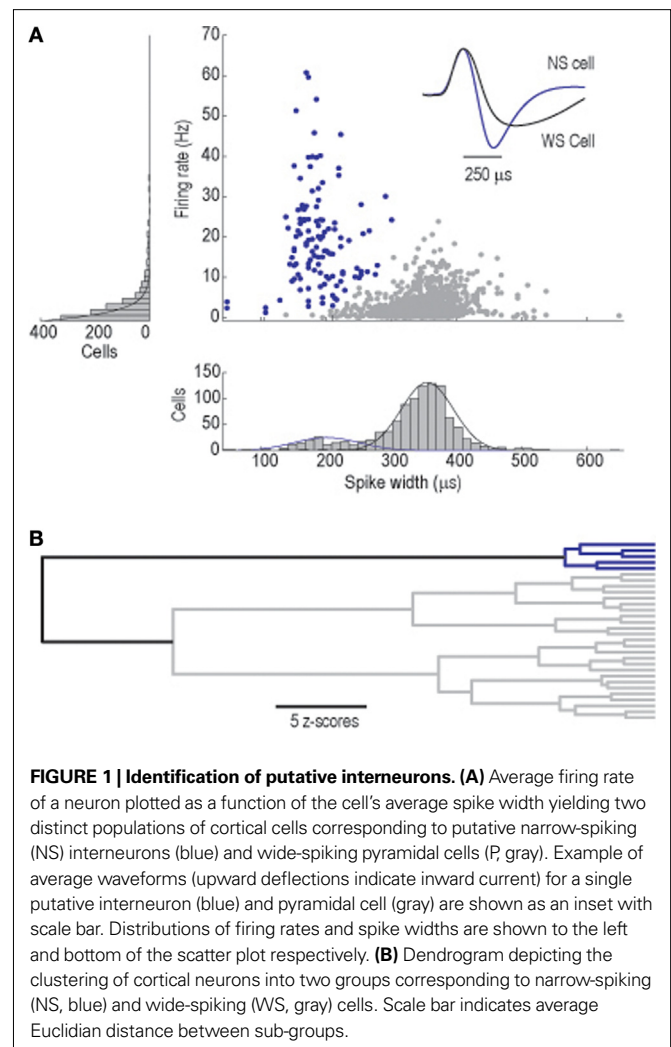
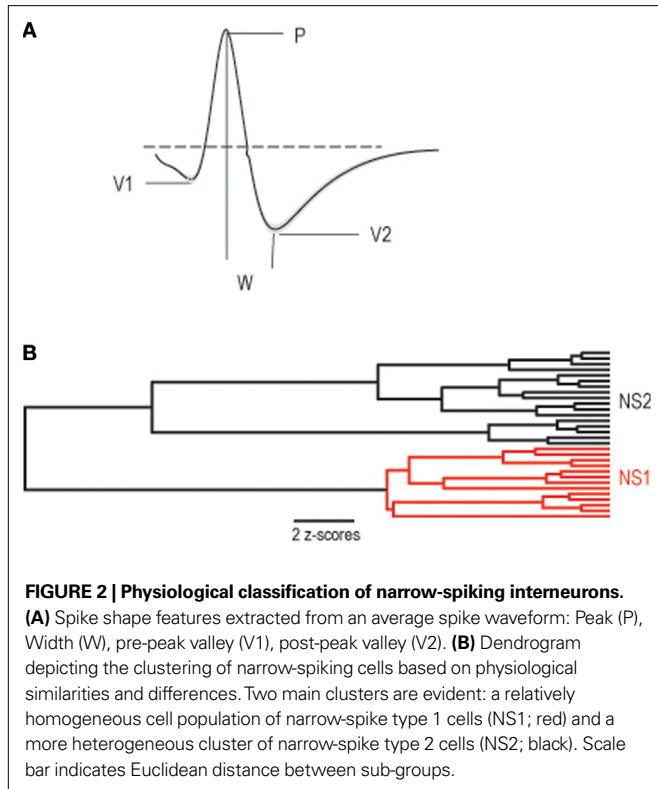


FIGURE 1 | Identification of putative interneurons. (A) Average firing rate of a neuron plotted as a function of the cell's average spike width yielding two distinct populations of cortical cells corresponding to putative narrow-spiking (NS) interneurons (blue) and wide-spiking pyramidal cells (P, gray). Example of average waveforms (upward deflections indicate inward current) for a single putative interneuron (blue) and pyramidal cell (gray) are shown as an inset with scale bar. Distributions of firing rates and spike widths are shown to the left and bottom of the scatter plot respectively. **(B)** Dendrogram depicting the clustering of cortical neurons into two groups corresponding to narrow-spiking (NS, blue) and wide-spiking (WS, gray) cells. Scale bar indicates average Euclidian distance between sub-groups.

ously to distinguish NS cells in the rodent hippocampus (Csicsvari 322
 et al., 1999). Unsupervised clustering based on these four features 323
 (see Experimental procedures) yielded two major divisions of 324
 NS neurons (Figure 2B). While further subdivisions are appar- 325
 ent within these two main divisions, because these two divisions 326
 roughly divided the NS population into two equal groups (see 327
 below) with numbers sufficient for subsequent statistical analysis, 328
 these two cell groups served as the main focus of this paper. 329

