Well-Timed, Brief Inhibition Can Promote Spiking: Postinhibitory Facilitation

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Submitted 18 July 2005; accepted in final form 10 January 2006

INTRODUCTION

From a classical perspective one role for synaptic inhibition is as modulation, biasing a cell’s membrane potential away from threshold. In consideration of transient effects, an inhibitory input could veto a brief suprathreshold excitatory input if the inhibition is delivered coincidentally or just before the excitation (see, e.g., Segev and Rall 1988). Awareness is growing that transient inhibition can also play a role in spike timing and phase locking of a neuron (Gauck and Jaeger 2000; Lytton and Sejnowski 1991; Perkel et al. 1964) or in network rhythrogenesis (Bartos et al. 2002; Wang and Buzsáki 1996; Whittington et al. 1995) or excitability of cells (Granit 1956; Kuffler and Eyzaguirre 1955). Even though the inhibition might affect spike timing, the overall activity in many such cases might be reduced. On the other hand, inhibition can sometimes induce firing. In classical postinhibitory rebound (PIR), sustained hyperpolarization as from a long inhibitory barrage sets the stage for a rebound spike after release (Granit 1956; Kuffler and Eyzaguirre 1955). Here we show that very brief, individual or tightly synchronized, inhibitory events can surprisingly also facilitate spike generation. We call this phenomenon postinhibitory facilitation (PIF). In PIF brief inhibition if delivered in an appropriate time window can facilitate firing in response to a subsequent, brief subthreshold excitatory input. We illustrate and analyze such behaviors in the context of auditory brain stem neurons and models, where the underlying biophysical mechanism is quite prominent and the timescales are strikingly fast. The facilitating effect of transient and well-timed inhibition on subthreshold excitation is a general phenomenon. PIF is likely used by many neuronal systems both at the cell and circuit levels; the inhibitory–excitatory pairings can occur as well in paired inputs, in periodic or even random trains. The requisite timescales of inhibition and the PIF window depend on the system’s intrinsic dynamics below threshold.

Recent in vitro (Grothe and Sanes 1994) and in vivo (Brand et al. 2002; Grothe 2003) studies have shown that precisely timed fast inhibition significantly affects the sensitivity of auditory brain stem neurons to detect bilateral excitatory input coincidences. Here, we demonstrate the essentials of PIF using neurons of the medial superior olive (MSO) and a conductance-based model of them. The biophysical components of MSO neurons seem well tuned for speed and precision and for the coincidence detection that they carry out in the localization of sound sources (Goldberg and Brown 1969; Spitzer and Semple 1995; Yin and Chan 1990). The intrinsic voltage-dependent conductances and synaptic currents show rapid gating (Gardner et al. 1999; Magnusson et al. 2005; Rathouz and Trussell 1998; Trussell 1999). These and other cells in the auditory brain stem have a low-threshold, dynamic, potassium conductance $g_{KLT}$ that is partially activated at rest. The gating properties of $g_{KLT}$ contribute to the cells’ temporal processing abilities (e.g., see Reyes et al., 1996; Rothman and Manis 2003; Svirskis and Rinzel 2003; Svirskis et al. 2002, 2003). We show here that $g_{KLT}$ also underlies PIF. The response to a normally subthreshold excitatory input is enhanced by a brief inhibition, $g_{inh}(t)$, enough to cause the cell to spike, provided the inhibition occurs in a favorable time window ahead of the excitation. The hyperpolarization induced by $g_{inh}$ transiently reduces $g_{KLT}$ and defines the time window for PIF. During this time window the cell is hyperexcitable: after $g_{inh}$ is virtually complete and while $V$ is returning to rest, but not too hyper-
polarized, and before $g_{KLT}$ recovers. This hyperexcitability allows a fast excitatory event that otherwise would be subthreshold, to bring the cell above threshold. If the time course of $g_{inh}(t)$ (with fixed amplitude) is too fast or too slow, relative to the intrinsic timescales of $V$ and the activation/deactivation rate of $g_{KLT}$, PIF is precluded. The decay rates that are sufficient for PIF in this system are sub- to a few milliseconds; for other systems they could be longer but not much longer than the suppressive factor that must be transiently relieved to create PIF.

Many of our results are demonstrated for a specific conductance-based model (Rothman et al. 1993). However, the geometrical analysis and underlying mathematical structure of our reduced two-variable version of the model suggest that features of PIF will be found in a class of models. The class is not necessarily auditory related and it includes the standard Hodgkin–Huxley (HH) model. The reduction also enables us to show that at one extreme even an adequate delta pulse of strong and fast inhibition can lead to PIF and, in some parameter regimes, even rebound spike generation without an excitatory input. This bears significant implications for systems with strong and fast inhibition. Interestingly, such a neuron or model has two thresholds for brief inputs, one for hyperpolarizing and one for depolarizing input. We illustrate the general principles of PIF further by formulating and analyzing a leaky integrate-and-fire (LIF) model that includes an accommodating dynamic threshold $\theta$. Hyperexcitability arises in this minimal model because $\theta$ decreases dynamically with hyperpolarization. Qualitatively, PIF can occur here when inhibition is faster than $V$ and $\theta$, regardless of the absolute timescales.

The effectiveness of brief inhibitory inputs in facilitating spike generation and the sensitivity to timing between excitation and inhibition emphasizes that we should be considering transient inhibition as an important component of temporal processing, well beyond just modulation of a cell's membrane potential or vetoing a would-be spike.

**METHODS**

**Computational**

The model incorporates HH-like sodium, delayed rectifier potassium, and leak currents as well as a low-threshold potassium current ($I_{KLT}$). The current balance equation is

$$C(dV/dt) = I_{app} - g_{Na}(V-E_{Na}) - g_{K}(V-E_{K}) - g_{KLT}(V-E_{KLT}) - g_{exo}(V-E_{exo}) - I_{inh}$$

where $g_{Na} = G_{Na}m^2h$, $g_{K} = G_{K}h$, and $g_{KLT} = G_{KLT}w$. This model was developed for bushy cells in the auditory brain stem (Rothman et al. 1993) and was also applied in conjunction with an in vivo experimental study on the role of precise inhibition in interaural time difference sensitivity in MSO neurons (Brand et al. 2002). $C$ is the membrane capacitance ($=23 \text{ pF}$). The reversal potentials of the various currents are $E_{Na} = 55 \text{ mV}$, $E_{K} = E_{KLT} = -77 \text{ mV}$, $E_{exo} = 2.8 \text{ mV}$, and the maximal conductance values are $G_{Na} = 985.2 \text{ nS}$, $G_{K} = 173.3 \text{ nS}$, $G_{KLT} = 86.6 \text{ nS}$, and $G_{exo} = 5.15 \text{ nS}$. The cell's resting potential is $-60 \text{ mV}$. The activation gating variables $m, n, w$, and the inactivation variable $h$ evolve according to $dx/dt = \alpha_{x}(1-x) - \beta_{x}x$, $x = m, h, n$, and $w$. The rate constants $\alpha_{x}$ and $\beta_{x}$ are voltage-dependent functions that for $I_{KLT}$ are determined from voltage-clamp experiments performed on bushy cell membranes by Manis and Marx (1991) and for $I_{inh}$ and $I_{K}$ are taken from other experimentally based models.

