A Fast-Conducting, Stochastic Integrative Mode for Neocortical Neurons In Vivo

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During activated states, neocortical neurons receive intense synaptic background activity that induces large-amplitude membrane potential fluctuations and a strong conductance in the membrane. However, little is known about the integrative properties of neurons during such high-conductance states. Here we investigated the integrative properties of neocortical pyramidal neurons under in vivo conditions simulated by computational models. We show that the presence of high-conductance fluctuations induces a stochastic state in which active dendrites are fast conducting and have a different dynamics of initiation and forward-propagation of Na⁺-dependent spikes. Synaptic efficacy, quantified as the probability that a synaptic input specifically evokes a somatic spike, was approximately independent of the dendritic location of the synapse. Synaptic inputs evoked precisely timed responses (milliseconds), which also showed a reduced location dependence. This scheme was found to apply to a broad range of kinetics and density distributions of voltage-dependent conductances, as well as to different dendritic morphologies. Synaptic efficacies were, however, modulable by the balance of excitation and inhibition in background activity, for all synapses at once. Thus, models predict that the intense synaptic activity in vivo can confer advantageous computational properties to neocortical neurons: they can be set to an integrative mode that is stochastic, fast conducting, and optimized to process synaptic inputs at high temporal resolution independently of their position in the dendrites. Some of these predictions can be tested experimentally.

Key words: computational models; random synaptic inputs; noise; high-conductance state; synaptic integration; dendritic democracy

Introduction

How the extended dendritic trees of central neurons integrate synaptic inputs is a problem that must be solved to understand how information is processed or coded in neurons (Yuste and Tank, 1996; Magee, 2000; Stuart et al., 2000). The high precision of patch-clamp and whole-cell recordings, together with the possibility of a fine control of synaptic inputs in vitro, has allowed significant advances in this field (Cash and Yuste, 1999; Pouille and Scanziani, 2001; Williams and Stuart, 2002). Cortical neurons possess several types of voltage- and calcium-dependent ion channels in their dendrites (Llinás, 1975; Johnston et al., 1996; Yuste and Tank, 1996; Stuart et al., 2000), which may significantly affect the impact of synaptic inputs at the level of the soma (Crill and Schwindt, 1995; Stuart and Sakmann, 1995; Williams and Stuart, 2000a; Berger et al., 2001), and generate calcium- and sodium-dependent spikes in dendrites (Spencer and Kandel, 1961; Wong et al., 1979; Benardo et al., 1982; Regehr et al., 1993; Andreasen and Lambert, 1995). Pyramidal neurons can generate action potentials (APs) that are initiated in the axon and propagate backward in the dendritic tree (Stuart and Sakmann, 1994; Stuart et al., 1997b; Häusser et al., 2000) or APs initiated in dendrites that propagate forward to the soma (Schwindt and Grill, 1997, 1998; Stuart et al., 1997a; Golding and Spruston, 1998; Williams and Stuart, 2002).

However, how neocortical neurons integrate synaptic inputs during activated states in the intact brain is a question yet unanswer, mainly because of the technical difficulty of controlling identified subsets of synaptic inputs in vivo. Here we address this problem by using computational models of morphologically reconstructed neocortical pyramidal neurons with active dendrites. In vivo conditions were simulated by random excitatory and inhibitory synaptic inputs in soma and dendrites based on constraining the model to intracellular recordings in vivo (Destexhe and Paré, 1999). In agreement with previous theoretical (Barrett, 1975; Holmes and Woody, 1989; Bernander et al., 1991; Rapp et al., 1992) and experimental (Borg-Graham et al., 1998; Paré et al., 1998b) studies, this approach revealed that background activity in vivo is responsible for a major tonic increase of conductance compared with quiescent states.

Thus, in vitro measurements have demonstrated the importance of active dendritic currents to capture the subthreshold and superthreshold dynamics of dendrites. On the other hand, in vivo studies demonstrated that cortical neurons are subject to high-conductance fluctuating activity during activated states of the brain. Here we use models based on both in vitro and in vivo measurements in an attempt to characterize the effect of these high-conductance fluctuations on the integrative properties of neocortical pyramidal neurons.

Materials and Methods

Computational models of morphologically reconstructed neocortical pyramidal neurons were simulated using NEURON (Hines and Carnevale, 1997) and were constrained by experimental data obtained from in vitro and in vivo preparations, as detailed below.
**Dendritic morphologies.** Dendritic morphologies were obtained from three-dimensional reconstructions of four pyramidal cells (one from layer II–III, two from layer V, and one from layer VI) obtained from cat cortex (Douglas et al., 1991; Contreras et al., 1997). The cellular geometries were corrected for spines assuming that spines represent ~45% of the dendritic membrane area (DeFelipe and Farinas, 1992).

