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# Slow Variable Dominance and Phase Resetting in Phantom Bursting

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#### Abstract

Bursting oscillations are common in neurons and endocrine cells. One type of bursting model with two slow variables has been called 'phantom bursting' since the burst period is a blend of the time constants of the slow variables. A phantom bursting model can produce bursting with a wide range of periods: fast (short period), medium, and slow (long period). We describe a measure, which we call the 'dominance factor', of the relative contributions of the two slow variables to the bursting produced by a simple phantom bursting model. Using this tool, we demonstrate how the control of different phases of the burst can be shifted from one slow variable to another by changing a model parameter. We then show that the dominance curves obtained as a parameter is varied can be useful in making predictions about the resetting properties of the model cells. Finally, we demonstrate two mechanisms by which phase-independent resetting of a burst can be achieved, as has been shown to occur in the electrical activity of pancreatic islets.

#### Keywords

Bursting; Multi-scale; Islet; Oscillations

#### **1** Introduction

Bursting oscillations, episodes of electrical activity followed by quiescence, are common in neurons and endocrine cells. Many mathematical models for bursting cells have been developed (Coombes & Bressloff, 2005), and singular geometric perturbation analysis (also called fast/slow analysis) has proven to be very useful in the analysis of such models (Rinzel, 1987; Rinzel & Ermentrout, 1989). This makes use of the separation of times scales

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between those variables that change rapidly (the *fast variables*) and those that change slowly (the *slow variables*). Many models contain a single slow variable, for example, (Butera *et al.*, 1999; Chay & Keizer, 1983), while others contain two or more slow variables, for example, (Bertram & Sherman, 2004; Rinzel & Lee, 1987). One type of bursting model with two slow variables produces *phantom bursting*, so named because the burst period can be a blend of the slow variables (Bertram *et al.*, 2000); therefore, the search for a single slow process with time constant similar to the burst period in experimental studies would be fruitless.

In the phantom bursting model described in (Bertram *et al.*, 2000) there are two slow variables,  $s_1$  and  $s_2$ , with very different time constants,  $\tau_{s_2} >> \tau_{s_1}$ . As a result,  $s_2$  changes appreciably slower than  $s_1$ . Depending on the values of other parameters, the bursting that is produced may be fast (short period), driven by  $s_1$ ; slow, driven by  $s_2$ ; or medium, driven by a combination of both slow variables. In pancreatic  $\beta$ -cells, the cells for which the model was developed,  $s_1$  could correspond to the fraction of  $K^+$  channels activated by cytosolic  $Ca^{2+}$  and  $s_2$  to the fraction of ATP-sensitive  $K^+$  channels activated by the ratio of ADP to ATP or to the  $Ca^{2+}$  concentration in the endoplasmic reticulum (Bertram & Sherman, 2004).

Phantom bursting has been analyzed using fast/slow analysis to understand the mechanism of bursting and the wide range of burst periods that can be produced (Bertram *et al.*, 2000; Bertram & Sherman, 2004, 2005). While this analysis clarified why the different slow variables control the fast or slow bursting and how the two work together to produce medium bursting, the relative contributions of the two slow variables to the generation of the medium bursting was not determined. That is, for a given medium bursting pattern, it was never determined quantitatively how much  $s_1$  contributed to the burst period and how much  $s_2$  contributed. In this article, we describe a measure, which we call the *dominance factor*, of the relative contributions of the two slow variables to the phantom bursting model described in (Bertram *et al.*, 2000). Since the contributions of the variables may be different during the active and silent phases of bursting, we compute dominance factors for both phases. Using this tool, we demonstrate how the control of different phases of the burst can be shifted from one slow variable to another by changing a model parameter. We then show that the dominance factor curves obtained as a parameter is varied can be useful for making predictions about the resetting properties of the model cells.