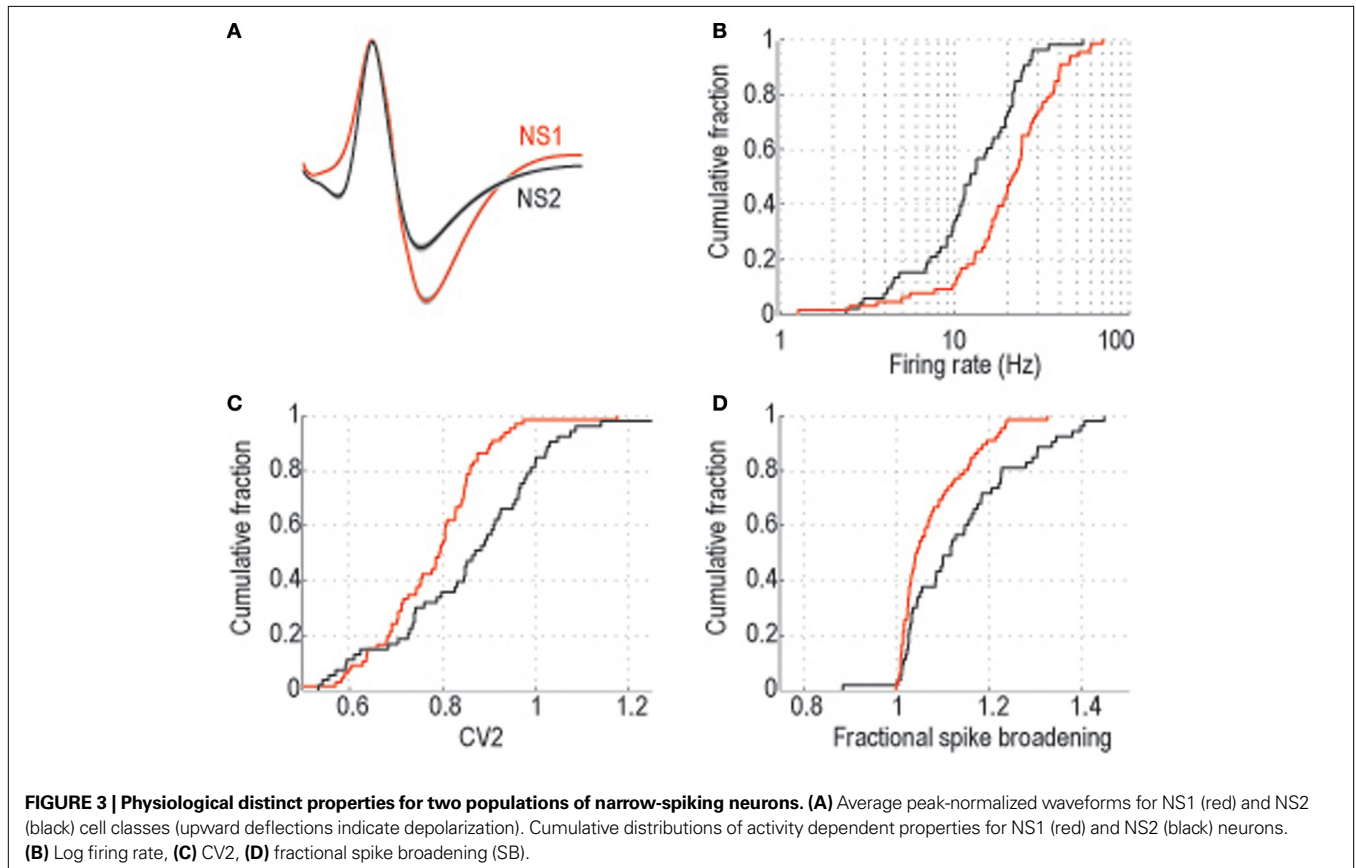
The first group of cells (NS1, $n = 66$) was a relatively homo- 330
 geneous cell class (average distance between cells: 1.80 *z*-scores) 331
 consisting of neurons whose average waveform was biphasic and 332
 exhibited a prominent post-peak valley, measured by the ratio of 333
 V2 to peak amplitude (P) ($V2:P = 1.00 \pm 0.025$) (Figure 3A). In 334
 contrast, the second cell population (NS2) was a more hetero- 335
 geneous group of 53 cells (average distance between cells: 2.48 336
 z -scores) characterized by a relatively shallow post-peak valley 337
 ($V2:P = 0.56 \pm 0.032$) and a prominent hyperpolarizing dip prior 338
 to the initial depolarizing phase of the cell's action potential 339
 (Figure 3A). In addition to these prominent waveform differ- 340
 ences, NS1 cells fired at higher rates than NS2 cells (24.1 ± 1.8 341
 vs. 14.9 ± 1.3 Hz; $P < 0.001$, Wilcoxon test; Figure 3B), with more 342



regular spike trains (0.78 ± 0.015 vs. 0.86 ± 0.024 ; $P < 0.005$, Wilcoxon test; **Figure 3C**) and significantly less activity dependent SB (1.08 ± 0.009 vs. 1.14 ± 0.018 ; $P < 0.005$, Wilcoxon test; **Figure 3D**). Given that there was no difference in the isolation quality of the two cell populations (ID, see Section “Materials and Methods”): 61.6 ± 5.5 vs. 63.0 ± 6.6 ; NS, t -test], the NS1 and NS2 classes are unlikely to reflect differences in recording quality. Furthermore, correlation analysis showed that spike shape features were highly correlated across spatially separated tetrode leads [V2:P: $r = 0.84$, $P < 0.01$; $\log(V1:V2)$: $r = 0.78$; $P < 0.01$, t -test], arguing that NS1 and NS2 features are not determined by distance from the recording electrode. Based on these cellular features, the NS1 class appears to be isomorphic with the FS class of GABAergic interneurons identified *in vitro*, whereas NS2 neurons appear to comprise a more heterogeneous population of cells (see Section “Discussion”).

ROBUSTNESS AND RELIABILITY OF NS SEPARATION CRITERIA

As documented above, the spike feature showing the greatest separation between NS1 and NS2 cells was V2:P. To explore the utility of using V2:P as a univariate metric for separating NS1 and NS2 cells, we performed a number of analyses. First, the work of Fee et al. (1996) clearly establishes the lack of stationarity and the large anisotropy associated with extracellular recordings. With reference to the current findings, it is therefore possible that non-stationarities in spike waveform shape could confound our



cell classification system. Thus, it is important to note that V2:P is a ratio measure. While absolute features such as amplitude, width, and valley can show a high degree of variability, these absolute features tend to covary such that the variability of the ratios are dramatically reduced. In our data set, V2:P values showed low variability ($CV = 0.2$) relative to spike width variability ($CV = 1.0$) across a recording session. Furthermore, across temporally separated epochs (pre- and post-saline injections) V2:P measurements were highly correlated ($R = 0.86$) and the average ratio of V2:P (pre-injection) vs. V2:P (post-injection) was not different from 1 (0.98 ± 0.035 ; NS, *t*-test) indicating no change in this measure across the two epochs.

We next evaluated the relationship between V2:P and other spike features within the unseparated NS population. **Table 1** shows the pair-wise correlation coefficients and significance values for all spike features used in this study. V2:P was most strongly correlated with V1:V2 and SB. Across all feature pairs the strongest pair-wise correlation was -0.6 and no single feature fully predicted any of the other extracted features with the NS population.

The observation that NS features were not completely predicted by any other singly observed feature supported the use of multi-dimensional hierarchical clustering in order to identify potential groupings within the NS population. Nevertheless, to establish the utility of single feature-based classification for subsequent *in vivo* studies, we evaluated the agreement of the classification based on clustering with a classification based on V2:P, the single waveform feature that showed the greatest separation between NS1 and NS2 cell populations (Mahalanobis distance 4.95 vs. 0.5799 for spike-broadening) and strong correlation with classification assignment ($r = -0.71$).

When the NS cell population was divided at a V2:P value of 0.80 the resulting two populations showed $>90\%$ overlap with the assignment obtained from our hierarchical classification procedure. Thus, V2:P values could be useful in identifying NS1 and NS2 cells under experimental conditions in which large samples of NS cells are unavailable. Furthermore, the relative ease of measuring V2:P and its relative invariance to changes in absolute spike amplitude suggests that this could be used to on-line screen cells during a recording session allowing for the targeted enrichment of specific cell populations for subsequent analysis.

Table 1 | Correlation between cell type and cell features.

	Type	Rate	CV2	Width	V2P	V1V2	SB
Type	–	-0.33^{**}	0.26^*	-0.16	-0.71^{**}	0.47^{**}	0.30^{**}
Rate		–	-0.52^{**}	-0.04	0.14	-0.23^*	0
CV2			–	0.14	0.03	-0.01	-0.06
Width				–	0.03	-0.32^{**}	-0.19
V2P					–	-0.6^{**}	-0.42^{**}
V1V2						–	0.43^{**}
SB							–

* $P < 0.02$, ** $P < 0.001$.