**Passive properties.** Passive properties (leak conductance, reversal potential, and axial resistance) were obtained by fitting the model to passive responses obtained intracellularly after application of tetrodotoxin and synaptic blockers (Paré et al., 1998b). Two sets of passive properties were used: (1) a set with uniform leak conductance obtained from sharp-electrode recordings (Destexhe and Paré, 1999) and (2) a non-uniform leak model with low axial resistance, as obtained from dual-patch recordings (Stuart and Spruston, 1998).

**Active properties.** Active properties were simulated using Hodgkin and Huxley (Hodgkin and Huxley, 1952) type models for voltage-dependent Na⁺, K⁺, and Ca²⁺ conductances. The densities used were as follows (in mS/cm²): 3–12 (soma and dendrites) for Na⁺ and 5–10 (soma and dendrites) for K⁺. Densities in the axon were chosen to be 5–10 times higher in the initial segment and nodes of Ranvier. Kinetics of the currents were taken from a model of hippocampal pyramidal cells (Taub and Miles, 1991), in which Na⁺ inactivation was shifted by 10 mV toward more hyperpolarized values to match voltage-clamp data of cortical pyramidal cells (Huguenard et al., 1998). Results were checked using different kinetic models for Na⁺ and K⁺ currents, as well as for different positions of the steady-state inactivation of Na⁺ channels. In some simulations, Ca²⁺ and Ca²⁺ -dependent K⁺ currents, as well as A-type K⁺ currents, were used (for details, see Appendix).

**Synaptic currents.** Synaptic currents were simulated by two-state kinetic models for glutamate, AMPA, NMDA, and GABA type-A (GABAA) receptor types (Destexhe et al., 1994). The densities of synapses were calculated in different regions of the cell based on morphological data (White, 1989; Larkman, 1991; DeFelipe and Farinas, 1992) and were (per 100 µm² of membrane) 10–20 (GABAA, soma), 40–80 (GABAA, axon initial segment), 8–12 (GABAA, dendrites), and 55–65 (AMPA–NMDA, dendrites). This led to a total of 49,699 glutamatergic and 10,669 GABAergic synapses for the layer V cell shown in Figure 1A (16,563 and 3376, respectively, for the layer VI cell shown in Fig. 1B). Quantal conductances were assumed to be uniform (Williams and Stuart, 2002) and were estimated from fitting the model to recordings of miniature synaptic events (Destexhe and Paré, 1999). The quantal conductances obtained were of 1.2 and 0.6 nS for glutamatergic and GABAergic synapses, respectively.

**Synaptic background activity.** Synaptic background activity was simulated by random (Poisson-distributed) release events at all synapses. The release parameters were estimated by fitting the model to intracellular recordings in vivo before and after suppression of background activity (Paré et al., 1998b; Destexhe and Paré, 1999). Release rates in the range of 0.1–1 and 0.55–5.5 Hz at glutamatergic and GABAergic synapses, respectively, gave average membrane potentials, input resistances, and fluctuation levels consistent with intracellular recordings. A correlation was included between release events (cross-correlation peak of c = 0.1) for details, see Destexhe and Paré, 1999). In these conditions, we calculated that the total membrane conductance attributable to inhibition is approximately four to five times larger than excitation, and, given the difference in driving force, excitatory and inhibitory currents are approximately balanced (with a slight excess for excitation).

**Correlations.** Correlations were introduced by forcing some of the synapses to corelease while keeping the random nature of the release at each synapse. This was achieved by generating Np Poisson-distributed random presynaptic trains and by redistributing these trains among the N synaptic sites in the model. If Np < N, all synapses still released randomly with identical statistical properties, but, at any given instant, some of the N synapses released simultaneously and were therefore “corelated.” The Np synapses were redistributed randomly among the N synapses at every time step, such that the average correlation was the same for every pair of synapses, regardless of their location in the dendritic tree. Because correlations selectively affect the amplitude of voltage fluctuations (Destexhe and Paré, 1999), this procedure can be used to control voltage fluctuations (by changing Np), with no change in the average conductance and membrane potential of the cell. Note that the correlation used corresponds to a pairwise Pearson correlation of ~0.1, which is consistent with the values measured experimentally for the “background” correlation in cerebral cortex (Zohary et al., 1994; Vaadia et al., 1995).

**Static conductance.** In some simulations, background activity was replaced by an equivalent static conductance. This conductance was obtained by inserting in each compartment a supplemental leak conductance, which was calculated to equate the average activity of the synapses converging to that compartment. The model obtained had a membrane potential, input resistance, and time constant that were equivalent to the model with background activity but had no membrane potential fluctuations.

