

## Chapter 5      Synaptic Integration by Electro-Diffusion in Dendritic Spines

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### I. Introduction

Many vertebrate and invertebrate neurons receive synaptic inputs on spines (Coss and Perkel, 1985). In hippocampal pyramidal cells, almost all synapses found on spines are excitatory (Harris and Stevens, 1989), but in the cat primary visual cortex, 7% of synapses on spines are inhibitory (Beaulieu and Colonnier, 1985), which comprise almost one-third of the total number of inhibitory synapses on a pyramidal cell. It has been suggested that shunting inhibition on a spine could perform a selective AND-NOT-like operation (Koch and Poggio, 1983b). The nonlinear effects would be localized to the spine, making it an effective computational module. In this chapter, we compare the predictions made by the cable model with those made by the electro-diffusion model for this problem.

The conduction of action potentials in axons can be accurately modeled by the Hodgkin-Huxley equation, which is based on the cable equation (Hodgkin and Huxley, 1952). The integration of postsynaptic signals in dendrites has also been studied with analytic solutions to passive cables (Rall, 1977), and recently, several investigators have used the cable model to examine the possibility of more complex signal processing in dendrites with complex morphologies, multiple synaptic inputs, and passive or excitable membranes (Shepherd *et al.*, 1985;

Koch and Poggio, 1983; Koch *et al.*, 1983; Rall and Segev, 1985; Perkel and Perkel, 1985; Wathey *et al.*, 1989). A central assumption of the cable model is that ionic concentrations do not change appreciably, so that the driving forces can be approximated by fixed batteries. However, if the intracellular volume is relatively small, as in dendritic spines, then ionic concentrations can change rapidly following a transient change in ionic conductances. Moreover, a sudden change in concentration at one location can lead to gradients of ionic concentration within a thin process, such as the neck of a spine, which violates another fundamental assumption of the cable model. Under these circumstances, it is necessary to consider the fundamental laws governing the movements of ions, as given by the Nernst–Planck equations for electro-diffusion (Jack *et al.*, 1975).

In this chapter, we summarize the results of applying an electro-diffusion model based on the Nernst–Planck equation to dendritic spines (Qian and Sejnowski, 1989, 1990). The electro-diffusion model gives more accurate predictions than the cable model, especially for small structures. The electro-diffusion model provides a unified framework for the computation of both the membrane potentials and the intracellular ionic concentrations during synaptic activation. We have also developed a modified cable model that is a better approximation to the electro-diffusion model than the standard cable model and is less demanding computationally.

## II. Cable Model Predictions

Consider first the case where the excitatory and inhibitory synapses are very close to each other. According to the cable model, the excitatory and the inhibitory synaptic currents are, respectively, given by:

$$I_e(t) = G_e(t)(V(t) - E_e) \quad (1)$$

and

$$I_i(t) = G_i(t)[V(t) - E_i] \sim G_i(t)[V(t) - V_{\text{rest}}], \quad (2)$$

where  $E_e$  and  $E_i$  are the reversal potentials of the excitatory and the inhibitory synapses,  $G_e$  and  $G_i$  are the transient synaptic conductances,  $V$  is membrane potential at the synapse, and  $V_{\text{rest}}$  is the resting membrane potential (Rall, 1977).

We have assumed in Eq. (1) that the reversal potential for shunting inhibition is very close to the resting membrane potential. In contrast, the excitatory inputs usually cause conductance increases to ions like  $\text{Na}^+$  or  $\text{Ca}^{++}$  that have a reversal potential,  $E_e$ , well above resting membrane potential. The inhibition will be effective if  $|I_i|$  is comparable to  $|I_e|$ . This requires, first, that  $G_i$  be larger than

$G_e$ . Second,  $V$  should be well above the resting membrane potential so that the driving force for the inhibition ( $V - V_{rest}$ ) is comparable to the driving force for the excitation ( $V - E_e$ ). This in turn requires that  $G_e$  be large and/or that the synapses are on small structures, such as spines or thin dendrites, where input resistances are large and small synaptic conductance change can cause a large depolarization. (Large inhibitory driving forces can also be achieved when the cell is firing an action potential.) In summary, for shunting inhibition to be effective when excitation and inhibition are located close to each other, the cable model requires that the synapses should be on small structures and  $G_i > G_e \gg G_{rest}$ , where  $G_{rest}$  is resting conductance of the membrane at the synapse.

In their analysis and simulations of shunting inhibition, Koch *et al.*, (1983) mainly considered large synaptic conductances on spines and distal (thin) dendrites that satisfy the inequalities discussed in the preceding. Their  $G_e$  was as large as 10 nS and  $G_i$  was 100 nS, but more recent physiological data suggest that  $G_e$  should be about 1 nS (Higashima *et al.*, 1986; Brown *et al.*, 1988). They also found that, for large excitatory conductances, inhibition on the direct path to the cell body was also effective, and that the most effective location for the inhibition moves toward the soma as the excitatory conductance increases (Koch *et al.*, 1982). Two opposing factors explain the phenomenon: When the inhibition is on the direct path from excitation to the soma,  $I_i$  is smaller because, at the site of inhibition, the membrane is less depolarized; but  $I_e$  is also smaller because, at the site of inhibition, the membrane is less depolarized; but  $I_e$  is also smaller because, at the site of excitation, the membrane is more depolarized. They also found that when the inhibition was more distal than the excitatory synapse, the inhibition was no longer effective. In this case, the resistance from the excitatory synapse to the cell body is much less than the resistance to the inhibitory synapse at the distal tip, so less current is shunted. Finally, Koch *et al.*, (1983) mentioned that increasing the value of the cytoplasmic resistivity and the membrane resistance increased the effectiveness of inhibition. This occurred because the membrane depolarization was larger, which made the driving force for the inhibitory current larger and the driving force of the excitatory current smaller.