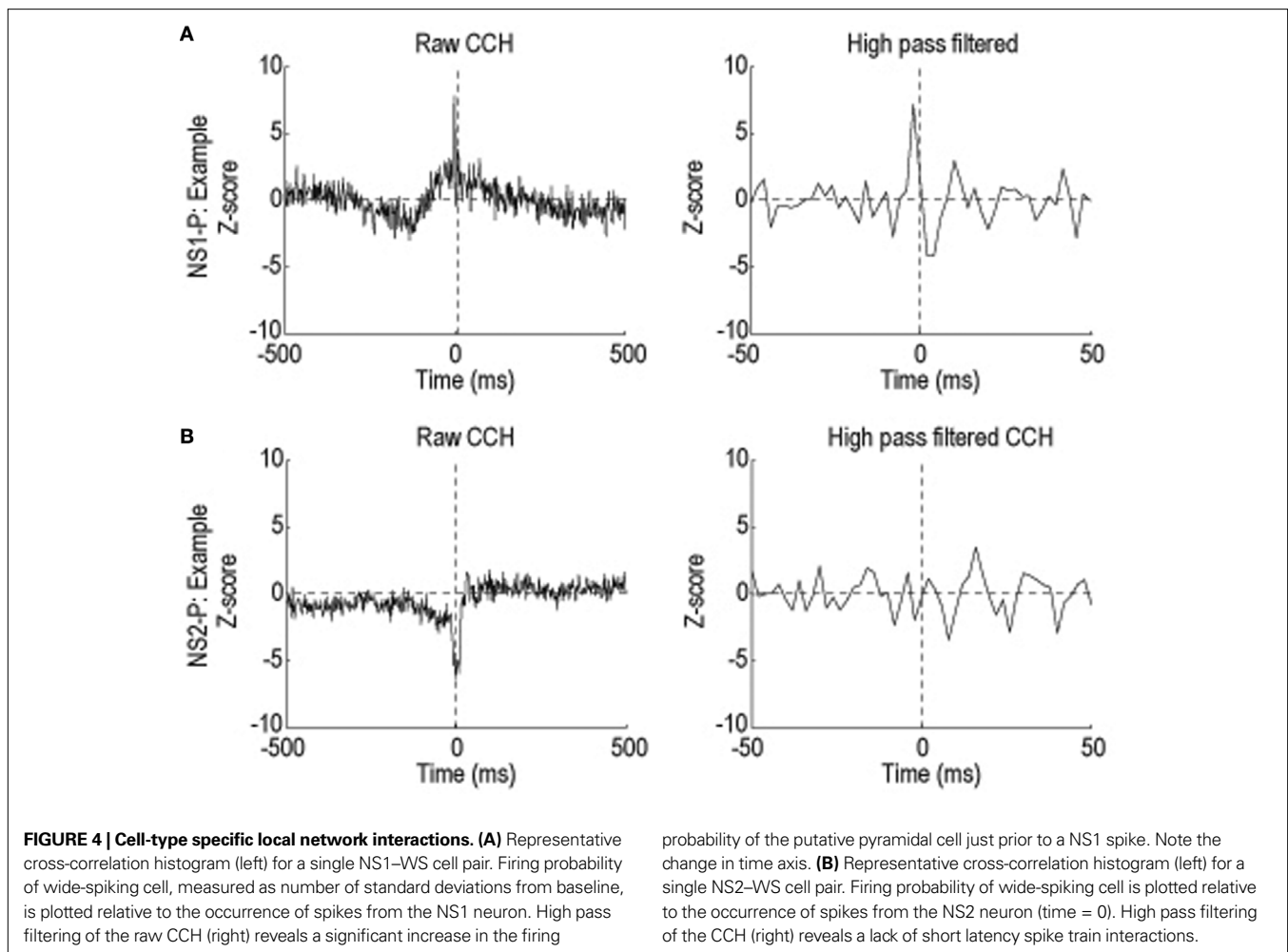
Pair-wise correlation coefficients and significance values for all cellular features used to separate NS1 and NS2 cells. Type refers to cell class (NS1 or NS2), as identified by hierarchical clustering.

NS1 NEURONS INTERACT STRONGLY WITH NEIGHBORING PYRAMIDAL CELLS

To provide independent validation of the distinction between the NS1 and NS2 populations, we examined the network properties of these two classes by constructing CCHs for pairs of simultaneously recorded neurons. **Figure 4A** (left), shows the CCH for a cell pair consisting of a NS1 cell and a WS neuron (NS1–WS pair). Two features of this CCH are notable. First, there is a substantial increase in the probability of the WS cell firing around the occurrence of a NS1 spike. Second, this increase in firing occurs over slow (10–100 ms) and fast (0–10 ms) time scales. Slow time scales of interaction are likely to reflect common input, while fast time scales can arise from direct synaptic interactions (Brody, 1999). To examine the fast time scale in more detail, we used high pass filtering to remove the low frequency component of the CCH (**Figure 4A**, right). The high frequency component of the NS1–WS CCH was dominated by an increase in the probability of WS cell firing just prior to the occurrence of an NS1 spike and a decrease in WS cell firing following the NS1 spike. This asymmetry, together with the short latency (2–4 ms) and narrowness (2–4 ms) of the correlation peak and trough, suggest a direct excitatory connection from the WS cell to the NS1 neuron and an inhibitory connection back onto the WS cell (Brody, 1999; Frank et al., 2001; Marshall et al., 2002; Bartho et al., 2004). Across the population of 116 local (same tetrode) NS1–WS cell pairs, 20% (24/116) showed similar evidence of an excitatory connection from WS neuron to NS1 cell. In contrast to NS1–WS cell pairs, NS2–WS pairs showed little evidence for direct excitatory coupling. **Figure 4B** shows a typical CCH for a single NS2–WS cell pair. Unlike the NS1–WS cell pair in **Figure 4A**, this NS2–WS cell pair had no obvious short-time scale interaction. Only 2 of 105 NS2–WS CCHs exhibited significant excitatory peaks, a proportion significantly smaller than that for NS1–WS cell pairs ($P < 0.01$, χ^2 test).

