Estimating firing rates from calcium signals in locust projection neurons in vivo

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INTRODUCTION

Combining intracellular electrophysiology and multi-photon calcium imaging in vivo, we studied the relationship between calcium signals (sampled at 500–750 Hz) and spike output in principal neurons in the locust antennal lobe. Our goal was to determine whether the firing of individual neurons can be estimated in vivo with calcium imaging and, if so, to measure directly the accuracy and resolution of our estimates. Using the calcium indicator Oregon Green BAPTA-1, we describe a simple method to reconstruct firing rates from dendritic calcium signals with 80–90% accuracy and 50 ms temporal resolution.

Keywords: locust, olfaction, projection neuron, two-photon microscopy, calcium imaging, Oregon Green BAPTA-1, electrophysiology, spiking activity

MATERIALS AND METHODS

Animal preparation

Young adult female locusts (\textit{Schistocerca americana}) from our crowded colony were prepared as described previously (Laurent and Davidowitz, 1994). The brain was exposed and held on a small platform to prevent any movements during imaging. The head and brain were oriented such that the glomerular arborization plane of most PNs was horizontal. The brain was desheathed, the antennal muscles denervated, and the brain superfused with locust saline (Laurent and Davidowitz, 1994). Odors (citrus, 3-hexen-1-ol, benzaldehyde, isoamylacetate, 1-hexanol, cherry, citral, etc.) were applied to the antennal socket at a concentration of 10–100 μM. Each odor was presented for 10 seconds at 30 seconds intervals. The odors were presented in a random order and the trials were averaged over 2-10 trials depending on the odor.

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In vivo two-photon imaging of the calcium indicator OGB-1 in locust projection neurons (PNs). (A) Frontal view of a PN labeled from a secondary dendrite via the recording electrode (el.). Eleven of the PN’s 12 glomerular tufts can be seen, forming a circle in one plane. The soma and axon, respectively above and below the dendritic plane, are not included in this projection. (B) Side projection of a different PN, showing the soma (top left; typically on the antenna lobe’s frontal surface), the plane of glomerular projections (view orthogonal to that in A), and part of the axon (pointing down). Fluorescence intensity was kept roughly constant over the image depth by tuning the laser excitation intensity. (C) Simultaneous voltage (V_m) and OGB-1-imaging (ΔF/F, 500 Hz) recorded after dye injection into the soma (PN different from those in A-B): no significant changes in calcium signal could be induced by odor-evoked activity or current injection. (D) Dendritic impalement configuration typical for simultaneous calcium imaging and voltage recording (PN different from those in A-C): within 10 minutes of impalement, one or several tufts could be seen (box 1); a 30 × 30 μm² area was then selected for line scans (stippled line, box 2).

cineole, 3-pentanone) were delivered using a custom-made odor-delivery system and a Teflon nozzle (diameter 1 cm) placed 1–3 cm away from one antenna. Odors were delivered in a constant stream of air (0.4–0.8 L/second) at final concentrations of ca. 0.5 to 50%. The results reported are based on data obtained from 47 PNs in 40 animals.

Electrophysiology and dye loading
Thick (1.0 mm OD, 0.58 mm ID) and thin (1.0 mm, 0.78 ID)-walled borosilicate electrodes with filament (Harvard Apparatus, Holliston, MA) were pulled on a Sutter Instruments (Novato, CA) P2000 horizontal laser puller and filled with a solution containing 2–3 mM Oregon Green BAPTA-1 (OGB-1; Molecular Probes, Eugene, OR) in 2 M potassium chloride (DC resistance: 70–150 MΩ). PNs were impaled by targeting the Ω-shaped neurite (~2 μm diameter) from which the glomerular spokes (secondary dendrites) originate (see Figure 1). We also impaled PNs from their primary neurite (leading to the soma), from secondary dendrites and from glomerular branchlets. Spike waveforms and amplitudes did not differ significantly across these recording sites, suggesting that spike-mediating currents are quite homogeneously distributed throughout. Usable recordings lasted between 15 and 90 minutes. Local neurons were easily distinguished from PNs by their anatomy and absence of sodium action potentials (Stopfer and Laurent, 1999). Data were acquired via an Axoclamp-2B amplifier (Axon Instruments, Union City, CA), a National Instruments A-D card (15 kHz sampling) and LabView software (National Instruments, Austin, TX).