### III. Limitations of the Cable Model

In the cable model, the membrane potential,  $V(z, t)$ , at distance  $z$  and time  $t$  along a cable obeys the equation (Jack *et al.*, 1975):

$$\frac{d}{dz} \frac{\partial^2 V}{\partial z^2} = C_m \frac{\partial V}{\partial t} + I_m, \quad (3)$$

where  $d$  is the diameter of the cable.  $R_i$  ( $\Omega\text{cm}$ ) is the total intracellular cytoplasmic resistivity,  $C_m$  ( $\mu\text{F}/\text{cm}^2$ ) is the membrane capacitance per unit area, and  $I_m$  ( $\text{mA}/\text{cm}^2$ ) represents the total non-capacitative membrane current density, which is the summation of all non-capacitative membrane current densities for each ionic species,  $I_{m,k}$ . If we assume that the movement of ionic species  $k$  across the membrane can be described by a membrane resistance of unit area  $R_{m,k}$  ( $\Omega\text{cm}^2$ ) in series with a battery whose electromotive force  $E_k$  is equal to the ionic equilibrium potential, then

$$I_{m,k} = \frac{V - E_k}{R_{m,k}} \quad (4)$$

and

$$I_m = \sum_k I_{m,k} = \frac{V - V_{\text{rest}}}{R_m} \quad (5)$$

where the resting membrane potential  $V_{\text{rest}}$  and the total membrane resistance  $R_m$  are given by:

$$V_{\text{rest}} = R_m \sum_k (E_k/R_{m,k}), \quad (6)$$

$$\frac{1}{R_m} = \sum_k (1/R_{m,k}). \quad (7)$$

Through these definitions, the electrical circuit can be reduced to a simpler equivalent circuit that has a single battery in series with a leak resistance. The standard equation for the cable model is obtained by substituting Eq. (5) into Eq. (3), assuming that  $V$  is measured from the resting potential  $V_{\text{rest}}$  (Rall, 1977):

$$\lambda^2 \frac{\partial^2 V}{\partial z^2} = \tau_m \frac{\partial V}{\partial t} + V, \quad (8)$$

where the space and time constants are defined as

$$\lambda = (d R_m / 4 R_i)^{1/2}, \quad (9)$$

$$\tau_m = R_m C_m. \quad (10)$$

The electromotive forces of the membrane batteries (equilibrium potentials) in the cable model are usually obtained from the Nernst equation and are considered constants. This is a good approximation in the squid giant axon or other large neurons, but may introduce errors if the concentrations of some ions change significantly. This applies to  $\text{Ca}^{++}$  in many situations and to synaptic events in

small structures such as dendritic spines (Rall, 1978; Koch and Poggio, 1983a; Zador *et al.*, 1990).

A second limitation of the cable model is in the treatment of longitudinal spread of current within neurons. In the cable model, the gradient of the electrical potential in the cytoplasm is the driving force for the ionic current, but there is no provision for the driving forces due to concentration gradients. This is usually a good assumption, but it may not be valid for small structures like dendritic spines where the spatial concentration gradients can be very large.

Finally, different ions may have different concentration-dependent cytoplasmic resistivities, but the cable model only incorporates the total cytoplasmic resistivity. This may not be a valid approximation when the concentration of ions are changing differentially. In summary, one expects that the cable model may not be appropriate when spatial and/or temporal ionic concentration changes are large and, especially when ionic concentration changes need to be determined.

#### IV. Electro-Diffusion Model Predictions

The cable model fails for small structure and large conductance changes, precisely the conditions required for effective shunting inhibition by the cable model (Qian and Sejnowski, 1988, 1989). The electro-diffusion model predicts that the shunting inhibition cannot be effective on small structures for the following reasons. Consider first the case when the conductance changes are large. If the inhibitory current is carried by  $\text{Cl}^-$  ions, then during a large conductance change the  $\text{Cl}^-$  concentration in a small structure such as a spine or a thin dendrite will very rapidly increase. The Nernst potential for  $\text{Cl}^-$  becomes more positive and the inhibition is ineffective. Changes in the  $\text{Cl}^-$  Nernst potential have been reported (Griffith *et al.*, 1986; Huguenard and Alger, 1986). If the conductance changes are small, then the concentration changes for  $\text{Cl}^-$  are small and the electro-diffusion model will reduce to the cable model. Thus, shunting inhibition will not be effective because the membrane depolarization is small and the driving force for the inhibitory current is much smaller than that for the excitatory current, as discussed in section II. As a consequence, the electro-diffusion model predicts that the shunting inhibition can never be very effective in small structures.

A similar analysis can be applied for hyperpolarizing inhibition carried by  $\text{K}^+$ . When both the excitatory and the inhibitory synaptic conductances are large on a small structure,  $\text{K}^+$  hyperpolarizing inhibition is just as ineffective as the  $\text{Cl}^-$  shunting inhibition because of large ionic concentration changes. However, the situation for small synaptic conductances is different. The reversal potential

for  $K^+$  is sufficiently below the resting potential that the driving force for the inhibition can be large even at the resting potential. In addition, the intracellular  $K^+$  concentration is much higher than  $Cl^-$  and, therefore, the percentage change is usually smaller. These statements will be made quantitatively precise in the model and the numerical simulations presented next.

#### A. Electro-Diffusion Model

The movement of ions in neurons is governed by the Nernst-Planck equation (Jack *et al.*, 1975):

$$\bar{J}_k = -D_k(\bar{\nabla}n_k + (n_k/\alpha_k)\bar{\nabla}V), \quad (11)$$

where  $V$  is the potential,  $\bar{J}_k$  is the flux of ionic species  $k$  (number of particles per unit area),  $D_k$  is the diffusion constant,  $n_k$  is the concentration, and the constant  $\alpha_k$  is defined as

$$\alpha_k = \frac{RT}{Fz_k}, \quad (12)$$

where  $z_k$  is the valence of ionic species  $k$ ,  $R$  is the gas constant,  $F$  is the Faraday constant, and  $T$  is the absolute temperature. The ionic concentrations and ionic currents must additionally satisfy the continuity equation:

$$\bar{\nabla} \cdot \bar{J}_k + \frac{\partial n_k}{\partial t} = 0. \quad (13)$$

Consider a cylinder of diameter  $d$  and assume that the longitudinal current and ionic concentrations are uniform across the transverse cross section of the cylinder. Assume also that transverse currents occur only at the surface of the cylinder and are independent of angle around the axis of the cylinder. These assumptions reduce the problem of electro-diffusion to a one-dimensional problem along the axis of the cylinder. The equations can be written in cylindrical coordinates and reduced to a single equation for the concentration as a function of the distance along the  $z$  axis of a cylinder:

$$\frac{\partial n_k}{\partial t} = D_k \frac{\partial^2 n_k}{\partial z^2} + \frac{D_k}{\alpha_k} \frac{\partial}{\partial z} \left( n_k \frac{\partial V}{\partial z} \right) - \frac{4}{d} J_{m,k}, \quad (14)$$

where  $J_{m,k}$  is the membrane flux of ionic species  $k$ , positive for outgoing flux.