NETWORK DIFFERENCES WITHIN AND BETWEEN NS CELLS

In addition to differences in their interactions with WS cells, there were significant differences in interactions between and within the NS1 and NS2 cell groups. The average CCH for NS1–NS1 cell pairs ($n = 25$) showed a prominent zero-lag peak indicative of strong spike synchronization (**Figure 5A**), even though 24 of 25 pairs were recorded on different tetrodes that were separated by more than around 300 μm . In contrast, the average CCHs for both NS2–NS2 cell pairs ($n = 22$) and NS1–NS2 cell pairs ($n = 34$) were much flatter (**Figures 5B,C**). To ensure that the lack of a synchronous peak for NS2–NS2 and NS1–NS2 cell pairs was not a consequence of averaging a mixed population of positively and negatively correlated cells, we classified the CCH for each cell pair into one of four groups: no correlation, positive correlation, negative correlation, and positive/negative (mixed) correlation (**Figure 5D**; see Experimental procedures). Sixty percent (15/25) of NS1–NS1 cell pairs exhibited significant CCH peaks, while none of the NS1–NS1 pairs had significant troughs within the temporal window. NS2–NS2 and NS1–NS2 cell pairs exhibited fewer positive correlations and more negative and mixed CCHs. Across the population of NS neurons, the proportion of cell pairs within each of the four correlation categories was significantly different for the three groups (NS1–NS1 vs. NS2–NS2 vs. NS1–NS2, $P < 0.05$, χ^2 test; $df(6) = 13.28$).



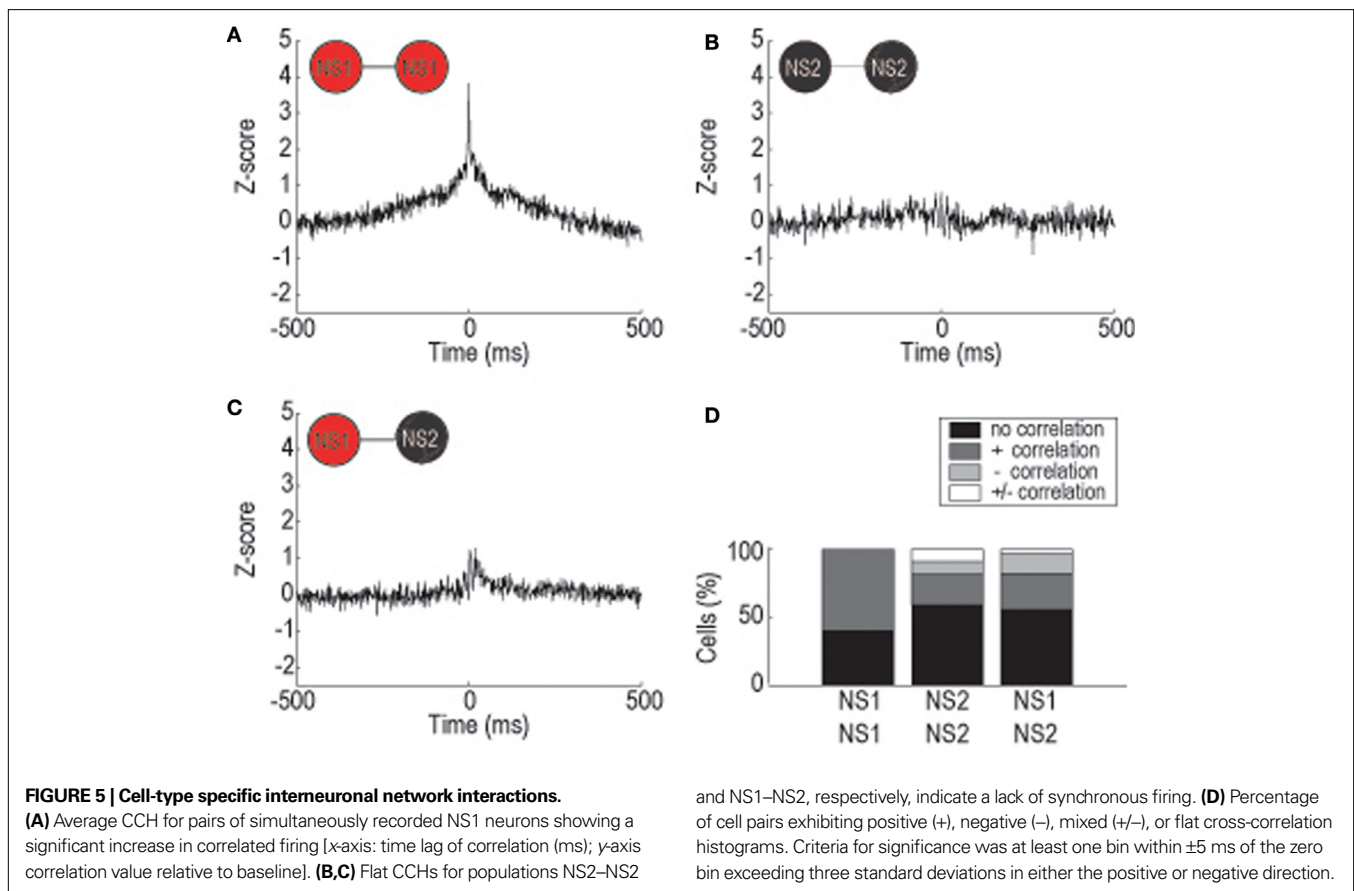
464 Thus, the distribution of network interactions within the NS
 465 population shows a strong dependence on the assigned identity
 466 of the individual units. This finding provides independent sup-
 467 port for the hypothesis that the NS1/NS2 classification reflects a
 468 true biological distinction. Moreover, the network features of the
 469 NS1 class are consistent with the synaptic properties of FS neurons
 470 studied *in vitro* (Gao and Goldman-Rakic, 2000).

471 CORRELATES OF NS1 ACTIVITY AND BEHAVIORAL TASK PERFORMANCE

472 The intrinsic and local network properties of NS1 cells together
 473 provide substantial support for the hypothesis that this group cor-
 474 responds predominantly to the FS class of GABAergic interneuron
 475 (basket and chandelier cells). To examine whether the relative physi-
 476 ological homogeneity of NS1 cells translates into similar activity
 477 patterns within functioning circuits during behavior, we examined
 478 the activity profiles of a subset of the NS1 neurons recorded from
 479 during an olfactory discrimination task (Uchida and Mainen, 2003)
 480 known to activate a large number of neurons in OFC (Feierstein
 481 et al., 2006). In this task, odor cues were associated with nose poke
 482 responses to one of two alternative choice or “goal” ports. This
 483 task allowed us to examine the relationship of the firing of NS1
 484 cells to sensory, motor and reward-related events during a simple
 485 goal-directed behavior.

486 If NS1 neurons form a coherent functional network relative to
 487 other cell populations within OFC, we would expect to find that they
 488 share common behavioral correlates. As a test of this prediction, we
 489 quantified the similarity between the activity profiles of individual
 490 NS1 neurons and the average activity profile of the NS1 popula-
 491 tion by computing the correlation coefficient between each cell’s
 492 activation profile (single cell post-stimulus time histogram PSTH
 493 time-locked to withdrawal from the odor port) and the average NS1
 494 PSTH. Across the population, the average correlation coefficient
 495 was 0.46 ± 0.059 indicating a strong degree of similarity between
 496 individual NS1 cells and the population average. By comparison,
 497 the same analysis showed much weaker homogeneity when applied
 498 to NS2 cells ($r = 0.25 \pm 0.042$). The average correlation coefficient
 499 between NS2 cells and the average NS1 PSTH was also signifi-
 500 cantly smaller than that of NS1 neurons ($r = 0.19 \pm 0.066$; $P < 0.05$,
 501 Wilcoxon test). Thus, NS1 neurons show relatively homogenous
 502 behavioral correlates compared to NS2 cells, a result consistent
 503 with their strong within-class interactions.