In vivo calcium imaging
A two-photon microscope based on a galvanometric-mirrors-scan-system (Denk et al., 1990) and proximal detection (Radiance, 2001; BioRad, Hercules, CA) coupled to an upright microscope (BX51-WF; Olympus, Japan) was used to image calcium signals from PNs. A mode-locked Ti: sapphire laser (Tsunami-10 W pump; Spectra Physics, Mountain View, CA) tuned to 850 nm was used as excitation light. A gradient contrast system (Yi et al., 2006) used in widefield reflection mode allowed precise positioning of the brain and electrode, mounted on XY-translation stages. Non-prechirped excitation pulses were focused onto labeled PNs using a 40 × 0.75 NA water immersion lens (Leica Microsystems, Germany). The epi-collected fluorescence was bandpass filtered (520 BP 40, Chroma Technology, Rockingham, VT) and detected with a photomultiplier tube (bi-alkali photo
our experimental conditions (locust PNs; OGB-1). Of those, the following steps for calcium-to-firing rate mapping: or sp/second). The algorithm smooths the calcium trace (Gaussian filter, trace represent changes in electrical activity (potential firing onset and interruption of firing, respectively), all portions of \( F(t) \) between a peak and the consecutive valley that last longer than \( 50 \) ms, \( \tau \), with a Gaussian decay function (\(\tau = 50 \) ms) and subsequently down sampled to 500 or 750 Hz. Missed (M) and false positive (FP) events were measured after detection of local maxima in the measured and estimated firing rate profiles (separated by \( t \geq 50 \) ms, Figure 6A).

Algorithm for firing rate prediction
Our algorithm depends on three parameters—\( F_0 \), \( \tau \), and \( S \)—derived from our experimental conditions (locust PNs; OGB-1). Of those, \( F_0 \) is the only experiment-specific calibration, used to measure baseline fluorescence at rest, in the absence of firing. \( F_0 \) is measured as the minimum value of \( F(t) \) during several seconds of baseline activity (typically, 6 seconds, preceding a stimulus pulse). \( \tau \) is the duration over which the effect of the summation of two successive spike-evoked calcium transients cannot be resolved from baseline noise. \( S \) is the scaling factor used to convert the normalized fluorescence (\( F_1 \approx F_0 \)) into a firing rate unit (spikes/second or sp/second). The algorithm smooths the calcium trace (Gaussian filter, 10 Hz), measures slopes, identifies peaks and valleys, and implements the following steps for calcium-to-firing rate mapping:

Step 1 is a rectification of \( F(t) \): because positive and negative slopes in the smoothed \( F(t) \) trace represent changes in electrical activity (potential firing onset and interruption of firing, respectively), all portions of \( F(t) \) between a peak and the consecutive valley that last longer than \( T_c \) are reset to \( F_0 \), with a Gaussian decay function (\( \sigma = 50 \) ms) from the peak. \( F(t) \) between a peak and a consecutive valley separated by \( t < T_c \) summate. This rectification brings most segments of negative slope to \( F_0 \).

Step 2 is a linear scaling of the rectified \( F(t) \) by \( S \) to yield the estimated \( R_p(t) \). To reduce the false assignment of action potentials to calcium signals triggered by subthreshold depolarization, we set a 4 sp/second threshold on the estimated \( R_p(t) \), that is, estimated rates below this value are automatically set to 0. The choice of 4 sp/second is justified in Figure 9.

The parameter \( T_c \) was determined empirically by generating the \( \tau \) decay time constant of the effect of a spike-evoked transient (\( \tau \approx 50 \) ms, Figure 5B).

Algorithm for firing rate prediction

Because measured odor-evoked firing rates did not exceed 40 sp/second, thus limiting dye saturation (Figure 6B). Normalized cross-correlation between predicted and measured firing rates, respectively \( R_p \) and \( R_m \), were calculated as \( R_p(t) \otimes R_m(t) / (R_p \otimes R_p(0) \cdot R_m \otimes R_m(0)) \), where \( \otimes \) is the linear correlation operator. Correlation coefficients were taken as absolute maxima of normalized cross-correlations (Figure 9C). FP and M activity peaks (separated by at least 50 ms) were counted from \( R_p(t) \) and quantified as percentages of the total number of real spiking events (determined from \( R_m(t) \)).

RESULTS

Locust projection neurons and experimental considerations
The locust antennal lobe contains about 830 PNs. The morphology of a typical locust PN is illustrated in Figures 1A and 1B. Somata are usually located near the anterior surface of the antennal lobe (Figure 1B) and connected by a thin primary dendrite to a planar, multi-glomerular arbor contained within a \(~\sim 50 \) \( \mu \)m-thick “slice” (front view: Figure 1A; side view: Figure 1B). The glomeruli are the sites where ORN axons contact PN dendritic tufts. Typically, each locust PN projects to 10–14 glomeruli (out of \(~\sim 1000 \) in toto); correspondingly, individual ORN axons \(~\sim 90 000 \) in toto project to several glomeruli (3–6). It is not known yet whether all ORN neurons of the same type converge to the same glomeruli, as they do in dipteras (Clyne et al., 1999; Halleen et al., 2004; Vosshall et al., 2000), nor whether the glomeruli visited by each PN are all innervated by ORNs of the same type, as observed with multi-glomerular mitral cells (MCs) in the mammalian accessory bulb (Del Punta et al., 2002; but see Wagner et al., 2006). In addition to receiving direct ORN input in the antennal lobe, PNs make direct, local excitatory connections with axon-less inhibitory neurons (Wilson et al., 2004; Jortner and Laurent, unpublished observations); because PN axons do not send collaterals in the antennal lobe, local PN output must originate from their glomerular dendrites. From this, we conclude that variations in intraglomerular calcium could reflect excitatory input from ORNs and spike-mediated PN output.