**Synaptic stimuli.** Synaptic stimuli consisted of a supplementary set of AMPA-mediated synaptic conductances inserted at different locations in the dendritic tree. Stimulation intensity was adjusted by varying the number of synchronously activated AMPA synapses (quantal conductance of 1.2 nS), colocalized at the same dendritic sites. The simulation was repeated every 50 msec. Successive stimuli can be considered as independent because the period was large compared with the typical duration of the responses (Fig. 2B). For each site, a total of 1200 stimulations (trials) were used to calculate the poststimulus time histogram (PSTH). For each parameter set, the model was run twice, with and without stimulus, and the spikes specifically evoked by the stimulus were obtained by subtracting spikes attributable to background activity. The time integral of the PSTH gives the probability that a somatic spike is specifically evoked by the stimulus, which is used as a measure of synaptic efficacy (for other measures, see London et al., 2002).

**Results.** We start by showing that background activity induces a stochastic dynamics that affects dendritic AP initiation and propagation. We next investigate the impact of individual synapses at the soma in this stochastic state, as well as how synaptic efficacy is modulated by different factors, such as morphology and the intensity of background activity itself. Finally, we investigate how this stochastic state affects the timing of synaptic events as a function of their position in the dendrites.

**A stochastic state with facilitated action potential initiation** We first characterized how synaptic background activity affects the dynamics of AP initiation and propagation in dendrites. Dendritic AP propagation was simulated in computational models of morphologically reconstructed cortical pyramidal neurons, which included voltage-dependent currents in soma, dendrites, and axon (Fig. 1A, top) (see Materials and Methods). In quiescent conditions, backpropagating dendritic APs were reliable up to a few hundred micrometers from the soma (Fig. 1A, bottom, Quiescent), in agreement with dual soma–dendrite recordings in vitro (Stuart and Sakmann, 1994; Stuart et al., 1997b). In the presence of synaptic background activity, backpropagating APs were still robust but propagated over a more limited distance in the apical dendrite compared with quiescent states (Fig. 1A, bottom, In vivo-like), consistent with the limited backward invasion of apical dendrites observed with two-photon imaging of cortical neurons in vivo (Svoboda et al., 1997).

APs could also be initiated in dendrites after simulated synaptic stimuli. In quiescent conditions, the threshold for dendritic AP initiation was high (Fig. 1B, left, Quiescent), and the dendritic-initiated APs propagated forward only over limited distances (100–200 µm) (Fig. 1C, Quiescent), in agreement with previous observations (Stuart et al., 1997a; Golding and Spruston, 1998; Vetter et al., 2001). Interestingly, background activity tended to facilitate forward-propagating APs. Dendritic AP initiation was highly stochastic because of the presence of random fluctuations, but computing the probability of AP initiation revealed a significa
icant effect of background activity (Fig. 1B, left, In vivo-like). The propagation of initiated APs was also stochastic, and a significant fraction (see below) of dendritic APs could propagate forward over large distances and reach the soma (Fig. 1C, In vivo-like), a situation that did not occur in quiescent states with low densities of Na⁺ channels in dendrites.

To explain this effect of background activity on dendritic APs, we compared different background activities with equivalent conductance but different amplitudes of voltage fluctuations (see Materials and Methods). Figure 1B (right) shows that the probability of AP initiation, for fixed stimulation amplitude and path distance, was zero in the absence of fluctuations but steadily raised for increasing fluctuation amplitudes (all simulations were at equivalent voltage). This shows that subthreshold stimuli are occasionally boosted by depolarizing fluctuations. Propagating APs can also benefit from this boosting to help their propagation all the way up to the soma. In this case, the AP itself must be viewed as the stimulus that is boosted by the presence of depolarizing fluctuations. The same picture was observed for different morphologies, passive properties, and various densities and kinetics of voltage-dependent currents (see below): in vivo-like activity induced a stochastic dynamics in which backpropagating APs were minimally affected, but forward-propagating APs were facilitated. Thus, under in vivo-like conditions, subthreshold EPSPs can be occasionally boosted by depolarizing fluctuations and have a chance to initiate a dendritic AP, which itself has a chance to propagate and reach the soma.

**Location independence of the impact of individual or multiple synapses**

We next evaluated quantitatively the consequences of this stochastic dynamics of dendritic AP initiation in terms of the impact of individual EPSPs at the soma. In quiescent conditions, the model was adjusted to the passive parameters estimated from whole-cell recordings in vitro (Stuart and Spruston, 1998), yielding a relatively moderate passive voltage attenuation (Fig. 2A, Quiescent; 25–45% attenuation for distal events). Taking into account the high conductance and more depolarized conditions of in vivo-like activity showed a marked increase in voltage attenuation (Fig. 2A, In vivo-like; 80–90% attenuation). Computing the EPSP peak amplitude in
these conditions revealed an attenuation with distance (Fig. 2A, lower panel), which was more pronounced if background activity was represented by an equivalent static (leak) conductance (see Materials and Methods). Thus, the high-conductance component of background activity enhances the location-dependent impact of EPSPs and leads to a stronger individualization of the different dendritic branches (London and Segev, 2001; Rhodes and Llinás, 2001). A radically different conclusion was reached if voltage fluctu-
locations were taken into account. In this case, responses were highly irregular, and the impact of individual synapses was assessed by computing the PSTH over long periods of time with repeated stimulation of single or groups of colocalized excitatory synapses (see Materials and Methods). The PSTHs obtained for stimuli occurring at different distances from the soma (Fig. 2B) show that the “efficacy” of these synapses is approximately location independent, as calculated from either the peak (Fig. 2C) or the integral of the PSTH (Fig. 2D). The latter can be interpreted as the probability that a somatic spike is specifically evoked by a synaptic stimulus. Using this measure of synaptic efficacy, we conclude that, under in vivo-like conditions, the impact of individual synapses on the soma is nearly independent on their dendritic location, despite a severe voltage attenuation.