One feature of bursting driven by a single slow variable is that it is possible to reset the oscillation from the silent to active phase, or vice versa, with a sufficiently large perturbation. The phase that follows the reset should be shorter than normal, since the slow variable has not had time to reach its typical starting point for that phase. Resetting experiments were performed on intact pancreatic islets by Cook and associates (Cook et al., 1981) to test these predictions. They found that resetting was indeed possible, but that the phase following the reset was often of full length. That is, for most silent-active resets the following active phase was no shorter than normal, and for most active-silent resets the following silent phase was of full duration. We refer to these as *full-length resets*. Fulllength resets in both directions (bidirectional full-length resets) were shown in the same islet in one case (Figs. 3 and 4 in (Cook et al., 1981)). A later study showed full-length silentactive resets, but short active-silent resets (Zimliki et al., 2003). The existence of a fulllength reset in either direction indicates that the bursting is driven by more than one slow variable. But how? We demonstrate that a full-length reset can be produced if one slow variable determines the silent phase duration while the other slow variable determines the active phase duration. This explanation was postulated earlier (Smolen & Sherman, 1994) and demonstrated with a  $\beta$ -cell model in which the time constants were adjusted so that one slow variable changes rapidly during the silent phase (so that the other slow variable controls the silent phase duration), and vice-versa for the active phase. We use a similar

approach to account for cases where full-length resets occur in both directions. However, unidirectional full-length resets can be accounted for with the phantom bursting model by simply adjusting a system parameter so that the dominance curves for the active and silent phases are well separated. We illustrate this, and the case of bidirectional full-length resets, later. Thus, the dominance factor is both a tool for understanding the dynamics of fast/slow systems with two slow processes and a practical tool for making testable predictions.

#### 2 Generic Phantom Bursting Model

The generic phantom bursting model for pancreatic  $\beta$ -cells is composed of fast and slow subsystems (Bertram *et al.*, 2000). The fast subsystem consists of the cell's plasma membrane potential (*V*) and the activation variable (*n*) for the delayed rectifier  $K^+$  current. The slow subsystem consists of two distinct slow negative feedback variables,  $s_1$  and  $s_2$ . These are activation variables for slowly activating  $K^+$  currents  $I_{s1}$  and  $I_{s2}$ , respectively. Both  $s_1$  and  $s_2$  are slow in relation to *V* and *n*, which operate on a time scale of tens of milliseconds. However, the  $s_1$  variable, with time constant  $\tau_{s1} = 1$  sec, is considerably faster than  $s_2$  with  $\tau_{s2} = 2$  min.

The model equations are:

$$\frac{dV}{dt} = -(I_{c_a} + I_{\kappa} + I_{s_1} + I_{s_2} + I_L)/C_m$$
(1)

$$\frac{dn}{dt} = (n_{\infty}(V) - n)/\tau_n(V)$$
(2)

$$\frac{ds_1}{dt} = (s_{1\infty}(V) - s_1)/\tau_{s1}$$
(3)

$$\frac{ds_2}{dt} = (s_{2\infty}(V) - s_2)/\tau_{s2},$$
(4)

with ionic currents:

$$I_{Ca} = g_{Ca} m_{\infty}(V) (V - V_{Ca}), I_{K} = g_{K} n(V - V_{K})$$
(5)

$$I_{s1} = g_{s1}s_1(V - V_K), I_{s2} = g_{s2}s_2(V - V_K)$$
(6)

$$I_L = g_L (V - V_L). \tag{7}$$

 $I_{Ca}$  is an inward  $Ca^{2+}$  current that activates very rapidly (assumed instantaneous),  $I_K$  is a rapidly activating outward  $K^+$  current, and  $I_L$  is a leak current.  $C_m$  is the membrane capacitance of the cell. The g parameters are the maximum current conductances, and  $V_{Ca}$ ,

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 $V_K$ , and  $V_L$  are the reversal potentials. The activation curves for *m*, *n*,  $s_1$ , and  $s_2$  are sigmoidal Boltzman functions, which increase with membrane potential:

$$m_{\infty}(V) = \frac{1}{1 + \exp[(-22 - V)/7.5]}, n_{\infty}(V) = \frac{1}{1 + \exp[(-9 - V)/10]}$$
(8)

$$s_{1\infty}(V) = \frac{1}{1 + \exp[(-40 - V)/0.5]}, \ s_{2\infty}(V) = \frac{1}{1 + \exp[(-42 - V)/0.4]}.$$
(9)