Equation (14) must be supplemented by an additional constraint between the membrane potential and the ionic concentrations. We adopt the same capacitive model of the membrane used in the cable model; that is, we assume that the

potential change in a short segment of a process is equal to the change of the total charge in the segment divided by its membrane capacitance:

$$V(z, t) = V_{\text{rest}} + (Fd/4C_m) \sum_k [n_k(z, t) - n_{k,\text{rest}}]z_k, \quad (15)$$

where  $V_{\text{rest}}$  is the initial potential and  $n_{k,\text{rest}}$  is the initial ionic concentration of species  $k$ .

Neuronal processes often branch and change their diameters. If branches are allowed, then these equations must be solved on a tree rather than a line. Continuous diameter changes can be approximated by segments having piecewise constant diameters. At points where the diameter jumps and/or branches occur, the solutions can be matched using the continuity of flux at that point. The continuity constraint at a branch point can be derived from Eq. (11). The continuity constraint at a branch point where three processes join is given by:

$$d_1^2 \left( \frac{\partial n_k}{\partial z} + \frac{n_k}{\alpha_k} \frac{\partial V}{\partial z} \right) \Big|_1 = d_2^2 \left( \frac{\partial n_k}{\partial z} + \frac{n_k}{\alpha_k} \frac{\partial V}{\partial z} \right) \Big|_2 + d_3^2 \left( \frac{\partial n_k}{\partial z} + \frac{n_k}{\alpha_k} \frac{\partial V}{\partial z} \right) \Big|_3, \quad (16)$$

where  $d_i$  is the diameter of the  $i$ th branch.

The square of the diameter enters into this equation because the flux through the areas of each branch must be matched. There is an analytic solution of the cable model for branching dendrites having passive membranes if Rall's *3/2 power law* (Rall, 1977) is satisfied. This law for an equivalent cylinder does not hold for our electro-diffusion model except in the limit when the concentration gradients go to zero and Eq. (4) is used to compute the membrane currents. A compartment approximation for the solution of Eq. (14) is inaccurate for large ionic fluxes if the continuity constraint in Eq. (16) is not used to match solutions on the two sides of a diameter jump or at a branch point.

### B. Simulations of Postsynaptic Potentials in a Dendritic Spine

For simplicity, only three types of ions,  $K^+$ ,  $Na^+$  and  $Cl^-$ , will be considered in this chapter. Excitation was modeled by a combination of transient conductance changes to  $Na^+$  and  $K^+$ , with the  $K^+$  conductance equal to one-tenth of  $Na^+$  conductance (Hille, 1984). The synaptic reversal potential under this combination is about 50 mV. We also made simulations with the reversal potential of the excitatory synapse equal to 0 mV and similar conclusions were obtained. Silent shunting inhibition and hyperpolarizing inhibition were modeled by transient  $Cl^-$

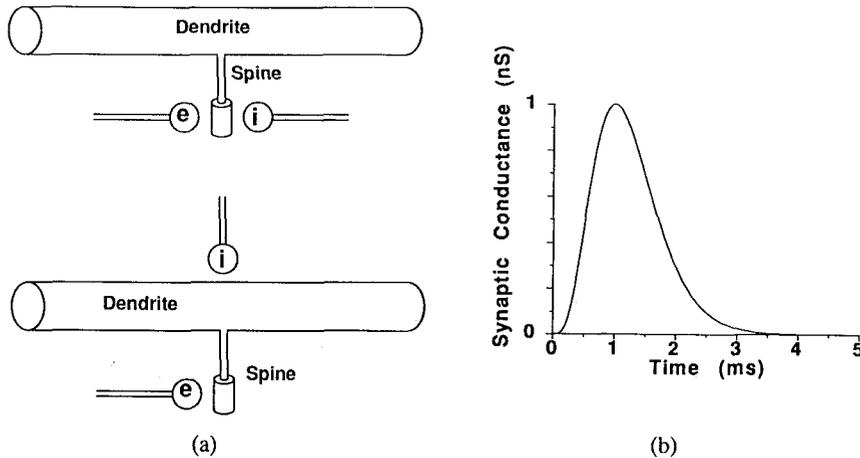


FIGURE 1. (a) Geometry of the dendrite and spine for the simulations showing excitation and inhibition on the spine head (top) and inhibition on the dendritic shaft at the base of the spine (bottom). The spine was located in the center of a dendrite with total length  $300 \mu\text{m}$  and diameter  $1 \mu\text{m}$ ; the spine neck was  $1 \mu\text{m}$  long and  $0.1 \mu\text{m}$  in diameter; the spine head was  $0.69 \mu\text{m}$  long and  $0.3 \mu\text{m}$  in diameter. In the simulations of the electrodiffusion model, sample points in the dendrite were  $10 \mu\text{m}$  apart and the integration time step was  $10^{-7}$  sec; in the spine head and neck, the spacing was  $0.173 \mu\text{m}$  and  $0.167 \mu\text{m}$ , respectively, and the time steps were  $2 \times 10^{-9}$  sec. The model had a total of 41 sample points: 31 in the dendrite, six in the spine neck, and four in the spine head. In the conventional cable model, only 33 lumped compartments were used (one for head, one for neck, and 31 for dendrite) due to the large spatial constant. The time step for spine head and neck was  $10^{-7}$  sec and that for the dendrite was  $10^{-6}$  sec. (b) Excitatory and inhibitory synaptic conductance changes were modeled by  $G(t) = G_M (et/t_{\text{peak}})^4 e^{-4t/t_{\text{peak}}}$ , where  $t_{\text{peak}}$  was the time to reach the peak conductance,  $G_M$ . A graph of this expression is shown with  $t_{\text{peak}} = 1$  ms, and  $G_M = 1$  nS. Parameters used in our simulations were:  $t_{\text{peak}} = 1$  ms; membrane capacitance  $C_m = 1 \mu\text{F}/\text{cm}^2$ ; diffusion coefficients  $D_K = 1.96 \times 10^{-5} \text{cm}^2/\text{sec}$ ,  $D_{\text{Na},\text{M}} = 1.33 \times 10^{-5} \text{cm}^2/\text{sec}$ , and  $D_{\text{Cl}} = 2.03 \times 10^{-5} \text{cm}^2/\text{sec}$ ; resting membrane conductances of unit area  $g_{\text{K},\text{rest}} = 1.95 \times 10^{-4} \text{Scm}^2$ ,  $g_{\text{Na},\text{rest}} = 1.63 \times 10^{-5} \text{Scm}^2$ , and  $g_{\text{Cl},\text{rest}} = 3.89 \times 10^{-5} \text{Scm}^2$ ; initial intracellular concentrations  $n_{\text{K},0} = 140$  mM,  $n_{\text{Na},0} = 12$  mM, and  $n_{\text{Cl},0} = 5.5$  mM; extracellular concentrations  $n_{\text{K},\text{out}} = 4$  mM,  $n_{\text{Na},\text{out}} = 145$  mM, and  $n_{\text{Cl},\text{out}} = 120$  mM. With this set of parameters, the resting membrane potential was  $-78$  mV. The Nernst potentials for  $\text{K}^+$ ,  $\text{Na}^+$ , and  $\text{Cl}^-$  were  $-90$  mV,  $63$  mV, and  $-78$  mV, respectively. Total membrane resistivity  $R_m$  was  $4000 \Omega\text{cm}^2$ . Total cytoplasmic resistivity  $R_i$  at rest was calculated (Qian and Sejnowski, 1989) to be  $87 \Omega\text{cm}$ . Total surface area of the spine head was  $0.65 \mu\text{m}^2$ . Some of these parameters were varied, as explained in the relevant figures or tables. The sources for these parameters are given in Qian and Sejnowski (1989).