We adopt the parameter dependency on temperature as in the original model (Rothman et al. 1993), here adjusted for 38°C. At this temperature the model does not show the classical PIR effect. The external applied current $I_{app}$ is set to zero, unless stated otherwise. The synaptic current is given by $I_{syn} = g_{inh}(t)(V-E_{inh}) + g_{inh}(t)(V-E_{inh})$ with $E_{ex} = -10 \text{ mV}$ and $E_{inh} = -66.5 \text{ mV}$. The value for $E_{inh}$ was also used in modeling bushy cells (Rothman et al. 1993); the fast inhibition in MSO was recently shown to be even more hyperpolarizing (Mannusson et al. 2005). $g_{exo}$ and $g_{inh}$ are alpha-function conductances given by

$$g_{exo}(t) = g_{exo}(t_0)/\tau_{exo}e^{-(t-t_0)/\tau_{exo}}H(t-t_0) \quad (2)$$

$$g_{inh}(t) = g_{inh}(t_0 + \delta)/\tau_{inh}e^{-(t-t_0-\delta)/\tau_{inh}}H(t-t_0 + \delta) \quad (3)$$

where $H(y) = 1$ if $y \geq 0$, and zero otherwise. Inhibition precedes the excitation by $\delta$ (>0) ms. Except for some cases in Fig. 1D we choose $g_{exo}$ to be subthreshold; $t_0$ is the initiation time of the excitatory synaptic event. In our simulations, the membrane is considered as having evoked an action potential if its voltage exceeds $-30 \text{ mV}$.

**Experimental**

Gerbils (*Meriones unguiculatus*) aged postnatal (P) days 17–20 were used to make 200-μm slices through the MSO. The artificial cerebrospinal fluid (ACSF) contained (in mM) 125 NaCl, 4 KCl, 1.2 KH$_2$PO$_4$, 1.3 MgSO$_4$, 26 NaHCO$_3$, 15 glucose, 2.4 CaCl$_2$, and 0.4 l-ascorbic acid (pH 7.3 when bubbled with 95% O$_2$-5% CO$_2$). The ACSF was continuously superfused in the recording chamber at 4–5 ml per min at 34°C. The internal patch solution contained (in mM) 127.5 potassium gluconate, 0.6 EGTA, 10 HEPES, 2 MgCl$_2$, 5 KCl, 2 ATP, 10 phosphocreatinine (Tris salt), and 0.3 GTP (pH 7.2). To block $I_{KLT}$ currents 8 nM of dendrotoxin I (DTXI) (Alomone Labs, Jerusalem, Israel) was added to the bathing solution. Data acquisition and stimulus generation were carried out under computer control at 10 kHz. We used the dynamic-clamp method for stimulus generation. The program determined in real time the time-varying current that was injected into the neuron according to the calculated conductance value and the measured instantaneous membrane potential (Reyes et al. 1996; Sharp et al. 1993). The reversal potential was 0 and $-70 \text{ mV}$ for excitation and inhibition, respectively. The decay time constant for the exponentially decaying “postsynaptic” conductance transients was 1 ms for excitation and inhibition.

**RESULTS**

We explore here subthreshold integration properties, in particular how brief inhibition can enhance membrane excitability and responsiveness to an individual subthreshold excitatory input. In the first two subsections we focus on models and neurons of auditory brain stem and the response to a paired input. In two later subsections we demonstrate the generality of the phenomena and mechanism by describing and analyzing the PIF behavior for a two-variable, reduced, conductance-based model and an LIF model with accommodation. In the final subsection, we show how favorably timed, periodic, fast inhibition can lead to spiking in response to a periodic train of subthreshold excitatory inputs.

**Selective time window for facilitation**

At the rest state ($V_{rest} \approx -60 \text{ mV}$) in the single-compartment, five-variable conductance-based model (Rothman et al. 1993; see METHODS) >80% of the outward current is attributed to the low-threshold potassium current, $I_{KLT}$; and its conductance $g_{KLT}$ is significantly activated (about 20%). Therefore a
transient reduction of this current arising from synaptic inhibition makes the cell hyperexcitable. Because the gating time scale for $g_{\text{KLT}}$ near rest is about 2 ms even a very brief inhibitory input can deactivate $g_{\text{KLT}}$. This hyperexcitability means that a brief excitatory input, which is below threshold when delivered alone, can cause the cell to spike if it is preceded in the proper time window by a fast priming inhibition.

To demonstrate the PIF effect we simulate the response to a brief subthreshold excitatory conductance (EPSG) that is preceded $\delta$ ms by an inhibitory conductance transient (IPSG) (Fig. 1). If the IPSG precedes the EPSG by too much (large $\delta$) then no spike occurs; the depolarization remains subthreshold, as if the EPSG were delivered to a resting membrane (Fig. 1A). If the IPSG occurs too near to the EPSG then it reduces the EPSP as one expects, resulting in no spike. However, if the IPSG occurs in an optimum narrow time window ahead of the EPSG it facilitates the response, evoking a spike. From the model simulation results we see the deactivation time course of $g_{\text{KLT}}$ (Fig. 1B). The reduction in $w$ (the activation gating variable for $g_{\text{KLT}}$) together with the membrane potential level at the time when the EPSG is delivered will determine whether a spike occurs (Fig. 1, B and A). When $\delta$ is small, say just after the optimal window, the membrane is still quite hyperpolarized from the IPSG. So even though $w$ shows a substantial reduction from its rest value the cell is still too hyperpolarized to elicit a spike; spike generation would require an EPSG stronger than the specified subthreshold value. When $\delta$ is large, preceding the optimal window, $V$ and $w$ have nearly returned to their resting values and there is no spike from the EPSG. If properly timed, the EPSG may find the cell just moderately hyperpolarized while still finding a reduced $w$ and then a spike is evoked. This time window of facilitation depends on several factors including the strengths of $I_{\text{KLT}}$ and of the EPSG. When $G_{\text{KLT}}$ is decreased sufficiently the PIF window collapses, but (for these parameter settings of the model) until it is lost there is little change in its location along the $\delta$-axis (Fig. 1C). For a fixed IPSG size the upper boundary, maximum $\delta$, of the window for PIF increases sensitively (Fig. 1D) with the EPSG’s peak value, $G_{\text{ex}}$, and without bound as $G_{\text{ex}}$ approaches the threshold value for spike evocation from rest, $G_{\text{ex,rest}}$ ($\approx 8.57$ nS); this is the right branch of the U-shaped curve in Fig. 1D. For $G_{\text{ex}} > G_{\text{ex,rest}}$ the EPSG causes a spike unless it follows too soon behind the IPSG (i.e., if $\delta$ is too small). In the
case $G_{ex} \geq G_{ex,rest}$, the threshold curve is not U-shaped but monotonically increases for $\delta \geq 0$.