504 We next examined the nature of the behavioral correlates of NS1
 505 neurons. **Figures 6A,B** show the raster and PSTH of a typical NS1
 506 neuron time-locked to withdrawal from the odor port. The firing
 507 rate of this neuron increased during the odor stimulus period and
 508 during the period preceding reward delivery. In contrast, the neuron



509 exhibited a transient depression (relative to baseline) in firing rate that
 510 is time-locked to initiation of the movement from odor port to choice
 511 port. This pattern of activity was also seen at the population level
 512 in the average peak-normalized PSTH obtained from all NS1 neu-
 513 rons recorded during the task ($n = 41$; **Figure 6C**, red trace). Further
 514 examination of NS1 cell correlates indicated that these neurons were
 515 sensitive to trial outcome. First, during the period immediately follow-
 516 ing an animal's nose poke into the response port, the firing of NS1 cells
 517 was consistently higher on error trials relative to correct trials (28 ± 4.2
 518 vs. 26 ± 4.1 Hz; $P < 0.05$, Wilcoxon test). Second, during the time
 519 period just prior to odor delivery, the firing rate of NS1 neurons was
 520 significantly lower on trials following errors relative to trials following a
 521 correct response (28.6 ± 3.8 vs. 31.9 ± 4.3 ; $P < 0.05$, Wilcoxon test).
 522 In contrast to the relative behavioral response homogeneity of NS1
 523 cells, the average PSTH of NS2 neurons ($n = 41$) showed much weaker
 524 behavioral modulation across all epochs considered, and no signifi-
 525 cant dip time-locked to movement initiation (**Figure 6C**; black trace).
 526 Furthermore, unlike the NS1 population, the NS2 population showed
 527 no significant relationship to trial outcomes.

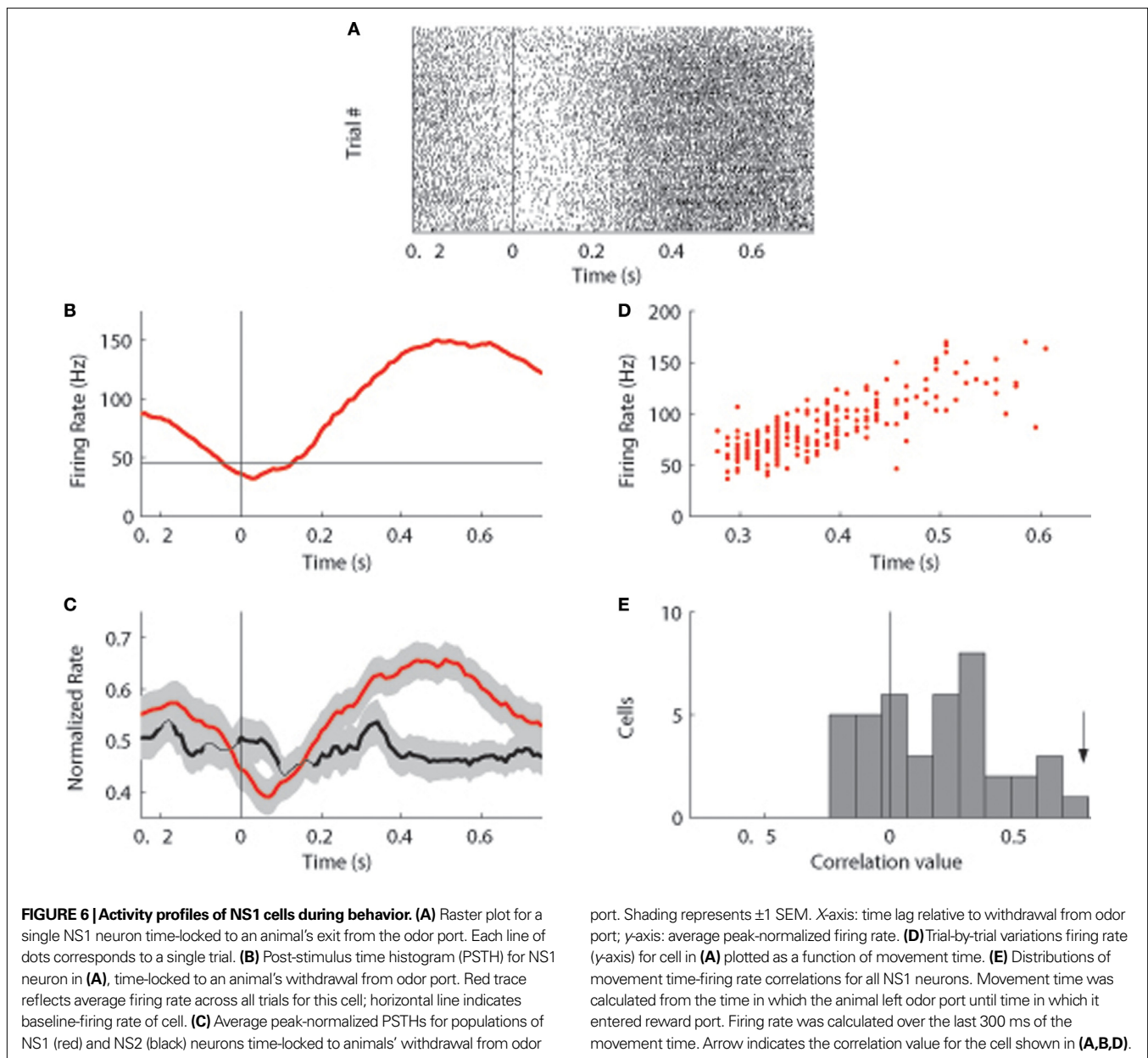
528 The dip in NS1 firing time-locked to the initiation of the
 529 behavioral response suggested a relationship between NS1 firing
 530 and the latency to respond, a commonly used behavioral metric
 531 of outcome expectancy (Schoenbaum et al., 2003). Shorter laten-
 532 cies are associated with better outcome expectancies and a more
 533 positive "motivational state" (Lauwereyns et al., 2002; Schoenbaum
 534 et al., 2003). To test whether OFC–NS1 cells show a specific link to
 535 such measures of motivational state, we examined the correlation

and NS1–NS2, respectively, indicate a lack of synchronous firing. **(D)** Percentage
 of cell pairs exhibiting positive (+), negative (–), mixed (+/–), or flat cross-correlation
 histograms. Criteria for significance was at least one bin within ± 5 ms of the zero
 bin exceeding three standard deviations in either the positive or negative direction.