Spike output from PNs is routed downstream to the mushroom body where odor memorization and recall are thought to take place (Hammer and Menzel, 1998; Heisenberg, 1998), and to the lateral protocerebrum, via an axon (see initial segment in Figure 1B, pointing down). When injected into the soma with sharp or patch microelectrodes (\( n = 10 \) PNs), BAPTA-based calcium indicators (Oregon Green 1 and 2) never diffused more than 10–20 \( \mu \)m away from the soma (for reasons thus far unknown). Despite large somatic voltage signals and the known expression of calcium currents in the soma membrane of insect neurons (Goldberg et al., 1999; Laurent et al., 1993), no calcium signal variations were observed (Figure 1C). We next impaled PNs in their primary (Figures 1A and 1D) or finer dendrites with sharp microelectrodes backfilled with the high affinity indicator Oregon Green BAPTA-1 (OGB-1, \( K_d \approx 200 \) nM). Dendritic resting potentials were between \(~\sim 55 \) and \(~\sim 60 \) mV and dendritic spike amplitudes between 40 and 60 mV (\( n = 47 \) PNs). Small constant currents \( (\sim 200 \) pA) were often injected for iontophoresis; glomerular tufts could be seen within 3–10 minutes of impalement. Once a labeled glomerulus had been located (Figure 1D, inset 1), we performed line scans within a \( 30 \times 30 \) \( \mu \)m equatorial area at 500–750 Hz (Figure 1D, inset 2). Electrophysiological and imaging data were acquired simultaneously over 24 second-long trials, and repeated several times with one or several of the eight odors in our stimulation apparatus (see Materials and Methods).

Calcium signals in vivo are noisy and heavily modulated
Single-trials containing a period of spontaneous activity and a 1 second-long odor stimulus (shaded bar) are shown in Figure 2 for three different PNs. At rest, these PNs fired an average of 2–5 sp/second, typical of locust PNs in vivo (Mazor and Laurent, 2005; Perez-Olive et al., 2003), superimposed on a noisy background of synaptic activity. Upon odor stimulation, the neurons displayed complex, odor- and PN-specific firing characteristics (Laurent et al., 1998; Perez-Olive et al., 2002), with firing rates distributed between 0 and 60 sp/second, as typical in locust.

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Figure 2. Simultaneous glomerular-calcium imaging and dendritic recording in PNs in vivo. (A–C) Single 24 second-long trials (16 seconds shown), each with a different PN and odor stimulus (indicated by vertical gray band). Top: xt-scan across one glomerular tuft (30 μm × 16 seconds); middle: corresponding ∆F/F (500-Hz sampling); bottom: intracellular membrane potential (V_m; 15 kHz sampling). Stippled line: average baseline fluorescence before odor pulse. Note subthreshold components of PN voltage and corresponding ∆F/F deflections (asterisks), and non-exponential decay of ∆F/F after cessation of firing (triangles).
Periods of spiking alternated with epochs of hyperpolarization, which could in turn be interrupted by phasic subthreshold depolarizations (\(^*\), Figure 2B). The recordings were free of movement artifacts, as indicated by the spatial stability of the fluorescence signals on each x-t scan (Figure 2). The normalized fluorescence fluctuations (\(\Delta F/F\), see Materials and Methods) seemed reasonably well correlated with those of \(V_m\); they showed complex features, reflecting both the noisy and elevated baseline activity typical of in vivo conditions, and odor-induced firing modulation. Odor-evoked subthreshold depolarizing events or \(V_m\) modulation (\(^*\), Figure 2) could produce fluorescence increases comparable to those correlated with spiking. Individual spikes were accompanied by small calcium transients (\(\sim 2\%\)–\(10\%\)), which we will examine in detail later. Interruptions in firing were generally accompanied by slow decays of fluorescence to below baseline average (▼), Figure 2A–C; suggesting that a mapping between fluorescence and firing rates might require consideration of the neuron’s past history of activity. We first examine OGB-1 fluorescence at baseline and PN dynamic ranges.

**Dynamic ranges**

We recorded intracellular potential and OGB-1 fluorescence signal fluctuations from PN glomerular dendrites in vivo and compared results obtained at resting potential (Figure 3A) and at a moderately hyperpolarized holding potential (Figure 3B). We made the following observations:

1. Average baseline OGB-1 fluorescence at resting potential was somewhere in the middle of the indicator’s dynamic range; this was confirmed by observing the consequences of short current pulses (Figure 3A, \(I > 0\): trial 1; \(I < 0\): trial 2) on OGB-1 fluorescence. It follows that periods of decreased firing are accompanied by a decrease in OGB-1 fluorescence, and that periods of firing following a period of silence produce fluorescence transients that may not exceed the baseline mean; this can be seen for many of the action potentials produced in response to the cineole odor pulse (Figure 3A).