and  $K^+$  conductance changes, respectively. Ionic driving forces similar to Eqs. (1) and (2) were used for the electro-diffusion model rather than the constant-field approximation used in Qian and Sejnowski (1989). The Nernst Potentials were updated at each time step according to the instantaneous ionic concentrations. We varied the magnitudes and durations of the conductance changes and the spine neck dimensions. We also compared the effectiveness of inhibition on the spine with inhibitory input on the dendrite at the base of the spine. The standard parameters used and details of the simulations are summarized in the caption of Fig. 1; any variation will be explicitly mentioned.

A measure of the effectiveness of shunting inhibition is the ratio of the maximum depolarization at a reference point in the neuron caused by an excitatory input alone to the depolarization when both the excitatory and the inhibitory inputs are present. This ratio, called the *F factor* (Koch and Poggio, 1983a), is equal to 1 if the inhibition has no effect on the excitation. One obvious requirement for effective inhibition is that its time course should overlap substantially with the excitatory synaptic conductance change. We modeled a spine located in the middle of a 300  $\mu\text{m}$ -long dendrites and the response at the spine head was used to calculate *F* factors. Our simulation results based on both the cable model and the electro-diffusion model are shown in Table I. The cable model indeed showed strong veto effects, especially when the conductances were large, as predicted. However, our electro-diffusion model showed no significant veto effect over a wide range of conductances. Note also that when the  $G_{\text{Cl},M}/G_{\text{Na},M}$  ratio was increased, there were cases where the *F* factor decreased slightly. This occurred because the  $\text{Cl}^-$  Nernst potential shifted so much that it depolarized the membrane away from its resting level. A  $\text{Cl}^-$  conductance change alone, however, did not cause any depolarization because there was no driving force and, therefore, no concentration change.

TABLE I. *F* Factors at the Spine Head when Both Excitatory and  $\text{Cl}^-$  Mediated Inhibitory Synaptic Inputs are Located on the Same Spine Predicted by Both the Electro-Diffusion Model and the Cable Model.

$G_{\text{Cl},M}/G_{\text{Na},M}$	$G_{\text{Na},M} = 0.1 \text{ nS}$		$G_{\text{Na},M} = 1.0 \text{ nS}$		$G_{\text{Na},M} = 10 \text{ nS}$	
	diffusion	cable	diffusion	cable	diffusion	cable
1	1.02	1.02	1.10	1.17	1.23	1.65
10	1.10	1.20	1.20	2.74	1.26	7.56
$10^2$	1.16	3.04	1.19	18.63	1.25	66.20
$10^3$	1.14	20.35	1.19	163.86	1.25	602.19

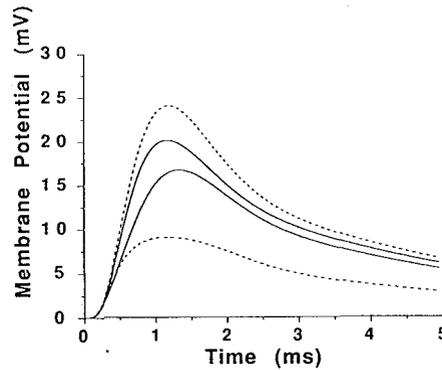


FIGURE 2. Postsynaptic responses relative to the resting level at the spine head calculated with the electro-diffusion model (solid lines) and the cable model (dotted lines). Two traces are shown for each model: The top trace is the response with excitatory synaptic input alone and the lower trace is the response to both excitatory and shunting inhibitory inputs. Excitatory synaptic input:  $G_{Na,M} = 1$  nS; inhibitory synaptic input:  $G_{Cl,M} = 10$  nS.

The difference between the cable model and the electro-diffusion model decreased as the synaptic conductances decreased. For  $G_{Na,M} = 0.1$  nS, the two models were essentially identical and both predicted that the inhibition was ineffective. However, for longer durations of the synaptic input, the two models may not agree even for synaptic conductance changes as small as 0.1 nS. (See Section VI.) The details of the postsynaptic responses on the spine head are shown in Fig. 2.

Quantal analysis on excitatory postsynaptic potentials in area CA3 of the rat hippocampus (Higashima *et al.*, 1986; Brown *et al.*, 1988) gave a quantal conductance of about 1 nS at mossy fiber synapses. Therefore, the synaptic conductance change due to a single presynaptic action potential should be about a few nS. Similar measurements of unitary inhibitory conductance performed on CA3 pyramidal cells of guinea-pig hippocampus obtained a value of 5–9 nS (Miles and Wong, 1984). However, the conductance of synapses on pyramidal cells in cerebral cortex may be much smaller. In the following simulations, we fixed  $G_{Na,M}$  at 1 nS and varied  $G_{Cl,M}$  unless otherwise indicated.