**Mechanisms underlying location independence**

To show that this location-independent mode depends on forward-propagating dendritic APs, we selected, for a given synaptic location, all trials that evoked a somatic spike. These trials represented a small portion of all trials: from 0.4 to 4.5% depending on the location and the strength of the synaptic stimuli. For these “successful” selected trials, the somatic spike was always preceded by a dendritic spike evoked locally by the stimulus. In the remaining “unsuccessful” trials, there was a proportion of stimuli (55–97%) that evoked a dendritic spike but failed to evoke somatic spiking. This picture was the same for different stimulation sites: a fraction of stimuli evokes dendritic spikes, and a small fraction of these dendritic spikes successfully evokes a spike at the soma–axon.

We analyzed the latter aspect by representing the probabilities of initiation and propagation along the distance axis (Fig. 2E). There was an asymmetry between these two measures: the chance of evoking a dendritic AP was lower for proximal stimuli and increased with distance (Fig. 2E, AP initiation), because the local input resistance varies inversely with dendrite diameter and is higher for thin (distal) dendritic segments. On the other hand, the chance that a dendritic AP propagates down to the soma and leads to soma–axon APs was higher for proximal sites and gradually decreased with distance (Fig. 2E, AP propagation). Remarkably, these two effects compensated such that the probability of evoking a soma–axon AP (the product of these two probabilities) was approximately independent on the distance to soma (Fig. 2E, Somatic response). This effect was observed only in the presence of conductance-based background activity and was not present in quiescent conditions or by using current-based models of synapses (data not shown). These results show that the location-independent impact of synaptic events under in vivo-like conditions is attributable to a compensation between an opposite distance dependence of the probabilities of AP initiation and propagation.

The same dynamics were present in four different pyramidal cell morphologies (Fig. 3), suggesting that this principle may apply to a large variety of dendritic morphologies. It was also robust to variations in ion channel densities and kinetics, such as NMDA conductances (Fig. 4A), passive properties (Fig. 4B), and different types of ion channels (Fig. 4C), including high distal densities of leak and hyperpolarization-activated I\(_{\text{h}}\) conductances (Fig. 4C, gray line). In the latter case, the presence of I\(_{\text{h}}\) affected EPSPs in the peri-somatic region, in which there is a significant contribution of passive signaling, but synaptic efficacy was remarkably location independent for the remaining part of the dendrites in which the I\(_{\text{h}}\) density was highest (see Appendix). Location independence was also robust to changes in membrane excitability (Fig. 4D,E) and shifts in the Na\(^+\) current inactivation (Fig. 4F). Most of these variations changed the absolute probability of evoking spikes but did not affect the location independence induced by background activity. The location-independent synaptic efficacy was lost when the dendrites had too strong K\(^+\) conductances, with either high I\(_{\text{KA}}\) in distal dendrites (Fig. 4G, black dotted line) or a high ratio between K\(^+\) and Na\(^+\) conductances (Fig. 4E). In other cases, synaptic efficacy was larger for distal dendrites (Fig. 4D, high excitability, F, inactivation shift of 0).