2. Only very high firing rates (obtained artificially by direct current injection; Figures 3A and 3B) led to likely saturation of the OGB-1 signal (70\% \(\Delta F/F\), relative to minimum, reached in response to negative current injection). From this, we conclude that OGB-1 is well suited to detect the variations of PN activity that occur within their normal operating range (0–20 sp/second, with rare transient discharges at rates not exceeding 60 sp/second).

3. Spontaneous and stimulus-evoked variations of fluorescence were often of similar magnitudes and time courses (for a PN held at its normal resting potential); care will thus be required to identify responses from these calcium signals.

4. Despite these complications, OGB-1 signals were reasonably consistent across trials (Figure 3A), and clearly correlated with the simultaneously recorded voltage: the variation of OGB-1 signals across trials clearly matched that of the PN’s firing profile over those trials. This suggests that OGB-1 signals may be sufficiently informative to reconstruct PN firing profiles in single trials.

5. While OGB-1 signals were high at baseline (i.e., at normal resting potential in vivo), much of the baseline fluorescence was reduced by simple hyperpolarizing current injection. This indicates that much of the baseline intradendritic calcium enters through voltage-gated channels, even if calcium channel activation is itself likely due primarily to (depolarizing) synaptic bombardment (see below).

From these observations, we conclude that the \(K_d\) and dynamic range of OGB-1 are well matched to the operating range of locust PNs in vivo, and thus potentially well suited to reconstruct spiking activity from calcium-dependent fluorescence signals.

**Caveats**

In several PNs, we sampled fluorescence signals from multiple glomerular dendrites. The PN in Figure 4A, for example, was impaled in a dendrite and examined successively in seven of its target glomeruli. This PN was held slightly hyperpolarized during the odor trials so as to minimize baseline firing. Figure 4A shows odor response segments from four of the seven glomeruli studied, together with the corresponding intracellular recordings. Successive odor trials evoked very similar voltage response patterns, typical of inter-trial variability with PNs (Stopfer and Laurent, 1999). The early odor-evoked spike burst, for example, contained \(\pm 4\)–\(\pm 1\) action potentials, at peak rates of \(32\%\) ± \(5\%\) sp/second. The amplitudes of the corresponding early OGB-1 transients were similarly clustered (37\% ± 5\% \(\Delta F/F\)). The decay phase of these transients, however, (\(r\), Figure 4A) depended neither on the number of action potentials (Figure 4B), nor on the firing rate (Figure 4C). Rather, the decay rate varied with the glomerulus from which the transients had been recorded (Figure 4A), ruling out decay phases of the OGB-1 signals as reliable correlates of PN firing rates.

We also noted that OGB-1 signals often rose even in the absence of action potential (Figure 4A, late response phase, blue trace). This is seen more clearly with the PN in Figure 4D: before, this PN was held slightly hyperpolarized to minimize baseline firing. Of the three successive depolarizing phases evoked by the odor, the first failed to evoke any action potential; yet, this depolarization was accompanied by an OGB-1 signal (1, Figure 4D) of amplitude comparable to the following two. These
Figure 4. Calcium signals are not always correlated with spiking. (A) Seven glomeruli were successively imaged in this PN, using constant odor delivery conditions. Single trials of $V_m$ and $F$ responses evoked by 1-second pulse of cherry odor are shown for four glomerular tufts (four colors). PN was held very slightly hyperpolarized. Note consistency of early $F$ and $V$ responses across glomeruli, to be contrasted with lack of correlation between $F$ (linearly quantified as a rate $r \pm SD$) in silent phase of the response and spiking response in preceding moments (in number of APs, B; or rate, C). (D) A different PN, held slightly hyperpolarized, in which $F$ is revealed to vary substantially even with subthreshold event (*).

Figure 5. PN glomerular fluorescence after blockade of nicotinic synaptic input with mecamylamine. Odor delivery failed to depolarize the PN and baseline activity was close to 0 sp/second. (A) Current-evoked spike trains lead to clear summation of $F$ transients, but current-evoked hyperpolarization leads to no further decrease in $F$, indicating that baseline PN fluorescence in intact animals is elevated because of tonic synaptic depolarization. (B) Spike-evoked calcium transients (single event: thin line; spike-triggered average (STA) of seven events: thick line) with decay time constant ($\tau = 50 \pm 7$ ms). (C) Upon large current injection, causing fast PN depolarization, $F$ increases prior to the first action potential, indicating low-voltage-gated calcium entry, consistent with high baseline fluorescence in intact animals.
examples suggest that the contribution of subthreshold depolarization may, under certain circumstances, be a potential source of error for the evaluation of PN spike output from OGB-1 signals. Predicting PN output from OGB-1 signals will thus require a separation of these superimposed spike-dependent (indicative of PN output) and subthreshold components (indicative of synaptic drive to PNs).