536 between NS1 firing and the speed of movement. **Figure 6D** shows
 537 for a single NS1 cell (same cell as in **Figure 6A,B**) the relationship
 538 between trial-by-trial fluctuations in the cell's firing rate and the
 539 time taken for the animal to respond (i.e., move from the odor port
 540 to the goal port). For this cell, there was a strong positive correlation
 541 ($r = 0.79$; $P < 0.01$) between firing rate and response latency (or
 542 equivalently, a negative correlation between firing rate and speed of
 543 response). Thus, the strongest firing of the cell was associated with
 544 the longest latency to reach the goal port. Interestingly, similar cor-
 545 relations were found throughout the NS1 population. The average
 546 correlation value for all NS1 neurons was 0.19 ± 0.04 , with 18/20
 547 significantly correlated cells showing trial-by-trial fluctuations in
 548 firing rate positively correlated with movement time (**Figure 6E**).
 549 This property of OFC–NS1 neurons contrasts with OFC–WS cells,
 550 measured in the same task, which showed no net correlation (i.e., a
 551 symmetric distribution of correlations; Feierstein et al., 2006).
 552 Thus, OFC–NS1 firing, unlike the firing of the overall population
 553 of OFC neurons, showed an overall negative correlation with an
 554 indicator of motivational state.

NS1 NEURONS ARE SELECTIVELY INHIBITED BY SUB-ANESTHETIC DOSES OF KETAMINE

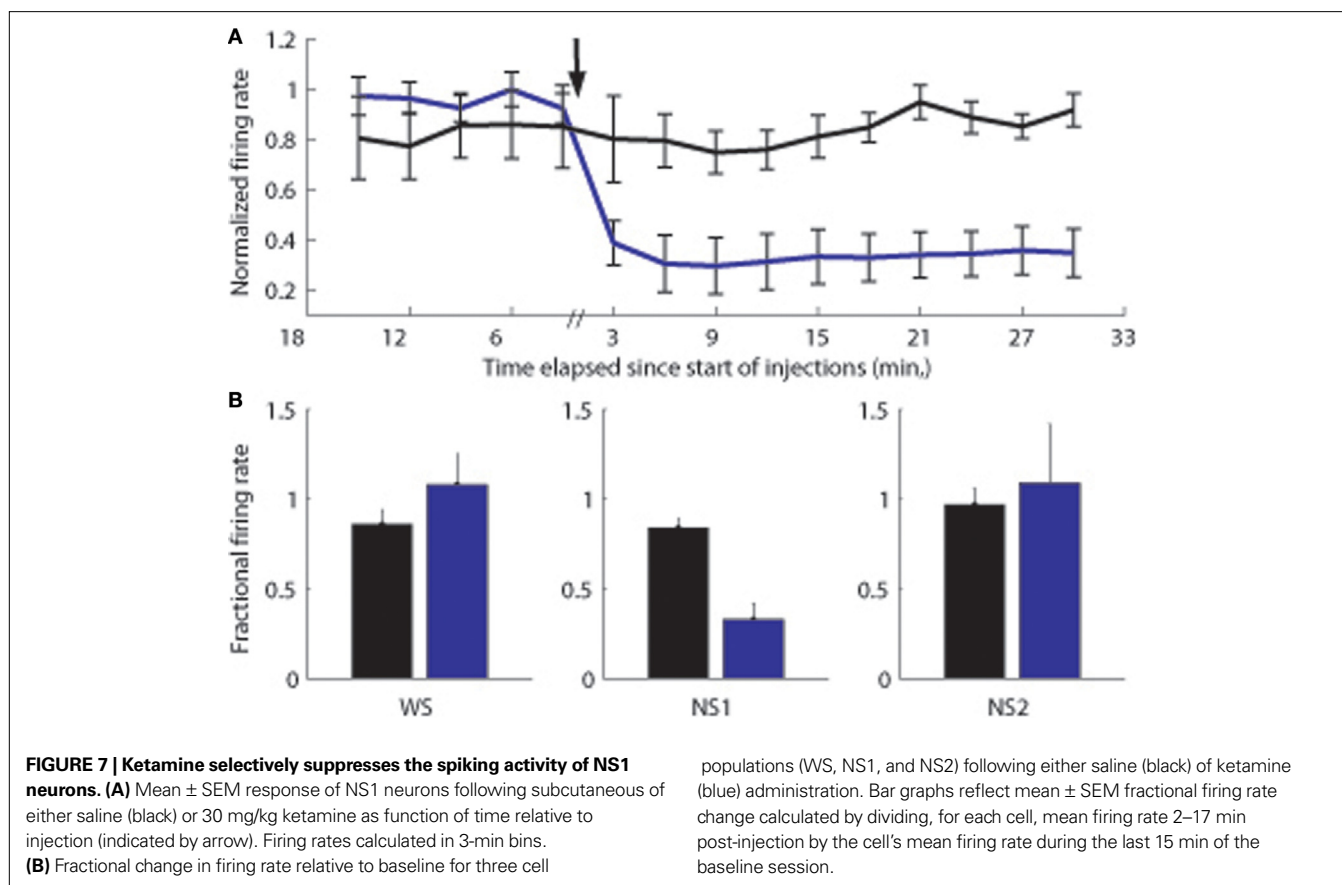
555 FS neurons have been implicated in the pathophysiology of psychosis,
 556 and are suggested to underlie the response of PFC to ketamine
 557 and PCP (Coyle, 2004; Lewis and Moghaddam, 2006; Homayoun
 558 and Moghaddam, 2007b). In particular, reduced FS function is
 559 theorized to result in dysfunction of PFC (Lewis and Moghaddam,
 560 561



2006; Homayoun and Moghaddam, 2007b). These hypotheses suggest that if NS1 neurons are indeed equivalent to FS cells, then ketamine or PCP administration should suppress their firing. To test this idea, in a subset of experiments, we recorded activity of OFC neurons during a drug-free baseline condition and following a subcutaneous injection of either 1% saline or ketamine (30 mg/kg). The 30 mg/kg dose was chosen based on earlier studies showing that 30 mg/kg elevated both glutamate and dopamine within PFC and produced behavioral impairments in a spatial working memory task (Moghaddam et al., 1997). Whereas gross behavioral performance following saline injection was indistinguishable from baseline conditions, ketamine sessions were characterized by complete disengagement from the task. Rather than performing the task, ketamine treated animals exhibited a number of stereotyped behaviors including both circling movements and repetitive limb

577 motions. These behavioral changes are consistent with the known
578 properties of dissociative anesthetics (Moghaddam et al., 1997),
579 and indicate that 30 mg/kg is a sufficient dose for producing acute
580 behavior effects without inducing an anesthetic state.