**Activity-dependent modulation of synaptic efficacy**

To determine how the efficacy of individual synapses varies as a function of the intensity of synaptic background activity, we repeated the same stimulation paradigms as in Figure 2 but by varying individually the release rates of excitatory (Fig. 5A) or inhibitory (Fig. 5B) inputs of the background, by varying both (Fig. 5C), or by varying the correlation with fixed release rates (Fig. 5D). In all cases, synaptic efficacy (integrated PSTH for stimuli that were subthreshold under quiescent conditions) depended on the particular properties of background activity but remained location independent. In the case of “balanced” excitatory and inhibitory inputs (Fig. 5C), background activity could be changed continuously from quiescent to in vivo-like conditions. In this case, the probability steadily rose from zero (Fig. 5C, clear region), showing that subthreshold stimuli can evoke detectable responses in the presence of background activity, and reached a “plateau” at which synaptic efficacy was independent of both...
The stimulation amplitude was, in all cases, 12 nS. Synaptic efficacy as a function of path distance and stimulation amplitude for a non-uniform passive model (Stuart and Spruston, 1998) (see Appendix). D, Synaptic efficacy as a function of path distance and membrane excitability. A both Na⁺ and K⁺ conductance densities were changed by a common multiplicative scaling factor. The dotted line indicates a dendritic conductance density of 8.4 mS/cm² for the Na⁺ current and 7 mS/cm² for the delayed rectifier K⁺ current. The stimulation amplitude was, in all cases, 12 nS. E, Synaptic efficacy obtained by changing the ratio between Na⁺ and K⁺ conductances responsible for action potentials (conductance density of 8.4 mS/cm² for I_{Na}; the dotted line indicates 7 mS/cm² for I_{Na}). F, Synaptic efficacy as a function of path distance obtained by varying the steady-state inactivation of the fast Na⁺ current. The inactivation curve was shifted with respect to the original model (Traub and Miles, 1991) toward hyperpolarized values (stimulation amplitude, 12 nS). The dotted line indicates a 10 mV shift, which approximately matches the voltage-clamp data of cortical pyramidal cells (Huguenard et al., 1988). All simulations were done using the layer VI cell in which AMPA-mediated synaptic stimuli were applied at different sites along the dendritic branch indicated by a dashed arrow in Figure 3.

Location dependence of the timing of synaptic events

We next tested whether location independence also applies to the timing of synaptic events. Figure 6A illustrates the somatic membrane potential after synaptic stimuli at different locations. In quiescent conditions, as predicted by cable theory (Segev et al., 1995; Koch, 1999), proximal synaptic events led to fast rising and fast decaying somatic EPSPs, whereas distal events were attenuated in amplitude and slowed in duration (Fig. 6A, Quiescent). The time-to-peak of EPSPs increased monotonically with distance (Fig. 6B, Quiescent). In the presence of background activity, the average amplitude of these voltage deflections was much less dependent on location (Fig. 6A, In vivo-like), consistent with the PSTHs in Figure 2B, and the time-to-peak of these events was only weakly dependent on the location of the synapses in dendrites (Fig. 6B, In vivo-like). Thus, in vivo-like conditions seem to set the dendrites into a fast-conducting mode, in which the timing of synaptic inputs shows little dependence on their distance to soma.

To investigate the basis of this fast-conducting mode, we simulated the same paradigm by varying a number of parameters. First, to check whether this effect could be attributable to the decreased membrane time constant as a result of the high conductance imposed by synaptic background activity, we replaced background activity by an equivalent static conductance, which led to an intermediate location-dependent relationship (Fig. 6C, Quiescent, static conductance) between the quiescent and in vivo-like cases depicted above. The reduced time constant therefore can account for some, but not all, of the diminished location dependence of the timing. Second, to check for contributions of dendritic Na⁺ channels, we ran the same stimulation protocol under in vivo-like conditions but by selectively removing Na⁺ channels from dendrites. This also led to an intermediate location dependence (Fig. 6C, In vivo-like, g_{Na} = 0), suggesting that Na⁺-dependent mechanisms underlie the further reduction of timing beyond the high-conductance effect. Finally, to show that this further reduction is attributable to dendritic APs, we used a quiescent state with equivalent static conductance but higher dendritic excitability (Na⁺ and K⁺ conductances that were twice as large), such that strong synaptic stimuli could evoke reliable forward-propagating dendritic APs. In this case only, the reduced location dependence of the timing could be fully reconstructed.
(Fig. 6C, Quiescent, static conductance, high dendritic g_{Na}). The dependence on dendritic APs was also confirmed by the intermediate location dependence obtained when EPSPs were constructed from trials that did not contain dendritic APs (Fig. 6B, No dendritic spikes). This analysis shows that the fast-conducting mode is attributable to forward-propagating APs in dendrites of fast time constant.

**Discussion**

In this paper, we modeled high-conductance states in neocortical pyramidal neurons that are associated with a depolarized and highly fluctuating membrane potential, as shown by intracellular measurements in vivo, in anesthetized (Contreras et al., 1996; Paré et al., 1998b) or awake animals (Matsumura et al., 1988; Steriade et al., 2001). The simulations suggest that these high-conductance states set pyramidal neurons into a stochastic integrative mode that is fast-conducting and in which the impact of inputs is location independent. The underlying mechanism depends on forward-propagating APs. The high conductance of background activity lowers the membrane time constant and enhances passive voltage attenuation; this enhanced attenuation limits the passive spread to soma and the triggering of backpropagating APs by dendritic stimuli and therefore leaves the opportunity for stimuli to influence the soma by forward-propagating APs. At the same time, the voltage fluctuations associated with synaptic activity induce a stochastic dynamics that facilitates the initiation and propagation of dendritic APs (this effect is consistent with the facilitation of dendritic APs by EPSPs shown by Stuart and Häusser, 2001). The combined result of these actions is that colocalized synaptic events have a small (but non-zero) chance to trigger a dendritic spike, which itself has a chance to propagate rapidly to the soma in which it can participate in firing the soma–axon. In contrast with an axon, however, the dynamics is here stochastic in nature, and there is a complex dependence on distance for the probability of AP initiation and propagation (Fig. 2E). These effects compensate such that the impact of a given synapse is approximately independent of its position in the dendrite. Different regions of the neuron, including distal dendrites, can therefore efficiently vote for firing the cell. The same dynamics was observed for different morphologies (Fig. 3), different passive and active properties (Fig. 4), and various background activities (Fig. 5).