Separating voltage-gated from synaptic contributions
Insect olfactory afferents are cholinergic (Bicker, 1999). Superfusion of the brain with 1 mM mecamylamine, a nicotinic receptor antagonist, led to drastic changes in PN activity at rest and during odor stimulation (n = 3 PNs). First, PN resting potential became more negative by 5–15 mV and baseline firing was suppressed, with the exception of occasional spontaneous bursts of spikes. Second, odor stimulation failed to evoke any depolarization. The effects of mecamylamine on electrical properties were reflected in the calcium signals (Figure 5). Mean OGB-1 signal intensity was drastically reduced in theglomeruli. Action potential trains, evoked by a small depolarizing current-pulse, elicited clear and summat ing calcium-transients (Figure 5A); each transient decayed considerably faster (τ ~ 50 ms, Figure 5B) than fluorescence signals correlated with inhibition or hyperpolarization from baseline in normal saline (Figure 3A). Third, hyperpolarizing current injection caused no further reduction in fluorescence (Figure 5A); this suggests that part of the elevated baseline calcium is due to non-inactivating voltage-gated channels, and that blockade of nicotinic synapses results in a net PN hyperpolarization below the activation threshold for voltage-gated calcium entry. Indeed, large depolarizing current injection triggered OGB-1 signals from below spike threshold (Figure 6C), at about ~60 mV.

We conclude that two spikes with a time interval ×50 ms are very easily detectable as OGB-1 fluorescence transients when PNs are held at moderately hyperpolarized potentials (~60 mV). In in vivo recording conditions, however, synaptic background due to ORN activation raises the resting potential of PNs above calcium-channel activation threshold; at rest, mean OGB-1 fluorescence is around the midpoint of the indicator’s dynamic range; from this baseline, single spike-related transients are thus compressed, and occur on a consistent synaptic background, due both to sub- and supra-threshold depolarizing events. With this information, our goal is now to estimate PN spike output from OGB-1 fluorescence signals.

Estimating firing rates from ∆RF(t)
Our qualitative results indicate that ∆RF(t) variations in PNs are influenced by at least three different factors: firing (Figures 3 and 5), subthreshold depolarization (Figures 4 and 5), and calcium clearance (Figure 4A). Having access to simultaneously recorded voltage and fluorescence, we now investigate whether a PN’s discharge can be extracted with reasonable precision from single trials using ∆RF(t) recorded at a glomerulus. Building on previous experimental and modeling work with other neurons and systems (Heimchen et al., 1996; Wang, 1998), we assume that sustained firing rate can, over a reasonable range, be linearly related to intracellular calcium levels. We will now describe a simple method to derive a smoothed estimate of PN firing rate R_{sm}(t) from ∆RF(t). To this end, we use simple rules based on two PN-specific parameters (S and Tₛ) and a single experiment-specific calibration, independent of the electrophysiological recording. Both parameters—S, a scaling factor related to average calcium entry caused by one action potential at the imaging site, and Tₛ, the duration over which spike-evoked calcium summation appears not to be relevant—are biologically meaningful and were obtained from our experimental measurements.

Method summary. We first estimated, from a recorded PN, the minimum fluorescence at baseline (F₀, measured in the absence of spike output). Because PNs recorded at rest and in vivo fire 2–5 sp/second on average (Perez-Orive et al., 2002), the minimum fluorescence measured during several seconds of baseline generally provides a reasonable estimate of F₀. F₀ can thus be estimated in the absence of electrical recording. Note that F₀ does not represent the lowest possible OGB-1 signal, but rather a rest position for the spike-independent components. After appropriate smoothing (100 ms kernel), we consider that ∆RF(>F₀) varies approximately linearly with Rₛ (always >0). Because a cessation of firing, due either to PN inhibition or to interruption of excitation, generally led to a decay of ∆RF(t), we assume that any segment of ∆RF(t) with a slope <0 and a duration >Tₑ represents a firing rate of 0. Our method thus sets all such segments of continuously decreasing ∆RF(t) to F₀. ∆RF, so rectified, was then scaled linearly to Rₛ using the constant S. To reduce the false assignment of calcium variations due to subthreshold events to action potentials, we set a threshold on the estimated Rₛ at a value roughly equal to the spontaneous neuron spiking rate (see exploration of parameters).