The PIF behavior and selective time window were also demonstrated experimentally in MSO cells in vitro (Fig. 2) using the dynamic-clamp technique to deliver the IPSG and EPSG pair. In control conditions the excitatory input generated by dynamic clamp stimulation was subthreshold when inhibition was delivered sufficiently before excitation, $> 3$ ms (Fig. 2A). However if the delay was between 2.5 and 1.5 ms, the same excitation when primed by inhibition evoked spike generation. The facilitatory effect of inhibition was mediated by $I_{KLT}$. After partial block of $I_{KLT}$ with 8 nM of DTXI, we reduced the amplitude of excitation and inhibition twofold to compensate for reduced membrane conductance. Under such conditions the subthreshold excitation was not facilitated to cause a spike (Fig. 2B). Furthermore, when the excitation was adjusted to be just-suprathreshold (Fig. 2C) it was able to evoke a spike only if it followed inhibition by $> 4$ ms. For smaller delays, inhibition always blocked spike generation, showing that DTXI eliminated facilitatory effects.

The PIF window depends on $\tau_{inh}$

The IPSG’s (alpha function) time constant, $\tau_{inh}$, plays an important role in positioning the PIF window. The IPSG’s peak value $G_{inh}$ is independent of $\tau_{inh}$ but the IPSG’s time integral increases linearly with $\tau_{inh}$, and is equal to $G_{inh}/\tau_{inh}$. For very short time constants, there is too little inhibitory current generated to adequately hyperpolarize the cell and deactivate $g_{KLT}$.—no PIF window appears (Fig. 3A). For larger $\tau_{inh}$, however, more current is generated by the IPSG, the cell hyperpolarizes more, and consequently $w$ falls well below its rest value. Subsequent to the IPSG the cell is hyperexcitable. If the membrane potential rises faster than $w$ can recover then a brief subthreshold EPSG may cause the cell to spike. If the EPSG is delivered too late then no spike will occur, thereby defining the upper $\delta$-value for the PIF window. If the EPSG is delivered too early, the membrane is still being actively hyperpolarized and shunted by the IPSG. To be effective, and for PIF to occur, an EPSG must wait until most of the IPSG has run its course. This explains why the onset time of the PIF window increases with $\tau_{inh}$; the dependency is nearly linear (as we will discuss below). If $\tau_{inh}$ is too large, the slow decay of the IPSG and its shunting effect forces the recovery of $V$ and $w$ to proceed hand in hand, disallowing $V$ to outrace the recovery of the deactivated $w$. Thus for large $\tau_{inh}$ there is no PIF window.

Within the PIF window, the degree of hyperexcitability is not uniform (Fig. 3B). The minimum strength of an EPSG to cause a spike depends on $\delta$, approaching $G_{ex,rest}$ (the from-rest threshold value) as $\delta$ becomes large, and rising to a vertical asymptote as $\delta$ becomes small. The just-adequate EPSG strength ($G_{ex}$ value at the curve’s minimum) for PIF increases with increasing $\tau_{inh}$ (except for very small $\tau_{inh}$). These plots show how the $\delta$-range for PIF varies, widening and its upper boundary increasing as $G_{ex}$ increases. However, the dominant factor for positioning the PIF window is the time constant for the IPSG.

The interplay between the opposing factors for PIF, hyperpolarization (with $V$ further from threshold) and deactivation of $g_{KLT}$ (hyperexcitability), may also be visualized with the help of phase-plane plots. In Fig. 3C we show projections from the five-variable model onto the $V$–$w$ plane of the trajectories for the cell response to a single IPSG, for different values of $\tau_{inh}$. The EPSGs are not included or needed here to illustrate the basic points; their associated EPSPs would overly complicate these images. A trajectory begins and ends at the rest state, with the flow direction being counterclockwise. The initial phase when the IPSG is applied shows hyperpolarization and

![FIG. 2. Experimental demonstration of postinhibitory facilitation. A: patch-clamp recordings (in vitro gerbil medial superior olive [MSO], P22) of membrane potential for a pair of inhibitory and subthreshold excitatory inputs, $G_{ex} = 20$ nS, are applied with the dynamic-clamp method with varying preceding times of inhibition, $G_{inh} = 100$ nS, (see text for experimental details). B: elimination of PIF after application of the $I_{KLT}$ blocker dendrotoxin I (DTXI). Over the same range of $\delta$ as in A the inhibitory conductance transient (IPSG) does not facilitate the subthreshold excitatory conductance (EPSG) to generate a spike. C: $G_{ex}$ was increased by 5% relative to B for just-suprathreshold EPSG, which was preceded by an IPSG in the presence of DTXI. In all cases of the $\delta$-range expected to show facilitation the IPSG suppressed spiking–no facilitation. To compensate for membrane conductance changes in B and C, the amplitudes of $G_{ex}$ and $G_{inh}$ were reduced approximately 2-fold from the control conditions.](http://jn.physiology.org/doi/10.1152/jn.00573.2005)
then reduction in $w$. Because $w$ lags behind $V$, the minimum in $w$ occurs after the minimum in $V$. The dot indicates when the IPSG is essentially complete and it has fallen to 10% of its peak value. Before the dot, $g_{\text{syn}}$ is strong and the net input resistance is reduced (below its resting value), enough to preclude a spike response to the EPSG. After the dot the PIF window is open, indicated by the heavy portion of a trajectory for the given amplitude of excitation $G_{\text{ex}}$ that was used for Fig. 3A. Later, as $V$ and $w$ are recovering to near their rest levels the window closes and the EPSG would not lead to a spike. We see here that the longer $\tau_{\text{inh}}$ is, the farther the PIF window migrates along the trajectory. For brief $\tau_{\text{inh}}$ the window opens even before $g_{\text{KLT}}$ is maximally deactivated. Note that for a fast-decaying IPSG the trajectory after the dot is basically reflecting the free (unstimulated) membrane’s subthreshold dynamics. Late in the time course all the trajectories return to the rest state, converging onto a common path and reflecting the eigenstructure for small deviations away from rest (see next section). For long $\tau_{\text{inh}}$ the slow decay means that there is still some residual synaptic conductance competing with the hyperexcitability and even some recovery of the also deactivated delayed rectifier conductance. Corresponding to the deactivation of $g_{\text{KLT}}$ the cell’s instantaneous (intrinsic, i.e., nonsynaptic) input resistance is transiently increased (Fig. 3D), contributing to the PIF by enhancing the depolarization from an EPSG. Because $g_{\text{K}}$ constitutes about 10% of the resting conductance its dynamics also contribute some degree (although much less than $g_{\text{KLT}}$) to hyperexcitability and these input resistance time courses show the combined effects.