This stochastic integrative mode was present despite considerable variations in the type or kinetics of ions channels in dendrites. It was present with voltage-dependent Ca^{2+} channels or with high distal densities of the hyperpolarization-activated current I_{h} (Fig. 4C), similar to the densities measured in the apical dendrite of layer V neocortical pyramidal neurons (Stuart and Spruston, 1998; Larkum et al., 1999a,b; Williams and Stuart, 2000a; Berger et al., 2001) (for a model, see Rhodes and Linás, 2001). Our study is therefore applicable to a wide range of dendritic excitabilities typical of cortical neurons. Because the mechanism is dependent on forward dendritic spikes, it can fail in situation with suppressed dendritic spike activity, such as strong dendritic K^+ conductances (Fig. 4C, I_{Na*}, I_{Kd*}, I_{Kf*}; E, large K/Na ratio). This suggests that this mechanism would be particularly relevant to states in which K^+ conductances are downregulated. This is the case for aroused states, which are maintained by neuromodulators such as acetylcholine, noradrenaline, or serotonin, all of which downregulate K^+ conductances in neocortical neurons (McCormick, 1992).

The fact that the same dynamics can be reproduced in different morphologies (Fig. 3) suggests that the location-independent efficacy of synapses could apply to different types of neurons, regardless of the complexity of their dendritic arborization. If this principle is true, cortical neurons would be free to optimize their dendritic arbor based on connectivity constraints only (and not on synaptic weight for example). This is valid, however, only for colocalized synaptic events, as investigated here. There is a large spectrum of possibilities that should be analyzed, including the synchrony of multiple inputs, their kinetics and receptor type, and the joint impact of paired synaptic events as a function of their relative distance and timing. Such an analysis may reveal a dependence on the local structure of the dendrites, similar to the “dendritic subunits” postulated in previous models of dendritic integration (Mel, 1994).

Another interesting property of this stochastic integrative mode is that synaptic efficacy is modulable by network activity. Changing the balance of excitation–inhibition, release rates, or correlation in background activity can reconfigure synaptic efficacies, but it does so by preserving location independence (Fig. 5). The model predicts that the highest synaptic efficacies should be obtained for states of intense “spontaneous” network activity, a somewhat counter-intuitive result. Note that the stochastic aspect also implies that neocortical neurons would have to use population codes to reliably encode information. In this respect,
the modulation of synaptic efficacies by network activity suggests that the number of neurons that respond to a given stimulus is determined by the spontaneous activity of the network.

Location independence was found and modeled in a number of previous studies (De Schutter and Bower, 1994; Magee and Cook, 2000; Andrásfalvy and Magee, 2001; London and Segev, 2001; Rudolph et al., 2001). In hippocampal pyramidal neurons, synaptic conductances are scaled according to their distance to soma to compensate for dendritic filtering (Magee and Cook, 2000; Andrásfalvy and Magee, 2001) (but see London and Segev, 2001), a situation that does not seem to apply to neocortical neurons (Williams and Stuart, 2002). In Purkinje cells, background activity can induce location independence for synchronized synaptic inputs by a subthreshold, Ca$^{2+}$-mediated mechanism (De Schutter and Bower, 1994). In contrast, we show here a stochastic mechanism based on Na$^{+}$-dependent APs that equally...
applies to single or synchronized synaptic inputs. It would be interesting to investigate the same behavior in other cell types, such as thalamic neurons, which also contain Na\(^+\) and Ca\(^{2+}\) channels in their dendrites (Destexhe et al., 1998; Williams and Stuart, 2000b) and are subject to intense synaptic background activity in vivo (Contreras et al., 1996).