Demonstration. A demonstration of this procedure over a single trial (response segment from PN in Figure 3, called PN4) is shown in Figure 6A; using only ∆RF(t) (red trace), we computed—as described above—a prediction (green trace) of the PN’s smoothed firing rate and compared it to the firing rate profile calculated from the intracellular recording (blue trace). Note that ∆RF(t) (red trace), either raw or simply thresholded at F₀, did not predict firing output well. The scaling parameter S—kept constant once estimated—was the slope of the linear fit of X (measured firing rate) versus Y (unscaled predicted firing rate) (Figure 6B), where X and Y were measured as in Figure 6A, for a set of 5 PN-odor pairs. S, so calculated, was 1.2 (sp/second)%/ (Figure 6B). A regression plot of X and Y after scaling is shown for PN4 in Figure 6C.

We measured the accuracy of this method in two ways: first, we cross-correlated the predicted and measured firing rate profiles (after smoothing, Figure 6D); in this example, our average estimate of PN4 firing rate from OGB-1 fluorescence was ~80% of the actual one. Second, we counted M or FP events in the prediction. In the trace in Figure 6A, for example, we illustrated one M and one FP event in our prediction. The proportions (in %) of such deviations relative to all real events are plotted in Figure 6E.

Fine tuning. Figure 7 illustrates the reliability of our procedure with other PNs and odors. Again, the only experiment-specific parameter is F₀, chosen as the minimum of ∆RF during baseline (i.e., before the odor pulse). Raw/smoothed ∆RF(t) (red), predicted (green), and measured (blue) firing rate profiles are plotted, together with cross-correlation and error rates for 3 PN-odor pairs (PNs 5–7, single-trials). Correlations at τ = 0 were close to 0.9 and errors were balanced overall; except for a few M and FP spiking events, the time structure of the predicted running firing rates matched the real ones almost perfectly. We showed earlier, however, that subthreshold depolarizations occasionally generate OGB-1 signals comparable to those evoked by action potentials (Figure 4D). Because the calcium signals corresponding to spiking and non-spiking events cannot be teased apart easily, we explored the use of a simple threshold (>F₀), below which all corresponding calcium events would be counted as subthreshold (0 sp/second). Because PN-mean firing rates at baseline vary between 2 and 5 sp/second across PNs, any threshold value within that range would generate most errors during baseline; we considered this acceptable because our goal is to reconstruct PN response profiles, not baseline activity, with maximum accuracy. Figure 8 shows the effect of varying this threshold on firing rate estimates (green) with two PNs selected for illustration because they respectively represent extremes of subthreshold electrical activity and spiking activity at rest. Our goal was to choose a single threshold value so as to minimize error rates (low M and FP) and maximize reconstruction accuracy (high correlations), over all PNs. The effect of threshold on error rates is plotted for both PNs; while a high threshold of 8 sp/second was optimal for PN8, a threshold of 2 sp/second was best for PN9. A threshold of 4 sp/second was found to provide the best average predictions over all PNs.

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Figure 6. Prediction of one PN's firing profile from its glomerular calcium signals alone. (A) Using the rules described in the text, we use the Gaussian-smoothed $\Delta F/F_0$ trace (dark red) to estimate the PN's discharge with ~50 ms resolution. Stippled line indicates $F_0$ (note that $F_0 \neq F_0$). Compare predicted, smoothed firing rate (green trace) with that calculated by 50-ms-Gaussian smoothing (blue trace) of actual spike discharge obtained from intracellular trace ($V_m$). X, Y: see B; M, FP: see E. (B) The proportionality constant $S$ between firing rate and calcium-plateau, used as a fixed scaling parameter in our method (see text), was extracted from 5 PNs (16-second-long single trials, 1-second odor presentation), each exhibiting moderate firing rate modulations (below 40 sp/second) upon stimulation. X and Y mark the spike-related peak amplitudes of, respectively, measured and predicted-smoothed firing rates (as illustrated in A). (C) Dispersion from assumed linearity of $Y(X)$, shown for PN4. (D) Cross-correlation (normalized to autocorrelation of measured firing rate) between estimated and measured firing rates; this measure is used as a global accuracy estimator of our predictions: PN4 prediction captures almost 90% of the actual firing rate. A perfect prediction would yield a peak at 1.0. (E) Match between predicted and measured firing rate peaks was also assessed piecewise; percentage of missed (M) and false positive (FP) events was measured (see examples in A). FP events are less than 5% of the total number of real events (n).
Estimating firing rates

Figure 7. Firing rate prediction from $\Delta F/F$ as applied to three different PNs. (A–C) Left: raw and smoothed (red and dark red) fluorescence traces ($\Delta F/F_0$), predicted (smoothed) firing rate (green), measured (smoothed) firing rate (blue). Right: assessment of match between estimated and measure rates; top: normalized cross-correlation; bottom: proportions of missed and false positive events. In each example, 15 seconds of 24-second-long, single trials are shown. The only experiment-specific parameter is $F_0$ (dashed lines), measured as minimum calcium signal before the odor presentation; $F_0$ thus requires no manipulation of membrane potential, and is not the same as $F_0$. All other parameters ($S$, $T_c$, and the threshold; see text and next figure) are fixed and identical for all. Note high accuracy for predicted firing for PN7 and small false positive percentage for all three examples.