The only voltage-dependent time constant is  $\tau_n$ :

$$\tau_n(V) = \frac{9.09}{1 + \exp[(V+9)/10]} \tag{10}$$

As V is varied over the interval [-55,-20] mV,  $\tau_n$  ranges from 6.2 to 9.0 msec. The fast subsystem governs spiking during the active phase of a burst, while the slow subsystem controls when the spiking is turned on and off. The spiking activity causes  $s_1$  and  $s_2$  to slowly increase. When these variables are sufficiently large,  $I_{s1}$  and  $I_{s2}$  suppress the action potentials, and the cell returns to a hyperpolarized silent state. Model equations were solved numerically using the CVODE algorithm implemented in the XPPAUT software package (Ermentrout, 2002). Bifurcation diagrams were also computed with XPPAUT. Computer codes are available as freeware from www.math.fsu.edu/~bertram/software/islet.

When the  $I_{s1}$  conductance  $(g_{s1})$  is large, bursting is driven by  $s_1$ . Since  $\tau_{s1} = 1$  sec, the burst period is only a few seconds (Fig. 1A). For this fast bursting,  $s_2$  is almost constant, while  $s_1$ varies with a sawtooth pattern (Fig. 1B). In fact, if  $s_2$  is clamped at its average value, the bursting continues almost unaltered. To analyze the bursting, we performed a fast/slow analysis. The  $s_1$  variable is treated as the bifurcation parameter for the fast subsystem with  $s_2$ held constant at its average value. The fast subsystem bifurcation diagram, or z-curve, is shown in Fig. 2A. The stationary solutions form the z-curve. The solid part of the curve represents the stable solutions, and the dashed part represents unstable solutions. There are two saddle node bifurcations (triangle) where the curve folds. A branch of periodic solutions emerges from a Hopf bifurcation (circle) and represents action potentials (both minimum and maximum voltage are indicated). The periodic branch terminates at an infinite-period homoclinic bifurcation (square). The burst trajectory is superimposed over the z-curve showing the system dynamics, with  $s_1$  no longer treated as a parameter. The z-curve now plays the role of a generalized V-nullcline, and the  $s_1$ -nullcline is added to the figure. During the silent phase, the burst trajectory follows the bottom of the z-curve, moving leftward, since it is to the right of the  $s_1$ -nullcline. Once the lower knee is reached, the phase point moves to the only attractor, the periodic branch that represents the spiking phase of the burst. Since it is now to the left of the  $s_1$ -nullcline, it moves rightward until the homoclinic bifurcation is reached, at which time the phase point returns to the bottom branch to restart the silent phase.

When  $g_{s1}$  is lowered, variations in  $I_{s1}$  are insufficient to move the system between active and silent phases. This requires a contribution from  $I_{s2}$ . During the active phase,  $s_2$  slowly increases and slowly decreases during the silent phase (Fig. 1D), increasing and decreasing the current, respectively. Medium bursting is produced (Fig. 1C) with period influenced by both  $s_1$  and  $s_2$ . If  $s_2$  is clamped, the bursting is replaced by a steady-state solution or

trajectory then enters the silent phase. While in the silent phase, the phase point gets caught at the intersection of the bottom branch of the z-curve with the  $s_1$ -nullcline. The burst period is determined both by the time required for the phase point to move along the z-curve (controlled by the  $s_1$  dynamics) and the time required to translate the z-curve and periodic branch back and forth (controlled by the  $s_2$  dynamics). Our analysis aims to quantify these contributions.