Let us consider a bit more the timing of when the PIF window opens. In Fig. 3, C and D we associated this opening with near completion of the inhibitory conductance transient;
i.e., when \( g_{inh}(t) \) had fallen to a small fraction, say \( \epsilon \), of its peak level \( G_{inh} \). Before this time the shunting effect of the inhibitory conductance dominates the excitatory input. Therefore we estimate roughly the minimum \( \delta \) when PIF first becomes available from the condition (recalling our notation in Eq. 3)

\[
G_{inh}(t_0 - \delta + \delta_{min}) = \epsilon G_{inh}\text{ i.e.}
\]

\[
\epsilon G_{inh} = G_{inh}(\delta_{inh}/\tau_{inh})\exp[1 - (\delta_{inh}/\tau_{inh})]
\]

After canceling \( G_{inh} \) we see that this equation defines the value \( \delta_{min}/\tau_{inh} \) as some constant, depending only on the criterion level \( \epsilon \). Thus we have that \( \delta_{min} \) is approximately proportional to \( \tau_{inh} \).

**Reduced V–w model**

We have argued that \( I_{KLT} \) is the dominant factor that underlies PIF in the model. This view is based on the contention that the gating variables \( n \) and \( h \) are recruited only at higher voltages, and do not come into play until the membrane has reached threshold and spike initiation is proceeding. Here, we show this more convincingly by implementing the contention explicitly in a reduced model that retains only the dynamics of \( V \) and \( w \). We simulate and analyze two-variable model, highlighting the important features that contribute to PIF, for the reduced model we freeze \( h \) and \( n \) at their resting values (\( n_0 \) and \( h_0 \), respectively) and we exploit the fact that sodium activation responds very rapidly to changes in \( V \), approximating its dynamics as instantaneous, setting \( m = m_e(V) \). This reduced two-variable model reproduces semi-quantitatively the subthreshold PIF behavior. It is valid for approximating a spike’s upstroke but it does not apply for the transient depolarized phase of an action potential or for the spike’s repolarization. In fact, for the control parameter values the reduced model does not repolarize, but with sufficient stimulus to provoke an upstroke the membrane potential latches up to a steady depolarized level because sodium inactivation and the delayed-rectifier potassium activation are frozen. The equations for the reduced \( V–w \) model are expressed as

\[
C(dV/dt) = -I_{inh}(V) - G_{KL,T}(V - E_{KL,T}) - I_{syn}
\]

\[
dw/dt = \alpha_w(1 - w) - \beta_w w
\]

where \( I_{fast}(V) = G_N m^2_e(V) h_t(V - E_{Na}) + G_k h_t(V - E_K) + G_L(V - E_L) \). The parameter values, resting potential (about \( -60 \text{ mV} \)), and form of synaptic input current, \( I_{syn} \), are identical to those of the previous sections. The PIF effect occurs as before (Fig. 4A): when inhibition falls within an optimal time window ahead of the subthreshold excitation, a spike (upstroke) is elicited. Here we use weaker synaptic conductances than for the full model. The \( V–w \) model is more excitable because the delayed rectifier is frozen and an upstroke is easier to attain. The PIF window for the reduced model, as for the full model, shows a similar dependency on \( \tau_{inh} \) with the elliptical closed region in the \( (\tau_{inh}, \delta) \) plane (compare Fig. 4B with Fig. 3A) and the window’s onset value of \( \delta \) increasing almost linearly with \( \tau_{inh} \). The reduced model shows similar dependencies of the PIF window on \( G_{ex} \) as in the full model. The window enlarges with \( G_{ex} \) (two cases shown in Fig. 4B; also compare Fig. 4C with Fig. 3B). The PIF window’s offset value depends sensitively on \( G_{ex} \) (not shown, but similar to Fig. 1D). These simulation results demonstrate that the reduced \( V–w \) model well represents the significance of \( I_{KLT} \) as a primary factor in PIF in the Rothman et al. (1993) model.

Because the reduced model has only two variables we can gain further insight into its properties by analyzing it in the phase plane. As in many models of membrane excitability with a fast-activating inward current and slower-activating outward current we find the familiar cubic \( V \)-nullcline and the sigmoidal \( w \)-nullcline (Koch 1999; Rinzel and Ermentrout 1998) (Fig. 4D, inset). In our case, because we have frozen the negative feedback variables, \( h \) and \( n \), the nullclines have three intersections, i.e., the system has three steady states. The lower \( V \) state corresponds to the rest state and the upper \( V \) state is the “excited” state, where the upstroke terminates. The “middle” steady state is a saddle point and determines the threshold characteristics. A magnification of the phase plane (Fig. 4D, main panel) better shows the subthreshold and upstroke behavior. There are two trajectories (double arrowheads) that exactly enter the saddle point, one from above and one from below: this latter trajectory, called the **threshold separatrix** (FitzHugh 1955), segregates state points from which the response evolves to the resting state or to the excited state. If after a transient input the system finds itself rightward of this separatrix a spike upstroke occurs. The \( V–w \) trajectories for two responses from Fig. 4A are shown in Fig. 4D; the flow is counterclockwise, reminiscent of Fig. 3C, but here we can also see the trajectories in relation to the nullclines and the separatrix (not possible for the two-variable projection of Fig. 3C from a five-variable phase space). After the brief excitatory input is essentially complete (marked by an open circle) the system is sitting below threshold in the case \( \delta = 1 \text{ ms} \) and above threshold for \( \delta = 5 \text{ ms} \). Note that the positive slope of the separatrix reflects that (in the range being viewed) the distance from \( V_{fast} \) to threshold decreases for lower values of \( w \), a geometric or phase plane correlate of hyperexcitability. Thus in the case of the upstroke trajectory, the EPSG is delivered after the IPSG is complete when (coincidentally) \( V = V_{fast} \) and the membrane is hyperexcitable, so that a spike is evoked.