The morphologies of spines vary greatly. The critical parameters for our simulations were the diameter and length of spine neck, which were varied from 0.1  $\mu\text{m}$  to 0.25  $\mu\text{m}$  and 0.4  $\mu\text{m}$  to 1.0  $\mu\text{m}$ , respectively, with the neck membrane area kept constant. We also considered the case where there was no spine neck and the spine head was connected directly to dendrite. The cable model gave large  $F$  factors when the neck was long and narrow and/or  $G_{Cl,M}$  was large, but

the electro-diffusion model produced no  $F$  factor larger than 2 over the entire range. The effectiveness of inhibition was not very sensitive to the dimensions of the spine neck because of two competing effects that cancel: As the spine neck length was decreased and the diameter increased, the concentration changes in the spine were reduced, making the inhibition more effective. However, the input resistance of the spine head was also decreased, resulting in a smaller depolarization and a reduced driving force for the inhibition.

**1. On-Path Inhibition.** Inhibitory synapses on pyramidal neurons are commonly located on the dendritic shaft at the base of the spine (Beaulieu and Colonnier, 1985; Martin, 1984). The simulations in Fig. 3 show that dendritic on-path inhibition is much more effective than inhibition on the spine head. The ionic concentration changes were much smaller for the dendritic inhibition because the dendrite had a diameter of  $1 \mu\text{m}$ , and hence the cable equation was a good approximation. Also, the driving force for the inhibition was strong because the spine was electrically coupled to the dendrite well enough that the excitation of the spine caused a large depolarization at the dendritic shaft.

**2. Duration of Inhibition.** Table II shows the  $F$  factors at the spine for a range of durations of the synaptic conductance (determined by  $t_{\text{peak}}$  in Fig. 1). A longer duration produced a larger membrane depolarization that increased the driving force for the inhibition and made the inhibition more effective. This

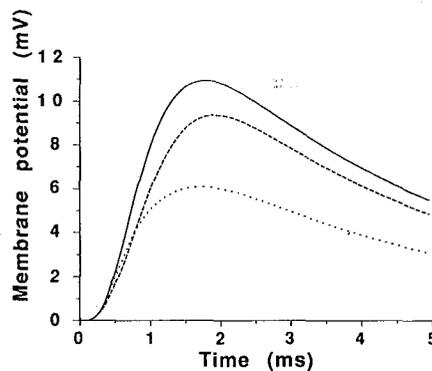


FIGURE 3. Responses relative to the resting level at the dendritic shaft at the base of a spine under three conditions: no inhibition and the excitatory input ( $G_{\text{Na},\text{M}} = 1 \text{ nS}$ ) on the spine head alone (solid), the excitatory input ( $G_{\text{Na},\text{M}} = 1 \text{ nS}$ ) and the inhibitory input ( $G_{\text{Cl},\text{M}} = 10 \text{ nS}$ ) both on the same spine head (dashed), and the excitatory input ( $G_{\text{Na},\text{M}} = 1 \text{ nS}$ ) on the spine head and the inhibitory input ( $G_{\text{Cl},\text{M}} = 10 \text{ nS}$ ) on the dendritic shaft at the base of the spine (dotted).

TABLE II.  $F$  Factors at the Spine for a Range of Durations of the Synaptic Conductance. The time to peak ( $t_{\text{peak}}$ , explained in caption to Fig. 1a) was varied for a synapse with  $G_{\text{Na,M}} = 1 \text{ nS}$  and  $G_{\text{Cl,M}} = 11 \text{ nS}$ .

$t_{\text{peak}}(\text{ms})$	diffusion	cable
0.5	1.32	2.46
1.0	1.20	2.73
2.0	1.17	3.01
3.0	1.16	3.30
4.0	1.17	3.53

explains the increase of  $F$  factors with  $t_{\text{peak}}$  predicted by the cable model. However, the longer the conductance change, the larger the concentration change, which makes inhibition less effective according to the electro-diffusion model. The latter factor predominated for inhibition on the spine head, as shown in Table II.

Repetitive stimulation of inhibitory interneurons in hippocampal slices for tens of seconds can cause disinhibition of inhibitory potentials in pyramidal cells (Thompson and Gahwiler, 1989). The most likely explanation is the intracellular accumulation of  $\text{Cl}^-$ . Our simulations suggest that a similar disinhibition can occur on spines within milliseconds. The difference between the time scales can be attributed to the difference between the intracellular volumes of cell bodies and dendrites compared with spine heads.

**3. Interactions between Synapses on Dendrites.** We next studied interactions between excitatory and inhibitory synapses at adjacent sites on dendrites ranging in diameter from 0.1 to 2.0  $\mu\text{m}$ . The predicted  $F$  factors for the cable model, given in Table III, were very large when the dendritic diameter was small and the  $G_{\text{Cl}}$  was large. For the electro-diffusion model, two competing factors determined the effectiveness of inhibition: For dendrites with large diameters, the concentration effects were small, so the Nernst potential did not change very much and the inhibition was effective. However, the polarization of the membrane from the resting level was smaller in larger dendrites, which made inhibition less effective. The  $F$  factors in Table IIIa were not monotonically increasing with increasing dendritic diameter because of these two factors and their interaction. For  $G_i = G_e$ , the first factor dominated and the inhibition was comparatively more effective on small dendrites. When  $G_i \geq 100 G_e$ , the second factor dominated and the inhibition was more effective on large dendrites. In any case,

TABLE III.  $F$  Factors for Both Excitatory and Inhibitory Synapses on Dendrites at the Same Site, with Different Dendritic Diameters.  $G_{Na,M} = 1$  nS. Top: Electro-Diffusion Model; Bottom: Cable Model.