Further reduction in  $g_{s1}$  leads to a further increase in the burst period. Bursting is now solely driven by  $s_2$  (Fig. 1E). Since  $\tau_{s2} = 2$  min, the burst period is nearly 2 min. The  $s_1$  time course is a square wave, characteristic of the fast variable in a relaxation oscillation. In fact,  $s_1$  is part of the fast subsystem. While  $s_1$  plateaus at its highest value during the active phase,  $s_2$  varies with a sawtooth pattern (Fig. 1F). In the extreme cases where  $g_{s1}$  is very big or very small, we can say that bursting is fast or slow based on the period of oscillations. However, it is difficult to define precisely where the transition occurs from fast to medium and from medium to slow bursting. Using the method of quantification described later, we will be able to define these transitions.

The generic phantom bursting model can be reduced to a phantom relaxation oscillator by making the activation kinetics of the delayed rectifier current instantaneous. That is,  $n = n_{\infty}(V)$  in Eq. 5. This replaces the spikes of an active phase of bursting with a depolarized voltage plateau. When  $g_{s1}$  is large, a fast relaxation oscillation is produced, which is driven by  $s_1$ . This can be analyzed in the  $s_1$ -V plane, with  $s_2$  held constant at its average value. In Fig. 3A, the solid z-shaped curve is the V-nullcline, given by:

$$s_1 = -\frac{I_{Ca}(V) + I_{\kappa}(V) + I_{L}(V)}{g_{s1}(V - V_{\kappa})} - \frac{g_{s2}}{g_{s1}}s_2.$$
(11)

The  $s_1$ -nullcline is the dotted curve in Fig. 3A and is given by:

$$s_1 = s_{1\infty}(V). \tag{12}$$

As in Fig. 2A the  $s_1$ -nullcline intersects the *z*-shaped curve, now the *V*-nullcline, on the middle branch, and the full-system equilibrium is unstable. The phase point travels along the bottom branch during the silent phase and the top branch during the depolarized phase. This is a standard relaxation oscillation (Fig. 3C). When  $g_{s1}$  is reduced the relaxation oscillation is driven by both  $s_1$  and  $s_2$  (Fig. 3D). As in Fig. 2B, in the  $s_1$ -*V* plane, the *V*-nullcline moves with changes in  $s_2$  to end the active and silent phases (Fig. 3B). In fact, Eq. (11) makes it evident that increasing  $s_2$  translates the *V*-nullcline leftward.

#### 3 Method of Quantification

We now develop a method for quantifying the contribution that each slow variable makes to the active and silent phases of the oscillation. We begin with the phantom relaxation oscillation and rely on the fact that activity is terminated and restarted as the slow variables increase during the active phase and decrease during the silent phase. The method is illustrated in Fig. 4. At the beginning of the active phase (AP) of a relaxation oscillation the

time constant,  $\tau$ , for one of the slow variables is increased by  $\delta\tau$ . This slows down the slow variable, so if slow variation of this variable contributes to the termination of the active phase, the active phase should increase by  $\delta$ AP. The larger the slow variable's contribution to the active phase duration, the larger  $\delta$ AP. If the variable has no influence on the active phase duration, then slowing it down will give  $\delta$ AP =0. The procedure is repeated for the second slow variable and the silent phase (SP). Note that we perturb only one variable's time constant at a time, at the very beginning of a phase, and only look at how this perturbation affects that phase. We do not let the system equilibrate after a time constant is changed, because then both variables will vary over a slightly different range than before the perturbation. This change in the range of variations of the slow variables may also lead to a change in the AP and SP durations, compounding the effect of the original perturbation in time constant.

We consider the system to be in the active phase when V > -40 mV and to be in the silent phase when V < -40 mV. Now, a measure of the contribution of  $s_1$  to the duration of the active phase is given by  $\delta AP_{s1}/\delta \tau_{s1}$ , an approximation to the derivative of the AP duration with respect to  $\tau_{s1}$ . Then, the normalized contribution of  $s_1$  to the AP duration  $(C_{s1}^{s1})$  is given

with respect to  $\tau_{s1}$ . Then, the normalized contribution of  $s_1$  to the AP duration  $(C_{AP})$  is given by

$$C_{AP}^{s1} = (\delta A P_{s1} / \delta \tau_{s1}) (\tau_{s1} / A P).$$
<sup>(13)</sup>