**Rebound excitation after long or brief hyperpolarization; geometric treatment for \( V–w \) model**

As one should expect, the reduced model’s applicability has a limited quantitative range. It works well for the qualitative feature of PIF, considering the responsiveness of priming transient inhibition. In this case the hyperpolarized levels do not fall below \( E_{inh} \) (−66.5 mV). For stronger hyperpolarizations, say in response to \( I_{syn} \) steps, the \( V–w \) model, unlike the full model, predicts classical PIR. This can be understood from looking at a zoom-out of the phase plane (Fig. 5A). Here we see as we move away from the saddle point along the threshold separatrix that it makes a U-turn, crossing the \( w \)-nullcline at \( V \approx -65.6 \text{ mV} \) (see arrow) and exiting this zoom-out’s left border near \( w = 0.4 \). Now, for setting up conditions for PIR, suppose that a long hyperpolarizing current step is applied. Then the state point approaches a holding steady state with \( V \) and \( w \) that must lie along the \( w \)-nullcline, below rest. If this holding level has \( V \) less than \( -65.6 \text{ mV} \) then the state point is below the threshold separatrix and when the current step is ended an upstroke (i.e., PIR) occurs. The full model does not show PIR for release from any maintained current steps. This is because \( n \) is not frozen and \( \tau_n \) is small enough so that when \( V \) begins
increasing on release and it increases beyond $V_{\text{rest}}$ the $n$-gates open rapidly, more than at rest (the frozen level in the $V$-$w$ model), and thereby preclude a spike. If $n$ is adequately slowed by increasing $\tau_n$ artificially enough the full model does show PIR. The sensitivity of PIR to the degree of $I_K$ activation around rest is also somewhat reflected in the $V$-$w$ model. For example, if the frozen value $n_0$ were increased from 0.0194 to 0.05 (still only 5% activation) PIR would not occur in the $V$-$w$ model. The separatrix would not cross the $w$-nullcline (Fig. 5B) and the reduced and full models would agree in this regard, as well as for PIF.

The $V$-$w$ model is quite responsive to such strong hyperpolarizing inputs, even if they are very brief. For example, a 1-ms (1-nA) anodal current of a given amplitude is adequate to cause a rebound upstroke (Fig. 5A), although a briefer input of 0.5 ms is not. However, a briefer hyperpolarizing current step if made very strong can lead to rebound. In the extreme case of a delta-function of $I_{\text{app}}$ the membrane is instantaneously hyperpolarized and if this $V$-displacement below rest is large enough [here, $\approx -90.8$ mV (see arrow), just leftward of where the threshold separatrix crosses upward through $w = w_{\text{rest}}$] then rebound occurs (Fig. 5A). In this case of such a U-turning separatrix, the cell model behaves as if it had two thresholds for brief $I_{\text{app}}$ inputs: one for depolarizing pulses and one for hyperpolarizing pulses. The observation that a strong instantaneous hyperpolarizing current pulse can evoke a rebound spike.
was reported by FitzHugh for the classical HH model (FitzHugh 1976); here, by showing it in a two-variable model, we provide a geometric phase-plane interpretation (see also Izhikevich 2001).

**Integrate-and-fire model with dynamic accommodating threshold**

Until now we have focused on a biophysical mechanism for PIF and hyperexcitability, transient deactivation of a voltage-gated potassium current mediated by brief inhibition. This is a particular instantiation of the general mechanism that underlies PIF, a transient decrease in effective threshold by hyperpolarization. Next we implement this idea explicitly in a bare-bones model, an enhanced leaky integrate-and-fire (LIF) model.

Here, PIF does not depend on a conductance mechanism per se. In this minimal model the voltage threshold is a dynamic variable \( \theta(t) \), which increases with depolarization and decreases with hyperpolarization with a time constant, and which can be chosen to be faster or slower than the membrane time constant. The dynamic behavior of \( \theta \) represents an idealized form of accommodation [in the spirit of Hill (1936) and others (see Rashevsky 1960)]. Transient, even brief, synaptic inhibition and hyperpolarization causes \( \theta(t) \) to decrease, inducing hyperexcitability. The model’s equations in dimensionless form are

\[
\begin{align*}
\frac{dv}{dt} &= -v + i_{\text{syn}} - i_{\text{sh}} \\
\frac{d\theta}{dt} &= \frac{(\theta - \theta_0) - v}{\tau_\theta}
\end{align*}
\]

Voltagelike quantities \((v\) and \(\theta)\) are expressed relative to rest and scaled by a reference voltage. Timelike quantities (e.g., \(t\) and \(\delta\)) are scaled by the membrane time constant; \(i_0\) sets the resting or holding voltage and \(\theta_0\) determines the resting level of threshold. The synaptic current is given by \(i_{\text{syn}} = g_{\text{syn}}(v - 2) + g_{\text{syn}}(t)\), where the synaptic conductances have the same functional form as in Eqs. 2 and 3; here, inhibition is pure shunting, although this aspect is not crucial. As in standard LIF models, \(v\) is reset to some subthreshold level (say, \(v = 0\)) when \(v\) reaches \(\theta\) and a spike event is recorded. Note, this LIF-\(\theta\) model is distinct from most other LIF models with variable threshold (or processes that affect threshold) that are meant to describe refractoriness and spike-frequency adaptation. In the latter, threshold depends on spiking but not typically on subthreshold \(v(t)\) (see, e.g., Dayan and Abbott 2001; Liu and Wang 2001; but also Perkel et al. 1981).

Our idealized LIF-\(\theta\) model indeed shows the PIF effect (Fig. 6, A and B). If excitation occurs too soon after inhibition the subthreshold EPSG is disadvantaged rather than facilitated. However, if the inhibition is delivered within a proper preceding time window then facilitation and a spike occurs. As in the previous models the PIF window’s position slides to greater \(\delta\) as \(\tau_{\text{inh}}\) increases (Fig. 6C). As we did for the \(V-w\) model we can analyze this two-variable LIF model in the phase plane (Fig. 6D). The nullclines and all behavior for \(\nu < \theta\) are linear. The trajectories for two cases, analogous to those in Fig. 6, A and B, show the expected counterclockwise flow with the threshold being reduced, and reaching a minimum as the \(\theta\)-nullcline is crossed. When \(\delta\) is too small the state point is still “below” (to the left of) the “firing line” just after the excitatory input, so no spike occurs. Waiting a bit longer (\(\delta\) somewhat larger but not by too much) means that the excitatory event pushes the system rightward over the firing line.

We can say more about this idealized two-variable model. Its eigenvalues are: \(-1\) and \(-1/\tau_\theta\); here, the rest or holding state does not have spiral structure. If \(v\) lags behind \(\theta\) (\(\tau_\theta > 1\)) then the decay back toward the holding state from hyperpolarizing input ultimately will have its trajectory in the phase plane becoming tangent to the \(v\)-nullcline. That is, the eigenvector...
associated with the slower decay \( \exp(-t/\tau_p) \) is vertical; consequently, there can be no overshoot of \( V \) in this model after inhibition. From the geometrical perspective we can also see how the competing factors of hyperpolarization and hyperexcitability determine where along the return-to-rest trajectory the maximal facilitation occurs. Suppose we ask: At each position along the trajectory after a priming inhibitory input what is the just-adequate value of a brief (delta-function) pulse of depolarizing \( I_{\text{app}} \) (delivering charge \( Q_{\text{ex}} \)) that would displace the membrane potential across the firing line? The answer, for \( Q_{\text{ex}} \) at time \( \delta \), is given by the horizontal distance from the trajectory to the firing line (Fig. 6D). Because the firing line is sloped positively (slope = 1) and the ultimate approach to rest is vertical we can see that as we back away from the rest point (\( \delta \) decreasing from large values) \( Q_{\text{ex}} \) decreases then reaches a minimum and increases again. The plot of \( Q_{\text{ex}} \) versus \( \delta \) has the same shape as the plots in Figs. 3B and 4C.