$G_{Cl,M}/G_{Na,M}$	dendritic diameter ( $\mu\text{m}$ )				
	0.1	0.25	0.5	1.0	2.0
0.1	1.07	1.04	1.02	1.01	1.00
1	1.47	1.45	1.18	1.07	1.03
10	1.87	1.66	2.26	1.66	1.31
$10^2$	1.91	2.67	3.43	3.10	3.25

$G_{Cl,M}/G_{Na,M}$	dendritic diameter ( $\mu\text{m}$ )				
	0.1	0.25	0.5	1.0	2.0
0.1	1.08	1.06	1.04	1.02	1.01
1	1.72	1.39	1.19	1.08	1.04
10	8.31	5.16	3.03	1.88	1.38
$10^2$	73.07	43.15	22.46	11.01	5.65

when the dendritic diameter was  $0.1 \mu\text{m}$ , the  $F$  factors were always less than 2, similar to the previous results for inhibition on spines.

**4.  $K^+$ -Mediated Inhibition.** The equilibrium potential for  $K^+$  is generally below the resting membrane potential (12 mV below in our model), so that an increase in  $K^+$  conductance leads to a hyperpolarization. In a previous study, we found that inhibition on spines mediated by  $K^+$  was not effective for excitatory conductances greater than 10 nS (Qian and Sejnowski, 1989). In this section, we consider excitatory conductances that are lower and more realistic for pyramidal neurons. We find that for smaller excitatory conductances, hyperpolarizing inhibition can be quite effective. Synaptic responses to  $G_{Na,M} = 0.1$  nS are shown in Fig. 4, which also shows that an inhibition of  $G_{K,M} = 1$  nS is very effective in reducing the response. In comparison, the inhibition due to a similar or much larger conductance change for  $Cl^-$  was not effective. For large excitatory conductance changes, the  $K^+$  inhibition became as ineffective as  $Cl^-$  because of the large  $K^+$  concentration changes that rapidly shift the  $K^+$  Nernst potential, as shown in Table IV.

Inhibition mediated by  $K^+$  in cortical neurons has a time course that can last for a significant fraction of a second when it is activated by  $GABA_B$  receptors

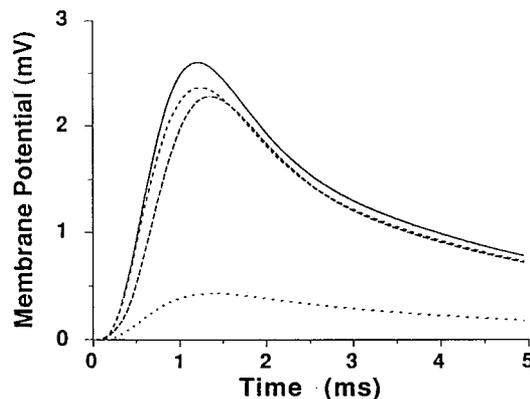


FIGURE 4. Response relative to the resting level at the spine head to an excitatory input of  $G_{Na,M} = 0.1$  nS and one of the following four different inhibitory synaptic inputs: no inhibition (solid),  $K^+$  inhibitory synaptic input with  $G_{K,M} = 1$  nS (dotted),  $Cl^-$  inhibitory input with  $G_{Cl,M} = 1$  nS (dashed), and  $G_{Cl,M} = 100$  nS (long-dashed).

through G proteins. We therefore studied the steady-state behavior following a step change in conductances of  $G_{Na} = 0.1$  nS and  $G_K = 1$  nS and found that the response at the spine head was about 6.9 mV with excitation alone and 1.3 mV with both excitation and inhibition. In steady state, the  $K^+$  efflux from the spine head was balanced by the  $K^+$  diffusion from the dendritic shaft to the head. Thus, the inhibition mediated by  $K^+$  conductances remained effective for slow inhibitory synaptic potentials when the excitatory conductances were small. Since an excitatory synaptic conductance typically lasts for only a few ms, an excitatory input arriving a few ms earlier than the inhibitory input will not be affected by the inhibition. Once an inhibitory input is active, there is a long time window during which arriving excitatory inputs are inhibited.

TABLE IV.  $F$  Factors at the Spine Head when Both Excitatory and  $K^+$  Mediated Inhibitory Synaptic Inputs are Located on the Same Spine Predicted by both the Electro-Diffusion and the Cable Models.

$G_{K,M}/G_{Na,M}$	$G_{Na,M} = 0.1$ nS		$G_{Na,M} = 1.0$ nS		$G_{Na,M} = 10$ nS	
	diffusion	cable	diffusion	cable	diffusion	cable
0.1	1.01	1.07	1.02	1.07	1.07	1.07
1	1.11	1.18	1.24	1.33	1.58	1.79
10	6.05	8.03	7.35	17.65	1.70	47.53
$10^2$	*	*	*	*	1.66	*

\* $F$  factors undefined because the responses were hyperpolarizing, indicating very effective inhibition.

### V. The Cable Model for Electro-Diffusion

The electro-diffusion model is highly computation-intensive and cannot be used routinely for large-scale simulations of complex dendritic trees (Wathey *et al.* 1989). It would be desirable to have a model that was as efficient as the cable model. We will prove here that a simple extension of the cable model can, in fact, provide an accurate approximation to the predictions of the electro-diffusion model.

The following modifications to the discrete approximation of the standard cable model should be made at each time step:

- (1) Calculate the intracellular concentration of each ionic species explicitly in each compartment from the membrane currents and the ionic currents flowing between compartments.
- (2) Compute the new membrane equilibrium potentials for each compartment using the intracellular and extracellular ionic concentrations according to the Nernst equation and update the membrane batteries:

$$E_k = \frac{RT}{Fz_k} \ln \frac{n_k(\text{out})}{n_k(\text{in})}, \quad (17)$$

where  $n_k(\text{out})$  is the ionic concentration of species  $k$  outside the membrane and  $n_k(\text{in})$  is the ionic concentration inside the membrane. This makes the membrane current expressions identical to those used in the electro-diffusion model.

- (3) Replace the single longitudinal resistance between compartments with parallel resistances  $R_{i,k}$  (Qian and Sejnowski, 1989).

$$\frac{1}{R_{i,k}} = (F^2/RT)D_k n_k z_k^2, \quad (18)$$

in series with batteries (step 4) for each ionic species.

- (4) Determine the longitudinal batteries by the Nernst potential for ionic concentrations of the two compartments they connect and update in the same way as the membrane batteries. The potential of the battery for species  $k$  between compartments  $j$  and  $j + 1$  is:

$$E_{i,k} = \frac{RT}{Fz_k} \ln \frac{n_k(j)}{n_k(j+1)}, \quad (19)$$

where a positive value for  $E_{i,k}$  means that the positive terminal of the battery is pointing to the  $j + 1$  compartment.