With  $C_{AP}^{s1}$  defined in this way, if  $s_1$  is the only slow variable contributing to the duration of the AP, an increase in  $\tau_{s1}$  of 5% so that  $\delta \tau_{s1}/\tau_{s1} = 0.05$  would result in an increase in AP of 5% so that  $\delta AP_{s1}/AP = 0.05$ , and therefore  $C_{AP}^{s1} = 1$ . If  $s_1$  has no effect on the active phase duration, then  $C_{AP}^{s1} = 0$ . In most cases both  $s_1$  and  $s_2$  will contribute, so  $0 < C_{AP}^{s1} < 1$ . Similarly, we can quantify the effect that  $s_1$  has on the silent phase duration by increasing the time constant at the beginning of the SP and measuring the effect that it has on the SP. Thus, we have

$$C_{sP}^{s1} = (\delta SP_{s1} / \delta\tau_{s1}) (\tau_{s1} / SP).$$
<sup>(14)</sup>

Likewise, we use the same technique on the  $s_2$  variable to obtain:

$$C_{AP}^{s2} = (\delta AP_{s2} / \delta \tau_{s2}) \left( \tau_{s2} / AP \right) \tag{15}$$

$$C_{sp}^{s2} = (\delta SP_{s2} / \delta \tau_{s2}) (\tau_{s2} / SP).$$
(16)

By comparing  $C_{AP}^{s1}$  to  $C_{AP}^{s2}$  and  $C_{SP}^{s1}$  to  $C_{SP}^{s2}$ , we can evaluate the respective contributions of  $s_1$  and  $s_2$  to AP and SP durations. This is facilitated by using a measure we call the dominance factor (DF) for each phase:

$$DF_{AP} = \frac{C_{AP}^{s1} - C_{AP}^{s2}}{\sqrt{(C_{AP}^{s1})^2 + (C_{AP}^{s2})^2}}, DF_{SP} = \frac{C_{SP}^{s1} - C_{SP}^{s2}}{\sqrt{(C_{SP}^{s1})^2 + (C_{SP}^{s2})^2}}.$$
(17)

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Defined in this way, the dominance factor has a trigonometric interpretation in the  $C^{s1} - C^{s2}$ 

plane (Fig. 5). The length of the vector  $\vec{C} = (C^{s1}, C^{s2})$  is  $|\vec{C}| = \sqrt{(C^{s1})^2 + (C^{s2})^2}$ . Then,  $C^{s1} = |\vec{C}| \cos \theta$ ,  $C^{s2} = |\vec{C}| \sin \theta$ , and from Eq. (17), DF =  $\cos \theta$ -sin  $\theta$ . When  $s_1$  dominates  $\theta = 0$  and DF = 1, and when  $s_2$  dominates  $\theta = \frac{\pi}{2}$  and DF = -1. For all  $\theta$  between 0 and  $\frac{\pi}{2}$ , DF is between these two extremes. The DF can go outside of this range if either  $C^{s1}$  or  $C^{s2}$  is negative (as discussed later).