581 Systemic ketamine injection led to a dramatic decrease in the
582 firing rate of OFC-NS1 cells. **Figure 7A** shows the average response
583 profile of NS1 cells following either saline or ketamine injections.
584 Across the population of NS1 neurons, there was an immediate and
585 long lasting (>25 min) reduction in firing rate following ketamine
586 injection. Although there was some reduction in average firing rate
587 during saline injections as well, the firing rate of NS1 cells was
588 significantly suppressed relative to saline injections (average frac-
589 tion of baseline firing rate: 0.33 ± 0.09 vs. 0.84 ± 0.05 , $P < 0.001$;
590 **Figure 7B**). In contrast, the overall firing rates of both the WS
591 and NS2 populations following ketamine administration was not



592 significantly different from that following saline (WS: 1.08 ± 0.17 vs. 593 0.86 ± 0.08 , N.S.; NS2: 1.05 ± 0.37 vs. 0.97 ± 0.09 , N.S.; **Figure 7B**). 594 Interestingly, for these populations there was an increase in the vari- 595 ance of firing rate. That is, OFC–WS and OFC–NS2 cells showed 596 large increases and decreases in firing rate relative to both the pre- 597 injection baseline period and the saline control condition (see 598 Supplementary **Figure 1**). For the population of WS cells, this vari- 599 ance difference was significant [$F(1,112) = 3.9$; $P < 0.05$] suggesting 600 the existence of two populations of WS neurons within OFC – one 601 that is significantly suppressed by ketamine and another that is 602 enhanced. Neither spike amplitude nor spike width changed signi- 603 ficantly following ketamine or saline administration, indicating 604 that the treatments did not compromise recording quality. These 605 experiments directly implicate the NS1 cell class in the behavioral 606 response to ketamine, but also suggest that ketamine either acts 607 on other neuronal types as well, or that, as might be expected, WS 608 changes result from changes in NS1 inhibition.

609 DISCUSSION

610 Neurons in PFC exhibit a high degree of cellular and functional 611 heterogeneity. Here we used a classification procedure based on 612 features of extracellularly recorded spike trains to show that NS cells 613 in rodent OFC form physiologically distinct subpopulations, which 614 we term NS1 and NS2. The NS1 population appears to coincide 615 with the most common type of cortical interneuron, the so-called 616 FS cells, comprising basket and chandelier cells (Somogyi et al.,

1998; Markram et al., 2004). The fidelity of our classification is 617 supported by three independent observations. First, interactions 618 between simultaneously recorded units are dependent on the 619 inferred cell type. Second, the NS1 population exhibits relatively 620 homogenous behavioral correlates. Third, the NS1 neurons are 621 preferentially inhibited by the glutamate (NMDA) receptor antago- 622 nist and “psychotomimetic” ketamine. This final observation sup- 623 ports the idea that NS1 (FS cell) dysfunction contributes to the 624 behavioral impairments associated with acute psychosis (Lewis 625 et al., 2005), and provides a direct physiological link between the 626 glutamatergic and GABAergic theories of psychosis (Jackson et al., 627 2004; Lewis and Moghaddam, 2006).

628 Neocortical cells can be broadly divided into two categories 629 based on spike width and firing rate (Mountcastle et al., 1969; 630 Constantinidis and Goldman-Rakic, 2002; Bartho et al., 2004). 631 Although narrow spikes are generally thought to arise from 632 GABAergic interneurons (McCormick et al., 1985; Kawaguchi 633 and Kubota, 1997; Cauli et al., 2000), narrow spikes are neither a 634 required nor exclusive property of inhibitory cells (Nowak et al., 635 2003; Markram et al., 2004). Nevertheless, several observations 636 suggest that NS1 neurons are most likely isomorphic with that 637 of FS interneurons. First, in addition to narrow spikes and high 638 firing rates, NS1 cells, like FS neurons, exhibit fast and prominent 639 repolarizations (Cauli et al., 2000; Nowak et al., 2003; Hasenstaub 640 et al., 2005). Second, like FS neurons, NS1 cells exhibit relatively 641 little activity-dependent SB (Cauli et al., 2000). Finally, the high 642

643 degree of synchronized firing for pairs of NS1 neurons is consistent
644 with the observation that FS neurons are strongly coupled to each
645 other via both electrical and chemical connections (Galarreta and
646 Hestrin, 1999; Gibson et al., 1999) and with the fact that FS cells
647 receive similar synaptic inputs (Yoshimura and Callaway, 2005).
648 Although it is impossible with our current recordings to definitively
649 determine the mechanisms responsible for NS1 synchronization
650 *in vivo*, it should be noted that spike synchronization was observed
651 under both basal and task conditions. Further studies selectively
652 targeting mechanisms known to promote synchronization *in vitro*
653 (i.e., gap junction blockers) may help to clarify this issue.

654 In contrast to relative homogeneity of the NS1 population, the
655 remaining population of NS neurons, termed NS2, is likely to con-
656 tain a mixture of cell types including both excitatory “chattering”
657 cells and dendritically targeting interneurons. Chattering cells are
658 a subclass of pyramidal cells exhibiting narrow spikes, high firing
659 rates, and a tendency to produce bursts of high frequency action
660 potentials at gamma frequencies (Gray and McCormick, 1996).
661 While chattering cells have not been previously reported in rodent
662 cortex, a few of our NS2 neurons matched these criteria (<15%).
663 More generally, the physiological properties of NS2 neurons are
664 consistent with that of certain subclasses of dendritically targeting
665 interneurons previously reported to exist in both the hippocampus
666 (OLM cells) (Csicsvari et al., 1999) and in the PFC of anesthetized
667 rats (bitufted interneurons) (Tierney et al., 2004). Similar to our
668 NS2 population, extracellularly recorded OLM cells and bitufted
669 interneurons produced action potentials with relatively shallow
670 post-peak valleys, low firing rates and irregular firing patterns.
671 Furthermore, like NS2 cells, a subset of OLM cells and bitufted
672 interneurons produce tri-phasic spikes (Csicsvari et al., 1999;
673 Tierney et al., 2004). Given the high degree of heterogeneity with
674 the NS2 population, it will be important to determine the extent to
675 which other physiological criteria (i.e., bursting) can reveal mean-
676 ingful subdivisions within NS2 neurons similar to the distinction
677 between NS1 and NS2 neurons.

678 Functionally, NS1 neurons exhibit a number of features that
679 distinguish them from other cell populations within OFC. Most
680 prominently, we found that NS1–WS pairs account for 90% of the
681 short-latency interactions between putative pyramidal cells and
682 NS neurons despite the fact that they made up only about 50%
683 of the NS–WS cell pairs. The overrepresentation of these strong
684 interactions is even more striking given that the 116 within-tet-
685 rode NS1–WS cell pairs represents less than 2% of the total cell
686 pairs within our data set, yet account for more than 70% of the
687 putative monosynaptic interactions. The powerful connection
688 between WS cells and NS1 neurons is consistent with the proper-
689 ties of pyramidal cell inputs onto FS neurons known from *in vitro*
690 work (Gao and Goldman-Rakic, 2000) and suggests that NS1 cells
691 provide the majority of rapid feedback and lateral inhibition to
692 local OFC circuits.