Steps 3 and 4 account for the effects of the longitudinal component of electrodiffusion in Eq. (11). We show here that this approximation is exact in limit that the compartments approach zero length.

Consider two compartments at  $x$  and  $x + dx$ . The longitudinal battery for ionic species  $k$  between these two compartments, according to step 4, is:

$$dE_{i,k}(x) = \frac{RT}{Fz_k} \ln \frac{n_k(x)}{n_k(x+dx)} = -\frac{RT}{Fz_k} \ln \left[ 1 + \frac{1}{n_k} \frac{dn_k}{dx} dx \right] = -\frac{RT}{Fz_k} \frac{dn_k}{n_k} dx.$$

For a nerve fiber of cross-sectional area  $A$ , the ionic current of species  $k$  between these two compartments is:

$$I_k(x) = \frac{dE_{i,k}(x) + V(x) - V(x+dx)}{R_{i,k} dx/A} = -AFz_k D_k \frac{dn_k}{dx} - \frac{AFz_k D_k n_k}{\alpha_k} \frac{dV}{dx}. \quad (20)$$

Equation (18) and the definition of  $\alpha_k$  in Eq. (12) were used in this derivation. The flux  $J_k(x)$  is defined as the number of ions moving across a unit area in unit time, of ionic species  $k$  between the two compartments. Thus, from Eq. (20),

$$J_k(x) = \frac{I_k(x)}{AFz_k} = -D_k \left[ \frac{dn_k}{dx} + \frac{n_k}{\alpha_k} \frac{dV}{dx} \right], \quad (21)$$

#### Modified Cable Model

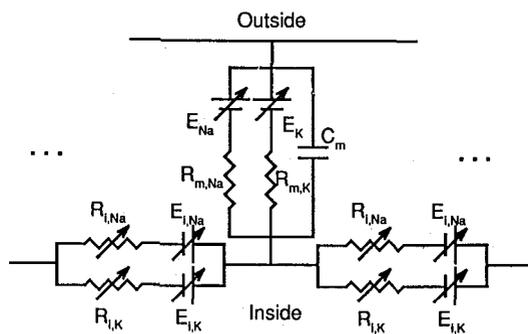


FIGURE 5. Electrical circuit for a single compartment of the modified cable model. The batteries in the membrane and between compartments are calculated from the Nernst potentials and change value during a response as concentrations change. The resistivities in the cytoplasm are also updated at each time step.

which is identical to the longitudinal component of the Nernst-Planck electro-diffusion equation for ionic species  $k$ . The continuity equation of the electro-diffusion model (Eq. (13)) is also satisfied in cable model because of Kirchoff's law. Therefore, results predicted by the modified cable model should be equivalent to those of the one-dimensional electro-diffusion model.

A schematic view of the modified cable model is represented in Fig. 5. This model was applied to the dendritic spine model in Qian and Sejnowski (1989). The results were in excellent agreement with the electro-diffusion model except when the conductances and ionic concentrations changed very rapidly. The differences were mainly due to the fact that we used the constant-field approximation for the membrane currents in that version of the electro-diffusion model, but used ohm's law for the batteries in the modified cable model.

## VI. Discussion

In most circumstances, the cable model of electrical conduction in neurons gives accurate predictions for membrane potentials during transient electrical events. A tiny amount of charge is enough to cause a substantial change in the membrane potential because the membrane capacitance is small, and as a consequence, the ionic concentrations usually are nearly constant (Qian and Sejnowski, 1989). However, the *individual* concentrations of certain ions could change significantly as long as changes in the total charge are nearly balanced out. This is more likely to happen in very small structures such as dendritic spines and for an ion such as  $\text{Ca}^{++}$ , which is normally maintained at a very low concentration inside a cell. Thus, the predictions from the cable model should not be used without a careful analysis.

We have developed a one-dimensional electro-diffusion model of electrical conduction, which reduces to the cable model when ionic concentrations are approximately constant. This model was used to study changes in ion concentrations and membrane potentials in dendritic spines in response to synaptic inputs. We found that ionic concentrations changed considerably in many circumstances. Thus, significant errors can be made in estimating membrane potentials and concentration changes using the cable model if the effects of diffusion and the changes in the driving forces for membrane current are not taken into account.

The major conclusion of this study is that  $\text{Cl}^-$  shunting inhibition on spines cannot be very effective regardless of how large the synaptic conductance changes are. Shunting inhibition is significantly more effective when it is on the dendritic shaft on-path to the cell body. This may partly explain the anatomical findings

that most synapses on spines are putatively excitatory and that the majority of the putative inhibitory synapses are found on dendritic shafts. Shunting inhibitory synapses on spines may have other functions. Although they may not contribute significantly to the electrical responses of the cell, they can certainly cause large local ionic concentration changes that may be important in regulating certain cellular functions.

The inhibitory synapses on spines may contribute to the electrical responsiveness of a cell if they are mediated through  $K^+$  currents. Our simulations predict that  $K^+$  hyperpolarizing inhibition on a spine head can be very effective when the excitatory synaptic conductance changes are less than 10 nS. The major inhibitory neurotransmitter in the visual cortex is GABA;  $GABA_A$  receptors are coupled to  $Cl^-$  channels and  $GABA_B$  receptors are linked to  $K^+$  channels. Therefore, we specifically predict that the inhibition on spines is mediated by the  $GABA_B$  receptors. This prediction is consistent with the finding that  $GABA_B$  input to hippocampal pyramidal cells is preferentially dendritic (Janigro and Schwartzkroin, 1988), where the majority of inputs are onto spines. Another way to have effective inhibition on a spine is through conductance decreases of either  $Na^+$  or  $Ca^{++}$ , although this type of inhibitory mechanism has not been found in cortical neurons.

All these results can be understood as a consequence of changing ionic concentrations and shifting equilibrium potentials. The postsynaptic responses are mainly determined by the ionic species with the largest transient membrane permeability. Regardless of the equilibrium potential for this species (excitatory or inhibitory), the opening of channels with sufficient duration will shift the equilibrium potential of the synapse toward zero because of the large concentration changes. Therefore, if the equilibrium potential of the ionic species is greater than zero ( $Na^+$  and  $Ca^{++}$ ), the cable model will overestimate the response. For ions whose equilibrium potential is less than zero ( $K^+$  and  $Cl^-$ ), the cable model makes predictions that can be qualitatively incorrect. For example, under certain conditions an inhibitory synaptic input that causes a large  $K^+$  conductance increase may, in fact, produce a postsynaptic depolarization.