This LIF-\( \theta \) model shows PIR behavior and this may be seen as follows. Because the firing line has slope of 1, it will intersect the \( v \)-nullcline at a low value of \( \theta \). If a steady hyperpolarizing \( I_{\text{app}} \) is strong enough, the prerelease steady state will lie at an even lower \( \theta \)-level, down on the lower left of the \( \theta \)-nullcline. Then, if \( \tau_p \) is large, the system on release will reach the firing line (a rebound spike) before \( \theta \) has time to recover very much; in fact, there could be several rebound spikes before \( \theta \) adequately recovers. One can easily identify the least negative steady \( I_{\text{app}} \) that results in rebound on release, i.e., by finding which starting point on the \( \theta \)-nullcline is closest to rest but still leads to PIR. Having identified this threshold for PIR from steady hyperpolarization we can conclude that the LIF-\( \theta \) model also gives a rebound spike for a strong enough delta pulse of hyperpolarizing \( I_{\text{app}} \): By following backward in time along the threshold PIR trajectory we see that it will ultimately cross the horizontal line that passes through the rest.
state. The $\nu$-coordinate of this crossing point defines the hyperpolarization threshold for the delta pulse that could produce a rebound spike, analogous to what we saw previously with the $V$–$w$ model. Finally, we note that the LIF-$\theta$ model can fire phasically. If $\alpha > 1$, then the $\nu$- and $\theta$-nullclines will intersect only to the left of the firing line for any level of steady depolarization $I_{\text{app}}$. Thus after the step is applied, the system’s trajectory will go to this target, maybe hitting the firing line a few times beforehand (phasic firing); the larger the value of $\tau_{\text{dp}}$, the more spikes will occur during this transient phase.

**PIF for periodic inputs**

Until this point we have considered PIF as it occurs for an IPSG–EPSG input pair. The essential mechanism that we have identified is also observable in the responses to longer input trains. Here we illustrate the PIF phenomenon for periodic trains of synaptic inputs, EPSG trains and IPSG trains of the same period but phase-shifted. The in vitro recording of an MSO neuron (Fig. 7A) shows facilitation when a phase-locked train of IPSGs is presented with a periodic train of subthreshold EPSGs and the phase difference is in an appropriate range. In the absence of inhibition, the pure EPSG train elicited no spikes (Fig. 7A, bottom trace). The spiking, arising from PIF, occurred with multiple ratios of frequency locking, but predominantly 1:1. Computations with the full model show PIF for phase-locked periodic inputs for a range of $\beta/\gamma$ values and for various frequencies (Fig. 7, B and C). For any periodic train of combined inputs, IPSG–EPSG pairs occur automatically with the same frequency as that of the input. If a single pair results in a spike by PIF, the combined input will now result in 1:1 firing as long as the input period is sufficiently longer than the refractory period. As the input frequency increases, the refractory period begins to overlap with the time window necessary for PIF spike production. This prevents some IPSG–EPSG pairs from producing a spike. Thus for increased input rate, the firing ratio falls below 1:1. In Fig. 7C the dominant spike mode is 1:1 for 100-Hz input, but for a 200-Hz input the ratio is almost 1:2.

In the context of this periodic stimulation protocol we further explore some parameter dependencies. The PIF regions in Fig. 7C (and in Fig. 3A) are for $\tau_{\text{inh}} < 1$ ms, primarily because of the fact that the membrane time constant and $\tau_{\text{w}}$ are small, say $\approx 2$ ms. Also, in the application of this model to MSO ITD tuning, Brand et al. (2002) used $\tau_{\text{inh}} = 0.1$ ms. This value is extremely short even when compared to some of shortest inhibitory decay time constants reported for auditory brain stem (Awatramani et al. 2004). Recent estimates for the glycine-mediated inhibition in MSO place $\tau_{\text{inh}}$ around 2 ms (Magnusson et al. 2005). By making modest modifications to the model we have extended upward the range of $\tau_{\text{inh}}$ for PIF. For the results in Fig. 7, D and E we increased the leakage conductance and used a fast-rising $G_{\text{inh}}(t)$ (instantaneous) with a pure exponential decay rather than an alpha-function time course. The primary effect of using the exponential $G_{\text{inh}}(t)$ is that the absolute phase shift $\delta$ is smaller (Fig. 7D) than that in Fig. 7C: $\approx 3$ ms for 100-Hz input and $\approx 4$ ms for 200-Hz input. For this case, $G_{\text{ex}}$ is 20% below the threshold value for spike generation from a single EPSG, and $G_{\text{inh}}$ is 25 nS. Figure 7E illustrates the PIF region in the $(G_{\text{inh}}, \delta)$ plane for $\tau_{\text{inh}} = 2$ ms and two input frequencies, 100 and 200 Hz. The enhanced leakage conductance allows PIF to occur for $G_{\text{inh}}$ values as low as 15 or 10 nS. This region is robust with respect to variations in $G_{\text{ex}}$ (data not shown).

**Discussion**

Brief, well-timed inhibition can induce a spike-generating response to a subsequent, otherwise subthreshold, excitatory synaptic input. The brief inhibition leads to a transient phase of hyperexcitability. The trade-off between hyperpolarization and hyperexcitability leads to a specific time window for the postinhibitory facilitation, PIF. This window’s width (possibly quite narrow) and temporal position are determined by the membrane time constant and that of the subthreshold negative feedback that underlies PIF. The PIF window’s onset time increases with $\tau_{\text{inh}}$ and its width/size increases with the excitatory synaptic strength. The PIF phenomenon in its simplest form is for an inhibitory–excitatory pair but the effects are also seen in longer trains. Our findings suggest that patterns of inhibition–(subthreshold) excitation timing can increase firing probability beyond that from just temporal summation of subthreshold excitation.

We demonstrate PIF here with a case study: experimentally, in MSO neurons in vitro and in a conductance-based model (Rothman et al. 1993), dissecting the latter to analyze the mechanism. In this system brief inhibition transiently decreases a low-threshold potassium conductance, below its partially activated level at rest. The basic mechanism of fast inhibition leading to temporary reduction of a subthreshold negative feedback is quite general. We formulate two different simplified two-variable models and demonstrate, with simulation and phase plane analysis, the essential and qualitative features of PIF. Both of these reduced models also show PIR. Moreover, they can rebond with a spike for extremely brief hyperpolarizing current injection, suggesting that some neurons have two thresholds for input pulses, depolarizing and hyperpolarizing.