693 In addition to receiving strong local drive, activity was strongly
694 synchronized between NS1 neurons at the 10–100 ms time scale,
695 as evidenced by their cross-correlations and similarity of activation
696 profiles during behavior. Consistent with the proposed role for
697 cortical interneurons in regulating the temporal flow of informa-
698 tion within cortex (Constantinidis et al., 2002), we found that NS1
699 neurons show enhanced firing while rats held their noses in the

700 sensory and reward ports, and reduced firing during movement
701 initiation. Importantly, NS1 firing was strongly correlated with
702 response latency, a measure of motivational state based on out-
703 come expectancy (Lauwereyns et al., 2002; Schoenbaum et al.,
704 2003). The further observation that NS1 neurons differentiate
705 between positive (reward) and negative (no-reward) outcomes,
706 both immediately and across trials can be interpreted in light of
707 recent data suggesting that primate FS cells play a role in atten-
708 tional gain modulation (Mitchell et al., 2007). Specifically, if we
709 assume that one function of NS1 FS cells is to regulate local network
710 excitability, one could posit that the elevation of NS1 cells firing
711 immediately following an error trial would shut down inappropri-
712 ate response networks, and that the subsequent reduction of NS1
713 activity during subsequent trials would be sufficient to unmask
714 normally suppressed response networks – a function consistent
715 with the role of OFC in error driven learning and behavioral flex-
716 ibility (Schoenbaum et al., 2003). Further behavioral studies of NS1
717 activity are needed to clarify these issues.

718 Together with the finding that OFC lesions impair normal dif-
719 ferences in response latencies associated with outcome expectancy
720 (Schoenbaum et al., 2003), our results support a central role for
721 NS1 neurons in mediating the expression of motivational state.
722 Because they collect inputs broadly from pyramidal neurons and
723 can in turn impose a broad influence, a network of NS1 neurons is
724 well positioned to integrate signals within the OFC and use them
725 to gate OFC output. Specifically, we hypothesize that NS1 cells play
726 a key role in selecting between “inhibitory” and “excitatory” influ-
727 ences of OFC on behavior (Figure 8). Increased activation of NS1
728 neurons, i.e., increased FS mediated inhibition within OFC, would
729 tend to favor “inhibitory” effects on behavior (slowed response
730 latency with lowered outcome expectancy) while decreased NS1
731 activation should favor “excitatory” output (speeded response
732 with heightened outcome expectancy). A specific prediction of
733 this hypothesis is that there should be two or more populations
734 of pyramidal neurons that receive opposite influences from NS1
735 neurons according to their output target and function. Similarly,
736 NS1 neurons could also regulate tradeoffs in a variety of functions
737 in which OFC output is implicated: exploration vs. exploitation in
738 foraging (Daw et al., 2006), high vs. low risk (Tobler et al., 2007),
739 and long vs. short time scale of reward discounting (Roesch et al.,
740 2006). In future experiments, the role of NS1 activity in mediating
741 the expression of motivational state can be tested using optogenetic
742 methods to selectively target and stimulate this specific population
743 of neurons during behavioral performance.

744 The functional properties of NS1 neurons take on heightened
745 relevance in light of the finding that ketamine strongly inhib-
746 ited this class of cells. Several prominent theories (e.g., Lewis and
747 Moghaddam, 2006) hypothesize that psychosis is associated with
748 a loss of “inhibitory control” within frontal cortex, resulting in
749 disorganized brain states. In addition, we observed a more vari-
750 able effect of ketamine on the activity of WS cells, which may be
751 due to the fact that there are a number of different kinds of WS
752 cells in the region of OFC from which we were recording, that
753 NMDA receptor subunit composition may differ between NS1
754 cells and WS cells, or that the local circuit organization in OFC
755 leads to the differential effect of ketamine we observed (e.g., WS
756 cells upstream of NS1 cells are the population of WS cells that are

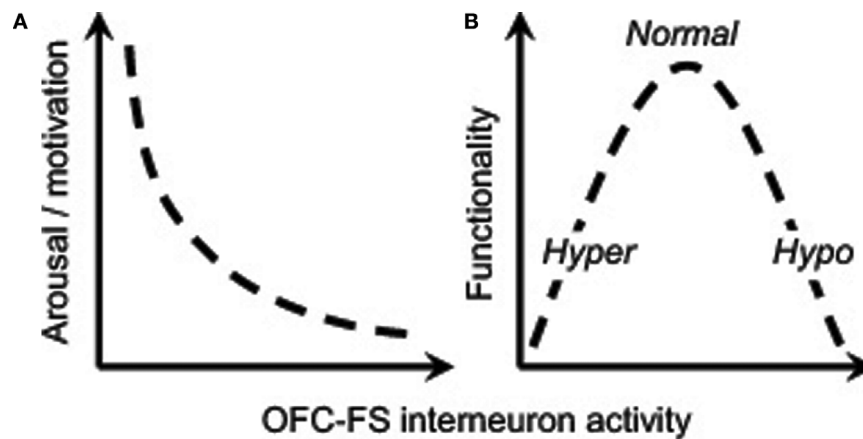


FIGURE 8 | Proposed role of OFC interneurons in regulation of behavior.

(A) The activity of OFC-FS type interneurons is negatively correlated with level of arousal or motivation, as indicated by correlations between NS1 firing and response latency. (B) An intermediate level of activity of OFC-FS interneurons is

expected to be associated with optimal behavioral functionality, with too low levels of interneuron leading to symptoms associated with hyperarousal (e.g., mania, positive symptoms) and too high levels leading to symptoms of hypoarousal (e.g. depression, negative symptoms).

757 inhibited by ketamine). The dramatic reduction in NS1 activity and
758 variable WS cell response we observed following ketamine ad-
759 ministration supports this class of theories. According to the above
760 proposal (Figure 8), a reduction of NS1 activity associated with
761 acute psychosis or mania would yield a reduced sensitivity to aver-
762 sive outcomes and a hyperexcitable motivational state (“positive
763 symptoms”). On the other hand, enhancement of NS1 activity
764 would be associated with a loss of motivational drive, a behavioral
765 phenotype associated with “negative symptoms” and depressed
766 brain states. The ability of ketamine to reduce the activity of a
767 hyper-excitable network of FS (NS1) cells could account for the
768 recently observed therapeutic benefit of acute ketamine treatment
769 on intractable depression (Berman et al., 2000; Zarate et al., 2006;

Maeng et al., 2007). While the complex relationship of brain func- 770
771 tion to psychiatric disease are undoubtedly a great deal more com-
772 plex than this simple account, this hypothesis may provide a useful
773 framework for further attempts to understand the link between
774 prefrontal neural circuits and psychiatric phenotypes.

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SUPPLEMENTARY MATERIAL

The Supplementary Material for this article can be found online 778
779 at [http://www.frontiersin.org/systemsneuroscience/paper/10.3389/](http://www.frontiersin.org/systemsneuroscience/paper/10.3389/neuro.06/013.2009/)
780 [http://www.frontiersin.org/systemsneuroscience/paper/10.3389/](http://www.frontiersin.org/systemsneuroscience/paper/10.3389/neuro.06/013.2009/)

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Conflict of Interest Statement:

Michael C. Quirk is currently employed at AstraZeneca Pharmaceuticals working on drug discovery programs in psychiatry. All work in present manuscript was conducted prior to joining AstraZeneca.

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