Our simulations have shown that discrepancies between the cable model and our electro-diffusion model increase with the magnitude and duration of the synaptic conductance changes. Since the cable model is valid only when the concentration changes are small, we can derive a condition under which the cable model is self-consistent. The intracellular concentration change of the  $k$ th ionic species caused by membrane current  $I_k$  within time duration  $\Delta t$  is  $\Delta n_k = I_k \Delta t / vzF$ , where  $v$  is the effective intracellular volume,  $z$  is the valence of the ion involved, and  $F$  is the Faraday's constant. (Of course,  $n_k$  will eventually stop changing with time when the membrane current is balanced by the intra-

cellular diffusion.) The criteria for the self-consistency of the cable model is simply  $|\Delta n_k|/n_{k,0} \ll 1$ , where  $n_{k,0}$  is the initial intracellular ionic concentration, or  $\Delta t \ll |vzFn_{k,0}/I_k|$ .

When the synaptic conductance  $G_k$  is small, say, 0.1 nS,  $I_k \sim G_k E_k$ , where  $E_k$  is the reversal potential relative to the resting potential. The preceding condition gives  $\Delta t \ll 10$  ms for  $\text{Na}^+$  in a spine, assuming that the effective volume  $v$  is equal to twice the volume of the spine. Therefore, the cable model may not be valid for spines if the duration of the conductance change is longer than 10 ms, even for synaptic conductance changes as small as 0.1 nS. The inclusion of ionic pumps would not alter the preceding conclusions for a typical Na-K pump current density of  $1 \sim \mu\text{A}/\text{cm}^2$  (Weer and Rakowski, 1984), in which case the total pump current of the spine head is about  $10^{-14}$  A, three orders of magnitude smaller than the synaptic current. Even when the pump molecules are close-packed in the membrane, the maximum possible pump current density is  $100 \mu\text{A}/\text{cm}^2$ , and the total pump current of the spine head is still about 10 times smaller than the synaptic current for a 0.1 nS conductance. The effect of the Na-K pump would be significant if we assume that the spine apparatus is also densely packed with pump molecules and its surface area is about 10 times that of the spine.

Ionic concentration changes are usually not explicitly considered in the cable model. Although ionic currents in the cable model can be integrated to yield concentration changes, this usually gives erroneous results (Qian and Sejnowski, 1989), even when the membrane potentials are predicted fairly well. Often, the cable model is solved first to find the membrane potentials and then diffusion processes are introduced to determine the ionic concentration changes (Gamble and Koch, 1987; Yamada *et al.*, 1989). Our model, however, considers the membrane potential and the ionic concentration changes at the same time and thus provides a more natural and accurate way for solving the problem.

In our simulations, we have assumed that extracellular ionic concentrations were constant to simplify our calculations. This may not be a valid approximation in restricted extracellular spaces for the same reasons that the cable model broke down in restricted intracellular spaces. If the extracellular space around a spine head that is effectively available for exchange in 0.5 ms is about the same as the volume of spine, then a change of concentration in the spine head would cause an equal change with opposite sign outside the spine head. For  $t_p = 0.25$  ms, the extracellular  $[\text{Na}^+]$  would change from 140 mM to about 110 mM, and  $[\text{K}^+]$  from 4 mM to about 34 mM in about 0.5 ms. (See Yamada *et al.*, 1989 for a similar estimate.) Although the extracellular  $\text{K}^+$  concentration would increase by a larger factor, the maximum value of an excitatory synaptic response

is mainly determined by the Nernst potential of  $\text{Na}^+$  because the  $\text{Na}^+$  permeability is much larger during an excitatory synaptic input. Also, glial cells are very effective in maintaining  $\text{K}^+$  homeostasis on a longer time scale so that the actual change during maintained activity is probably less.

Thus, the main effect of a limited extracellular space is on the Nernst potential of  $\text{Na}^+$ . Based on the preceding estimates, this would reduce the peak response of the postsynaptic potential by about 6% at the spine head. For large  $t_p$  and multiple synaptic inputs, the  $\text{Na}^+$  concentration change is greater but is achieved over a longer period of time. The corresponding effective extracellular space around the spine head would then be larger because more time allows ions to diffuse further. Thus, the modification would not be much greater. For an excitatory input driven by a  $\text{Na}^+$  current, the effects of restricted extracellular space always reduce the amplitude of response and thus will tend to make the differences with the cable model even greater.

For synapses on large dendrites rather than on spines, we do not expect any significant difference between the electro-diffusion model and the cable model because the ionic concentration changes are negligible. Synaptic inputs on thick dendrites should not suffer the saturation caused by shifts of the Nernst potential, the absence of temporal summation, and lack of an inhibitory veto effect that we have demonstrated for synapses on spines. On the other hand, large compartments are also harder to depolarize, which is needed to increase the driving force of the inhibitory currents. This second factor may not be as important if a compartment also receives a large number of convergent excitatory synapses. Note that the more depolarized a cell, the more effective the inhibitory synapses become and the less effective the excitatory synapses. Indeed, if the depolarization is large enough to trigger an action potential, the driving forces for the inhibitory currents on soma and dendrites reach their maximum and the driving forces for the excitatory currents reach their minimum, especially if the effects of the action potentials propagate up the dendritic tree. Thus, inhibitory synapses on the cell body and proximal dendrites could control the effects of action potentials propagating up dendritic trees and the temporal firing patterns of the neuron (Lytton and Sejnowski, 1991).

This raises the interesting possibility that otherwise identical synapses could have different functions depending on their location. It has been reported that during learning and development, the spine neck shortens and merges into the dendrite (Rausch and Scheich, 1982; Coss and Globus, 1978; Brandon and Coss, 1982; Coss *et al.*, 1980). The concentration changes in spines are caused not just by the small volume of the spine head but also by the long, narrow spine neck, which helps to maintain the large concentration changes that occur in the spine head. Thus, synapses on spines with long necks could switch to a different

functional state if the neck were to shorten sufficiently for the spine to merge with the dendritic shaft.

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