**Biophysical factors that affect PIF**

A number of factors influence the PIF effect and the window for facilitation, its width and position. We have shown (Figs. 3A, 4B, and 6C) that the window’s onset time increases with the IPSG’s time constant $\tau_{\text{inh}}$. The onset time also depends on the time constant for the determining subthreshold negative feedback. This time constant (together with the membrane time constant) influences the membrane’s intrinsic dynamics during the recovery after synaptic inhibition, i.e., during the phase of hyperexcitability. In the five-variable model, and in our reduction of it, $\tau_{\text{KLT}}$ sets the scale of a few milliseconds for the window’s onset; $\tau_{\text{p}}$ is the analogous factor in the LIF-$\theta$ model. More generally, other dynamic negative feedback processes could contribute to PIF. For example, if $I_{\text{Na}}$ is partially inactivated at rest then reducing this inactivation by transient hyperpolarization could be a significant player in the integration of subthreshold inputs (Svirskis et al. 2004) and PIF, and then $\tau_{\text{p}}$ would influence the PIF timescales. In the full model (Rothman et al. 1993) $I_{\text{Na}}$ inactivation ($h$) and $I_{\text{K}}$ activation ($n$) are secondary to $I_{\text{KLT}}$ gating for PIF. However if, say, $I_{\text{KLT}}$ were blocked or $w$ were frozen in the full model then $n$-dynamics would play a dominant role (not shown here).
FIG. 7. PIF occurs for periodic inputs in experiments and simulations; in the model, for extended parameter regimes: \( \tau_{\text{inh}} \) slower (\( \leq 3 \) ms, from previously \( < 1 \) ms) and weaker inhibitory strengths. A: recordings of an in vitro MSO cell for periodic trains of EPSPs and phase-shifted IPSPs for different phase shifts measured with respect to the input period at 200 Hz; the relative phase shift is \( \delta/T \). Bottom trace: response, nonspiking, for pure subthreshold periodic excitation. For a particular range of \( \delta/T \) inhibition facilitates firing. Inputs (delivered with the dynamic-clamp method) are pure exponential functions with \( G_{\text{ex}} = 40 \) nS, \( G_{\text{inh}} = 100 \) nS. B: computed voltage responses of the complete model (Rothman et al. 1993) showing phase- and frequency-locked spiking for inputs similar to those in A. Inputs are modeled as alpha functions (\( \tau_{\text{inh}} = 0.3 \) ms) with rate 150 Hz; all other parameters are the same as in Fig. 1A. C: computed PIF regions for input rates of 100 and 200 Hz for parameters as in Fig. 3A. Color coding in C–E indicates the firing rate of the membrane relative to the input rate. D and E: with enhanced leakage conductance (12.2 nS) and for inputs that are simple exponentials (not alpha functions) the model shows PIF for extended parameter ranges: longer inhibitory time constants and weaker synaptic strengths. In D the PIF region in the \( (\delta/T, \tau_{\text{inh}}) \) parameter plane extends up to \( \tau_{\text{inh}} = 3 \) ms, for these frequencies, 100 and 200 Hz. Note, the relative phase shifts for the PIF range correspond to absolute time differences between inhibition and excitation, say for 100 Hz, of 0.5 to 2.5 ms, considerably less than those in Fig. 3A. Here, \( G_{\text{ex}} \) (15.6 nS) is 20% below the threshold value for a single excitatory input to generate a spike. In E the PIF regions are shown in \( (G_{\text{inh}}, \delta) \) space for \( \tau_{\text{inh}} = 2 \) ms. Facilitation occurs for \( G_{\text{inh}} \) values as low as 15 nS (or smaller); in this regime an isolated inhibitory transient input hyperpolarizes the membrane only 1.6 mV (or less).
onset time of the PIF window would be reduced, down to a millisecond or so, because \( \tau_w \) is much smaller than \( \tau_v \). As we have shown, the offset time of the PIF window increases with the EPSG strength, in each of the models that we have discussed.

**PIF and subthreshold dynamics**

If the time constants of the subthreshold negative feedback are too fast compared with the membrane time constant we would not observe PIF; the hyperexcitability phase would end too soon. If too slow then brief hyperpolarization will not enable hyperexcitability to appear. When, say, \( \tau_{KLT} \) is not too different from the membrane time constant, however, then we find an opportunity for a PIF window for some IPSG and EPSG amplitudes. The important point is the dynamic interplay between \( V \) and the subthreshold process that induces hyperexcitability. This interplay is visualized as the counterclockwise flow in our phase-plane views (it would look clockwise in the \( V-h \) plane if \( I_{Na} \) inactivation were the dominant factor). Various transient inputs can push the system into this region of phase space and onto a trajectory that contains a PIF window. It would be an oversimplification to say that PIF occurs only if the subthreshold behavior involves a strictly spiraling-like behavior after transient perturbing inputs. Although subthreshold (damped) resonance is sufficient (see also Izhikevich 2001), and might even result in spike “probability oscillation” after an inhibitory input in a stochastic background (Luk and Aihara 2000), it is not necessary for PIF. Our LIF-\( \theta \) model shows PIF behavior but the ultimate trajectory in returning toward rest is not spiral, and \( V \) does not overshoot \( V_{\text{rest}} \).

**Beyond specified pairings**

In most of our illustrations of PIF the input was an IPSG-then-EPSG input pair with specified timing. However, the essential mechanism is general and also underlies enhancement effects of spiking for longer input trains and with temporal ordering unspecified. We illustrated (Fig. 7) the PIF phenomenon for periodic trains of synaptic inputs, EPSG trains and IPSG trains of the same period but phase-shifted. Although we understand the enhancement of firing by considering the effect of an IPSG priming the membrane for hyperexcitability for a succeeding EPSG the notion of which PSG leads is not obviously a priori meaningful for periodic inputs. More generally, for input trains with randomly arriving EPSGs and IPSGs some PIF-favorable timings will occur by chance. Indeed, our simulations with the classical HH model show that introducing random, fast IPSGs to a train of subthreshold EPSGs enhances the spontaneous firing rate. Identification of spike-triggered synaptic events reveals the enhancement in firing is attributed to timely IPSG–EPSG pairings (Dodla and Rinzel 2006).

**Delivery and release rate of hyperpolarization**

Even very, very fast hyperpolarizing input can induce PIF. For our \( V-w \) and LIF-\( \theta \) models we saw that, in the extreme case of a delta-pulse input, we get essentially instantaneous hyperpolarization and the subsequent recovery trajectory not only contains a PIF window but, on its own, may lead to a rebound spike (Fig. 5). The geometrical (phase-plane) view enables us to see that for such a case of rebound spiking after a delta pulse that classical PIR (a rebound spike after release from adequate and steady \( I_{\text{app}} < 0 \) was also guaranteed to occur. This is an interesting connection between the effect of long duration and very briefly applied hyperpolarization.