Figure 6 shows the results of applying this method of quantification to the phantom relaxation oscillator for various values of  $g_{s1}$ . Here and in other figures, we use  $\delta \tau / \tau = 1$ . The rationale for using this somewhat large value is discussed later for the case of phantom bursting. Fast oscillations occur with high values of  $g_{s1}$ , while slow oscillations occur with low values of  $g_{s1}$  (Fig. 6A). The C values for various values of  $g_{s1}$  are shown in Fig. 6B.  $C_{sp}^{s1}$ (open circle) and  $C_{sp}^{s1}$  (closed circle) start near 0 for small values of  $g_{s1}$ , then increase to 1 as  $g_{s1}$  increases, while  $C_{AP}^{s2}$  (open triangle) and  $C_{SP}^{s2}$  (closed triangle) start at 1 and decrease to 0. Figure 6C shows  $DF_{AP}$  (open circles) and  $DF_{SP}$  (closed circles). For low values of  $g_{s1}$ , DF is close to -1 indicating that  $s_2$  is the variable driving the oscillations, which therefore have a large period (Fig. 6A). For high values of  $g_{s1}$ , DF is close to 1 indicating that  $s_1$  is the variable driving the oscillations, which have a short period since  $\tau_{s1}$  is small. It also shows that the switch between  $s_1$ -driven oscillations and  $s_2$ -driven oscillations occurs near  $g_{s1} = 20$ pS. However, the switch of control does not occur simultaneously for the AP and the SP. When  $g_{s1} = 20$  pS the AP is driven primarily by  $s_2$ , while the SP is primarily driven by  $s_1$ . This difference in contribution to the AP and SP between  $s_1$  and  $s_2$  is not simply due to the difference in their time constants. The difference in their time constants leads to  $s_1$ 's dominance of fast bursting and  $s_2$ 's dominance of slow bursting. Rather, it is due to the difference in their activation. Figure 3B illustrates that the phase point gets stuck in the AP shown by the vertical trajectory as  $s_2$  moves the V-nullcline leftward, ending the AP. The slow increase in  $s_2$  moves the V-nullcline leftward, while the phase point is at the upper intersection of this nullcline with the  $s_1$ -nullcline. Once this intersection disappears, the phase point moves vertically downward, and then to the left. While it is not as clear for the SP,  $s_2$  moves the lower knee before the phase point reaches it. Therefore, the phase point does not get stuck. So, the contribution of  $s_2$  to the termination of the SP is minimal, while  $s_2$  controls the termination of the AP.

Figure 7 shows the results of applying the method of quantification to the phantom bursting model for various values of  $g_{s1}$ . These results are similar to those obtained for the relaxation oscillator (Fig. 6). Thus, we see a similar transition between  $s_1$ -dominated and  $s_2$ -dominated dynamics, except that now the switch of control between  $s_1$  and  $s_2$  occurs at a lower value of  $g_{s1}$ . At  $g_{s1} = 8$  pS, the control is mixed;  $s_1$  controls the length of the SP, while  $s_2$  controls the length of the AP. While a small  $\delta \tau$  works well with the relaxation oscillator, it can produce very jagged contribution (C) curves when used with bursting. This is because asmall  $\delta \tau$  can lead to the addition of an extra spike in some cases, but not others. The curves are smoother with a larger  $\delta \tau$  ( $\delta \tau = \tau$ ). We verified that, in the case of the phantom relaxation oscillator, the dominance factor curves are similar for  $\delta \tau = 0.05\tau$  and  $\delta \tau = \tau$  (used in Figs. 6,7,9).

During the active phase of a fast burst  $s_1$  increases monotonically (Fig. 1B). During the active phase of a medium burst  $s_1$  first increases, but then decreases (Fig. 1D and Fig. 8). This decrease occurs when the trajectory is "stuck" near the end of the periodic branch. The value of  $s_1$  averaged over a spike declines as  $s_2$  rises and shifts the *z*-curve leftward, since now the spike spends a longer period of time at its nadir, underneath (and to the right of) the  $s_1$ -nullcline In other words,  $s_1$  declines due to the decrease in spike frequency near the end

of the active phase. If  $\tau_{s2}$  is now increased, the duration of the decreasing phase of  $s_1$  will be extended. This extra decrease in  $s_1$  provides an extra increase in the AP duration. Figure 8 is an exaggerated picture of the decrease in  $s_1$  leading to burst prolongation. As  $s_1$  declines, the hyperpolarizing current  $I_{s1}$  also declines, tending to increase the AP duration. As a result,

 $C_{AP}^{s2}$ >1, as seen in Fig. 7B for  $g_{s1}$  near 8 pS. This does not occur in the relaxation oscillator since there are no spikes to bring the trajectory to the right of the  $s_1$ -nullcline.

There are also cases during medium bursting where  $C_{AP}^{s1} < 0$ , so that increasing  $\tau_{s1}$  decreases the active phase duration. This is again due to the decline in  $s_1$  during the latter part of the active phase in medium bursting. If  $\tau_{s1}$  is increased,  $s_1$  rises more slowly during the active phase and enters its declining phase much later in the burst. Once it enters the declining phase it declines more slowly. Together, the active phase prolongation, due to the  $s_1$  decline during the active phase, is reduced. The end result is that slowing down  $s_1$  makes the active

phase shorter, so  $C_{AP}^{s1} < 0$ .