Some of the present observations suggest additional theoretical work. First, the fact that dendrites subject to high-conductance fluctuations can compensate for the opposite location-dependent effects of AP initiation and propagation (Fig. 2E) should be investigated using the stochastic cable theory to understand the underlying mechanisms. Second, the observation of an increased traffic of Na\(^+\)–dependent dendritic APs in these conditions suggests possible consequences on spike-timing dependent plasticity (STDP). STDP is thought to occur according to the relative timing between release events and postsynaptic APs (for review, see Bi and Poo, 2001). The present results suggest that, during active states, a significant fraction of postsynaptic APs are local dendritic spikes, which opens the possibility for local dendritic regions to manage their synaptic plasticity mechanisms independently of the soma (Golding et al., 2002). Additional work is needed to evaluate to what extent such a local plasticity rule could duplicate the computational power of cortical neurons (Mel, 1994).

Finally, this model formulates a series of experimentally testable predictions. The predicted stochastic initiation and propagation of dendritic APs should be observable by single-cell imaging techniques in vivo (Svoboda et al., 1997), if such recordings can be made in activated states with desynchronized EEG. The prediction that location-dependent synaptic efficacies should become location independent in the presence of intense background activity could be tested by using dual soma–dendrite recordings (for methods, see Stuart and Sakmann, 1994; Larkum et al., 1999a; Berger et al., 2001; Williams and Stuart, 2002). This could be performed in slices either with artificially induced active states (Buhl et al., 1998; Brumberg et al., 2000) or by recreating in vivo-like activity by intracellular injection of fluctuating conductances (Destexhe et al., 2001). In both cases, protocols similar to Figure 2B–D could be used to calculate the efficacy of synaptic conductance waveforms injected in dendrites and to compare the efficacy between quiescent and active states.

### Appendix

This appendix contains details of the different models used in the paper. Computational models and scripts for the NEURON simulation environment are available at http://cns.iaf.cnrs-gif.fr/supplement.html.

### Morphology

Simulations were performed using multicompartment models of morphologically reconstructed neocortical pyramidal neurons from layer II–III, V, and VI of cat parietal cortex (Contreras et al., 1997), as well as a layer V cell from cat cortex (Douglas et al., 1991) (Fig. 3) with both simple and detailed axonal geometry (Mainen et al., 1995; Paré et al., 1998a). In all models, passive properties and active conductances in dendrites were changed to account for the surface correction attributable to dendritic spines, assuming that ~45% of the dendritic membrane area are represented by spines (DeFelipe and Farinas, 1992).

### Uniform passive model

Passive model parameters were adjusted to intracellular (sharp-electrode) recordings obtained after application of TTX and synaptic blockers (Destexhe and Paré, 1999), yielding an axial resistance \( R_s = 250 \, \Omega \, \text{cm} \), a membrane resistance \( R_m = 22 \, \text{k}\Omega \, \text{cm}^2 \) in soma and dendrites (\( R_m = 50 \, \text{\Omega} \, \text{cm}^2 \) in axon), and membrane capacity \( C_m = 1 \, \mu\text{F/cm}^2 \) in soma and dendrites (\( C_m = 0.04 \, \mu\text{F/cm}^2 \) in initial and myelinated segments of axon).

### Nonuniform passive model

A non-uniform leak model obtained from dual-patch recordings in neocortical layer V pyramidal neurons (Stuart and Spruston, 1998) was used with \( C_m = 1.54 \, \mu\text{F/cm}^2 \) in soma and dendrites (\( C_m = 0.04 \, \mu\text{F/cm}^2 \) in initial and myelinated segments of axon), and \( R_s = 68 \, \text{\Omega} \, \text{cm} \) and \( R_m = 50 \, \text{\Omega} \, \text{cm}^2 \) in axonal nodes of Ranvier. The membrane resistance in dendrites scaled in a sigmoidal manner with path distance, with values \( R_m(\text{end}) = 5.36 \, \text{k}\Omega \, \text{cm}^2 \), \( R_m(\text{soma}) = 39.06 \, \text{k}\Omega \, \text{cm}^2 \), steep = \( 50 \, \mu\text{m} \), and \( d_{\text{half}} = 300 \, \mu\text{m} \).

### Active membrane conductances

Voltage-dependent conductances were inserted in the soma, dendrites, and the axon. Currents included the fast sodium current \( I_{\text{Na}} \), the delayed rectifier potassium current \( I_{\text{Kd}} \), a slow voltage-dependent potassium current \( I_{\text{Kd}} \), a Ca\(^{2+}\)-dependent potassium current \( I_{\text{CaL}} \), a persistent sodium current \( I_{\text{NaP}} \), an A-type potassium current \( I_{\text{KA}} \), and a hyperpolarization-activated conductance \( I_{\text{h}} \). All currents were described by Hodgkin–Huxley-type models (Hodgkin and Huxley, 1952). Various combinations and densities in dendrites and soma were used, as follows.