Finally, we examine the relevance of PN discharge statistics to the accuracy of our procedure. The transform we propose relies on a few approximations (smoothing, mean slope, and threshold), best adapted to particular firing regimes. We thus examined the reliability of our firing rate estimation with many PNs and multiple stimuli—each evoking a different response pattern (Figure 9). The example in Figure 9A illustrates the responses of one PN to six different odors (single trials); the OGB-1 signal is plotted in red, and the predicted and measured firing rates are shown in green and blue, respectively. The accuracy of these estimates varies between 80 and 90% (Figure 9B). Measured over 9 PNs and 21 PN-odor pairs, the mean correlation was 81% (Figure 9C).

DISCUSSION

Somatic calcium signals combined with simultaneous cell-attached recordings have been used more recently in vivo in anesthetized ani-
Different thresholds are tested (green traces) and compared with the firing rate profile (blue) measured. The PNs in A and B are chosen because they illustrate two extremes of PN activity patterns (A: sparse firing, large subthreshold potential variations; B: high and variable resting activity). Because the threshold is a fixed parameter, it should be optimized such that reconstruction accuracy is maximized over all PNs. (A) Following odor presentation, PN8 undergoes many successive subthreshold potential variations, leading to a large proportion of false positive events in the prediction from ∆F/F. A threshold of 8 sp/second would be optimal for this neuron. (B) By contrast, PN9 displays a high and irregular spike discharge both at rest and after odor stimulation, causing very few false positive events. A high threshold, however, leads to many missed events. A threshold of 2 sp/second would be optimal for this neuron. After similar analysis with several PNs, the threshold parameter was set at 4 sp/second; this value represents a trade-off between missing and erroneously detecting spiking events, when assessed over a large span of PN firing patterns.

mals to identify temporally sparse (∼1 Hz) spontaneous activity of neuron ensembles (Kerr et al., 2005). Because activity was restricted to single action potentials or short bursts, a simple template-matching method applied to the calcium recordings successfully extracted spike times with good accuracy; a simple linear relationship between calcium elevation and action potential number (with a maximum at about 4) for moderate interspike intervals was experimentally established and used to extract spike time series during sparse bursting events.

More recently, Yaksi and Friedrich (2006) developed a method to extract spiking activity from soma calcium signals, using a more complex deconvolution approach. This method, used on populations of adult zebrafish MCs stimulated with odors, is based on the assumptions that (1) spike-evoked calcium transients are the dominant source of calcium elevation, (2) each action potential gives rise to a stereotyped calcium transient with mono-exponential decay over time, and (3) calcium summation caused by a train of action potentials is described well by a simple convolution of a mono-exponential kernel with the spike-train. Calcium signals were recorded at MC somata with Rhod-2, a moderate affinity indicator (K_d ∼ 570 nM); calcium kinetics appeared to be relatively slow (single-action potential decay time >1 second), and calcium variations appeared to be related only to action potentials. When recorded spike trains and reconstructions were compared, Yaksi and Friedrich’s approach gave very good firing rate reconstructions (∼85%), with image sampling rates of ∼30 Hz. These reconstructions were done while MC baseline activity (typically 5–10 Hz) had been suppressed by hyperpolarizing current injection; the thresholding problem caused by a non-zero baseline activity is thus similar to the one we faced (see below).

In the present study, we examined the predictability of PN-spike output from calcium signals recorded at high sampling rate (line-scan, 500–750 Hz) from glomerular dendrites with a high-affinity indicator (OGB-1, K_d ∼ 200 nM). In each glomerulus, individual PNs receive direct cholinergic input from ORNs (Homberg et al., 1995) and make cholinergic output to LNs (Drosophila: Ng et al., 2002; Wilson et al., 2004). We found that PN-dendrite OGB-1 signals are fast (single-spike decay time ∼ 50 ms) after blockade of nicotinic synaptic input, but slower and more variable (e.g., non-stereotyped decay times across multiple glomeruli of a single PN) in intact in vivo conditions. We also determined that OGB-1 signals are not exclusively correlated with action potentials, a likely consequence of...
Figure 9. Reliability of firing-rate prediction over various stimuli and response patterns. (A) Comparison of predicted (green) and measured (cyan) smoothed firing rates. All plots are from single-trials, with the same PN (PN10), and six odors (1: cis-3-hexen-1-ol, 50%; 2:1-hexanol, 50%; 3: benzaldehyde, 50%; 4: benzaldehyde, 0.5%; 5: isoamylacetate, 50%; 6: isoamylacetate, 0.5%). Corresponding raw fluorescence traces (red) are juxtaposed, to emphasize the difference between calcium signals and spike discharge. (B) Normalized cross-correlations between predicted and measured (50 ms-Gaussian smoothing) rates for the six trials in A. Peak correlations are between 0.80 and 0.90. (C) Measure of peak cross-correlation (at dt = 0, as in B) for 21 PN-odor pairs calculated with 9 PNs. These correlations are calculated over the entire duration of each trial, and thus represent both stimulus-evoked and spontaneous activity, as shown in A. In all cases, the only trial-specific parameter is FB (minimum fluorescence during pre-stimulus baseline); all other parameters are fixed and identical (see text).
the indicator’s affinity. We developed simple rules to predict firing patterns from imaging data with ~80–90% mean accuracy. The errors made in our prediction originated from two main causes: (i) an imprecise evaluation of the fluorescence levels (F0) that correspond to 0 action potential; this evaluation might be improved by recording longer periods of baseline activity; (ii) more importantly, a deviation from the assumed linearity between ΔF/F and firing rates, due to nonlinearity summarization of spike-related events (mostly at high rates) or to the superposition of subthreshold and spike-dependent components. We suspect that the combination of these two components, presumably non-linear, is mainly responsible for the variance of the scaling factor S (Figure 7B). A calcium indicator with a lower affinity than OGB-1 but similar unbinding rate would likely help diminish the weight of the subthreshold components.