In either of these cases the hyperpolarizing stimulus is instantaneously terminated. This feature of classical PIR is sometimes overlooked. One should think more generally that a fast enough return to zero stimulus (\( I_{\text{app}} = 0 \)) is required to induce rebound (Ferragamo and Oertel 2002; Gutierrez et al. 2001). This necessity is consistent with our observations on PIF that \( g_{\text{inh}}(t) \) must be fast enough (Figs. 3 and 7) and underscores, in particular, that the decay must be fast enough. If the stimulus relaxes too slowly the system would merely track the slowly recovering hyperpolarized state back to rest. The opportunity would be lost for exploiting the hyperexcitability and have \( V \) outtrace the slightly lagging recovery of the negative feedback.

For the brain stem model we found PIF for \( \tau_{\text{inh}} \) in the submillisecond (Fig. 3A) to millisecond (Fig. 7D) ranges. In the experimental/computational study of gerbil MSO ITD tuning (Brand et al. 2002) the modeling results were based on \( \tau_{\text{inh}} = 0.1 \) ms. Recent patch-clamp data are finding in these neurons that \( \tau_{\text{inh}} \) is closer to 2 ms (Magnusson et al. 2005). Such brief inhibition is not unique to the auditory system. In hippocampal slices \( \tau_{\text{inh}} \) values of 2 ms or so have been reported (Bartos et al. 2002).

The reduced \( V-w \) model

The saddle-point threshold behavior visualized in the phase plane of Figs. 4D and 5, A and B is not necessary for PIF behavior, but this instantiation of excitability enables a clear understanding. Moreover, similar geometry is common to a class of neuronal models and therefore supports the generality of our findings for the illustrative case study from the auditory brain stem. This same geometry and similar behaviors are obtained for a two-dimensional reduction of the classical HH model with \( m = m_w(V) \) and \( h \) frozen at its rest value (Dodla and Rinzel 2006).

We note here that our formulation of the reduced \( V-w \) model was not based on a mathematical derivation from the full model of Rothman et al. (1993). Usually such reductions lean on timescale differences in dynamic processes: e.g., allowing fast variables to track instantaneously quasi-steady-state values [as we did, by setting \( m = m_w(V) \)] or by fixing slow variables to their values at a rest or holding state. In our case, we ignored the dynamics of \( h \) and \( n \) by noting that they were not significant players as dynamic negative feedback in the subthreshold voltage range; this simplification proved reasonably successful for addressing some issues, such as responses to IPSGs and EPSGs. Interestingly, in this model \( n \) is noticeably faster than \( w \) for \( V < V_{\text{rest}} \), it is just very small. This leads us to caution that assessing the contribution of various gating factors to subthreshold integration depends not only on their \( V \)-dependent gating amplitudes but on the \( V \) dependency of their time constants. In some cases the tail-end values of the typically bell-shaped \( V \) dependency of \( \tau \) (or the atypically shaped profiles of \( \tau_h \) and \( \tau_w \) in this model) may be near zero and so the identification of fast/slow timescales requires careful consideration.
Timed inhibition and coding

In considering a system’s temporal coding properties one often asks about the precision in detecting nearly coincident subthreshold excitatory inputs. Interestingly, as we have shown here, the same intrinsic biophysical factors that contribute to high-fidelity coincidence detection in the auditory brain stem case endow the neurons with enhanced firing probability for precise and preferred timing between subthreshold excitation and preceding brief inhibition. Although timed and fast inhibition plays a role in temporal processing and ITD tuning in the gerbil MSO (Brand et al. 2002; Grothe and Sanes 1994) the evidence at this point suggests that the contribution there is to reduce firing to deselect some ITDs and create asymmetry in the tuning curve. Whether PIF plays a role in ITD sensing by MSO or in other signal processing by auditory brain stem remains to be seen. Nevertheless, other timing computations in the auditory system involve transient inhibition that precedes excitation. In the bat auditory system delay sensitivity (for outgoing and incoming sonar signals) is seen in some neurons in the cortex and inferior colliculus. Localized pharmacological blocking of transient inhibition onto these cells degrades or eliminates the delay-sensitive firings (Wenstrup and Leroy 2001). Bat IC neurons that show sensitivity to sound duration are also seen to fire for appropriately timed (otherwise subthreshold) excitation after offset of inhibition (Casseday et al. 1994; Covey and Casseday 1999; Covey et al. 1996). Although the absolute timescales of biophysical mechanisms are longer in higher structures such as IC than in MSO, it is the relative values that determine whether PIF might occur. The mechanism that we describe here is a viable candidate for such phenomena. Our modeling results, especially for the reduced models, suggest that postinhibitory facilitation is a quite general feature. Even for random inputs one can expect favorable timings for PIF to occur by chance, thereby providing a contribution to spontaneous activity (Dodla and Rinzel 2006). Surely many neural systems for implementing computations can take advantage of the synergy between transient subthreshold excitation that is primed by relatively brief or rapidly decaying timed inhibition.

Appendix

The voltage dependencies of the rate constants $\alpha_m$ and $\beta_m$ of the full model described in the paper, identical to those used by Rothman et al. (1993), are listed below:

$$\alpha_m = 0.36 \times T_f(3)(V + 49)/(1 - \exp(-(V + 49)/3))$$

$$\beta_m = -0.4 \times T_f(3)(V + 58)/(1 - \exp(V + 58/20))$$

$$\alpha_s = 2.4 \times T_f(3)(1 + \exp(V + 68/3)) + 0.8 \times T_f(10)(1 + \exp(V + 61.3))$$

$$\beta_s = 3.6 \times T_f(3)(1 + \exp(-(V + 21)/10))$$

$$\alpha_n = -0.0282 \times T_f(3)(V + 9)/(1 - \exp(-(V - 9)/12))$$

$$\beta_n = 6 \times T_f(3) \exp(-(V + 144)/30) + 6 \times T_f(3)(1 + \exp(V + 62))$$

$$\alpha_w = 0.107 \times T_f(3)(1 + \exp(-(V + 33)/13.1))$$

$$\beta_w = 0.01881 \times T_f(3) \exp(-(V + 30)/30.3)$$

The conductances are given by $G_K = 325 \times T_f(2), G_L = 40 \times T_f(2.5), C_{K_L,T} = 20 \times T_f(2.5), \text{ and } G_L = 1.7 \times T_f(2)$. The temperature factor is $T_f(T) = Q_{10}^{(T-20)/10}$, where $T = 38^\circ C$. Integration of the equations was done using a fourth-order Runge–Kutta method with a fixed time step of $\leq 0.01$ ms. All the nullclines and separatrices in the phaseplanes were obtained using the software package XPPAUT (Ermentrout 2002).

Acknowledgments

The authors are grateful to the Mathematical Biosciences Institute at Ohio State University (http://mbi.osu.edu/) for providing support and a stimulating environment during and following a workshop on the Auditory System (May 2003) when many of the ideas and results presented here were developed.

Grants

This research was supported by funding from National Institute of Mental Health Grant MH-62595.

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