\[
\begin{align*}
I_{\text{Na}}, I_{\text{Kd}}, I_{\text{M}} \\
I_{\text{Na}}, I_{\text{Kd}}^{\text{p}}, I_{\text{M}}, I_{\text{Kd}}^{\text{C}}, I_{\text{CA}} \\
I_{\text{Na}}, I_{\text{Kd}}^{\text{p}}, I_{\text{M}}, I_{\text{Kd}}^{\text{C}}, I_{\text{CA}}, I_{\text{Kd}} \\
I_{\text{Na}}^{\text{p}}, I_{\text{Kd}}^{\text{p}}, I_{\text{M}}^{\text{p}}, I_{\text{Kd}}^{\text{C}}, I_{\text{CA}}^{\text{p}}, I_{\text{Kd}}^{\text{p}}
\end{align*}
\]

The standard kinetic models for these currents were taken from a model of hippocampal pyramidal cells (see Materials and Methods) (Traub and Miles, 1991), adjusted to match voltage-clamp data of cortical pyramidal cells (Huguenard et al., 1988). In the standard setup, constant peak conductance densities of 8.4 mS/cm\(^2\) (soma and dendrites; 84 mS/cm\(^2\) in axon initial segment and nodes of Ranvier) for \( I_{\text{Na}} \), 7 mS/cm\(^2\) (soma and dendrites; 70 mS/cm\(^2\) in axon initial segment and nodes of Ranvier) for \( I_{\text{Kd}} \), and 0.35 mS/cm\(^2\) (soma and dendrites) for \( I_{\text{h}} \) (no \( I_{\text{h}} \) in axon) were used. These densities correspond to the values found experimentally in adult hippocampal pyramidal neurons (Magee and Johnston, 1995).

\[
\begin{align*}
I_{\text{Na}}^{\text{p}}, I_{\text{Kd}}^{\text{p}}, I_{\text{M}}^{\text{p}}, I_{\text{Kd}}^{\text{p}}, I_{\text{CA}}^{\text{p}}, I_{\text{Kd}}^{\text{p}}
\end{align*}
\]

These currents are as the standard model with additional Ca\(^{2+}\)-dependent potassium current (for kinetics, see Yamada et al., 1989) with conductance density of 1 mS/cm\(^2\) in dendrites and soma and high-threshold Ca\(^{2+}\)-current \( I_{\text{CaL}} \) (for kinetics, see McCormick and Huguenard, 1992) with peak conductance density of 3 and 1.5 mS/cm\(^2\) for proximal and distal dendrites, respectively.

\[
\begin{align*}
I_{\text{Na}}^{\text{p}}, I_{\text{Kd}}^{\text{p}}, I_{\text{M}}^{\text{p}}, I_{\text{Kd}}^{\text{p}}, I_{\text{CA}}^{\text{p}}, I_{\text{Kd}}^{\text{p}}
\end{align*}
\]

These currents are as the standard model with an additional persistent sodium current \( I_{\text{NaP}} \) (for kinetics, see French et al., 1990; McCormick and Huguenard, 1992) with conductance density of 0.1 mS/cm\(^2\) in soma and dendrites.

\[
\begin{align*}
I_{\text{Na}}^{\text{p}}, I_{\text{Kd}}^{\text{p}}
\end{align*}
\]

These currents are described by a different model of action potential generation, in which the fast Na\(^+\) current and the delayed rectifier K\(^+\) current were taken from a previous model of cortical pyramidal neurons (Mainen et al., 1995). Peak conductance densities were 8.4 mS/cm\(^2\) (soma and dendrites; 84 mS/cm\(^2\) in axon initial segment) for \( I_{\text{Na}}^{\text{p}} \) and 7.0 mS/cm\(^2\) (soma and dendrites; 70 mS/cm\(^2\) in axon initial segment) for \( I_{\text{Kd}}^{\text{p}} \).
I_{Na}^{*}, I_{Kd}^{*}, I_{KA}

These currents are according to a model of action potential generation for hippocampal pyramidal neurons (Migliore et al., 1999), including an A-type K^+ current, which conductance density linearly increased with path distance (for details, see Migliore et al., 1999). Conductance densities were as follows: 7.4 mS/cm^2 (soma and dendrites; 1800 mS/cm^2 in axon initial segment and nodes of Ranvier) for I_{Na}^{*}, 10 mS/cm^2 (soma and dendrites; 2500 mS/cm^2 in axon initial segment and nodes of Ranvier) for I_{Kd}^{*}, and 48 mS/cm^2 for I_{KA} in the soma. Two different kinetics for proximal and distal I_{KA} were used, with densities increasing with 48 mS/cm^2 per 100 μm.

I_{Na}, I_{Kd}, I_{h}

These currents are as the standard model for I_{Na} and I_{Kd} with an additional hyperpolarization-activated conductance I_{h} (kinetics and density according to non-uniform model by Stuart and Spruston, 1998): sigmoidal scaling with path distance, peak conductance density of 0.02 mS/cm^2 at soma and 20 mS/cm^2 in most distal dendrites, d_{half} = 400 μm, and steep = 50 μm.

References