We now compare our approach to the deconvolution method proposed by Yaksi and Friedrich (2006), which is, to our knowledge, the most relevant work in the interpretation of calcium imaging data to extract firing activity.

Computation time
Our algorithm is, by contrast to deconvolution, computationally trivial; it can thus be applied at low cost to long calcium recordings regardless of sampling rate. For a single 24-second trial sampled at 500 Hz, the calculation time on MatLab 7.0 running with Pentium(R) IV1.50 GHz processor is less than 1 second, data-loading time included.

Noise
Because, in practice, indicator concentrations must be kept as low as possible to minimize their influence on neuronal function, S/N ratios are usually low, especially so if averaging is excluded (single trials, high temporal resolution). Deconvolution, like all inverse problems, is very sensitive to noise; Yaksi and Friedrich (2006) thus applied a clever but time-intensive local lowpass filtering on their 30 Hz-calcium traces. Our algorithm performed well with a simple three-point median filter on 500 Hz-line scans.

Calibrations
Our approach necessitated simultaneous voltage and calcium recordings from a few neurons to establish the three parameters needed for our prediction algorithm. These parameters are likely to vary with neuron types and preparations. In many cases, such simultaneous calcium and electrical recordings may be difficult to obtain in vivo. On the other hand, the calibration procedure could be done in intact, in vivo conditions, thus providing information about the transform’s parameters for neurons in their natural state, in the absence of any artificial manipulation.

Assumptions
We made no assumption on the waveform and time constant of calcium signal decays (we rarely observed in our data a mono-exponential decay). We made no assumption on the waveform and time constant of calcium signal decays (we rarely observed in our data a mono-exponential decay). Deconvolution, like all inverse problems, is very sensitive to noise; Yaksi and Friedrich (2006) thus applied a clever but time-intensive local lowpass filtering on their 30 Hz-calcium traces. Our algorithm performed well with a simple three-point median filter on 500 Hz-line scans.

Spatial sampling
A great strength of Yaksi and Friedrich’s (2006) approach is the high density of their spatial sampling: they could image activity in tens to hundreds of neurons quasi-simultaneously, after extracellular application of the calibration procedure could be done in intact, in vivo conditions, thus providing information about the transform’s parameters for neurons in their natural state, in the absence of any artificial manipulation.

CONCLUSION
Using a simple algorithm and minimal calibration, we estimated with good accuracy the smoothed firing rates of principal sensory neurons from dendritic calcium signals. By examining carefully the relationship between calcium and spiking, we found that, while decipherable, the relationship between the two is not a simple linear scaling, even after appropriate thresholding; sources of possible error are many, illustrating the need for care in using calcium imaging as a proxy for electrophysiology. In many insects (moth, honey bee, cockroach, fly) the glomerular structure of the antennal lobes is highly ordered (Couto et al., 2005; Fishilevich and Vosshall, 2005; Galizia et al., 1999), usually apparently, and easily accessible for imaging techniques. Previous work (Carlsson et al., 2005; Sachs and Galizia, 2002) showed that many PNs-glomeruli can be labeled simultaneously with dextran-based calcium indicators injected in an axon tract. Genetic calcium indicators are now also available (Miesenbock and Kevrekidis, 2005; Miyawaki et al., 1999; Nakai et al., 2001; Pologruto et al., 2004; Reiff et al., 2005; Wang et al., 2003, 2004) and thus usable in certain model systems. It is hoped that the algorithm we propose—as that developed by Yaksi and Friedrich (2006)—combined with increasingly optimized imaging techniques (Iyer et al., 2006; Roorda et al., 2004; Salome et al., 2006; Svoboda and Yasuda, 2006), will be useful to the investigation of electrical activity in large populations of neurons in vivo.

CONFLICT OF INTEREST STATEMENT
The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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