We can use the dominance curves to provide, for the first time, a quantitative distinction between the types of phantom bursting. For some small  $\varepsilon > 0$  (we choose  $\varepsilon = 0.15$ ), bursting can be defined as "fast" if  $DF_{SP}$ ,  $DF_{AP} > 1-\varepsilon$ . Bursting is "slow" if  $DF_{SP}$ ,  $DF_{AP} < -(1-\varepsilon)$ . Bursting is "medium" if  $-(1-\varepsilon) < DF_{SP}$ ,  $DF_{AP} < 1-\varepsilon$ . From Fig. 7, slow bursting occurs for  $g_{s1} < 6.75$  pS; medium bursting occurs for 6.75 pS  $< g_{s1} < 11$  pS;and fast bursting occurs for  $g_{s1} > 11$  pS.

In computing the dominance curves in Fig. 7,  $g_{s1}$  was varied to produce the different types of bursting. For fast bursting, both  $DF_{AP}$  and  $DF_{SP}$  were close to 1, while for slow bursting both  $DF_{AP}$  and  $DF_{SP}$  were close to -1. This indicated that either  $s_1$  or  $s_2$  controlled both phases of the burst. On the other hand, for medium bursting ( $g_{s1} \approx 8$  pS), we can have  $DF_{AP}$ < 0 and  $DF_{SP} > 0$ , showing that each slow variable controls one phase (Fig. 7C). If the dominance curves are computed by varying  $g_{s2}$  and keeping  $g_{s1}$  constant at 8.5 pS, one slow variable controls the active phase while the other controls the silent phase over most of the range (Fig. 9). As  $g_{s2}$  is increased, the burst period decreases (Fig. 9A). Figure 9B shows the DF values for a range of values of  $g_{s2}$ . For low values of  $g_{s2}$ ,  $DF_{AP}$  is close to -1 and  $DF_{SP}$ is close to 1 indicating that  $s_2$  drives the active phase, while  $s_1$  drives the silent phase. In other words, as  $g_{s2}$  is varied, the z-curve shifts from left to right, changing which phase the trajectory gets stuck in. Therefore, the DF curves intersect. Figure 10C shows the fast subsystem bifurcation diagram for  $g_{s2} = 20$  pS. Here the phase point gets stuck in the AP and has to wait for  $s_2$  to move the z-curve to the left, terminating the AP. The phase point does not get stuck in the SP. However, for high values of  $g_{s2}$  the  $DF_{AP}$  is close to 1 and  $DF_{SP}$  is close to -1 indicating that  $s_1$  is the variable driving the AP, while  $s_2$  is driving the SP. Figure 10A shows the fast subsystem bifurcation diagram for  $g_{s2} = 100$  pS. Now, the phase point gets stuck in the SP, and has to wait for  $s_2$  to terminate the SP.Figure 10B shows the fast subsystem bifurcation diagram for  $g_{s2} = 40$  pS. Now, the phase point never gets stuck;  $s_1$  is in control of both the AP and SP. Thus, at extreme values of  $g_{s2}$  each slow variable contributes to a phase of the burst, while with  $g_{s2}$  held constant and  $g_{s1}$  varied, a single variable controls both phases at the extreme  $g_{s1}$  values. By applying our definition of medium bursting, as  $g_{s2}$  varies, the bursting goes from medium to fast then back to medium.

#### 4 Resetting

When a relaxation oscillator is perturbed from the silent (active) to the active (silent) phase half way through the silent phase, the immediately following active phase is reduced. This is also true for a bursting oscillation driven by a single slow variable. When there is more than one slow variable, the resetting properties can be different. In fact, if the dominance curves

are appropriate, full-length resets may be achieved for both the phantom relaxation oscillator (not shown) and the phantom burster (shown below).

The condition required for a full-length reset is that one slow variable,  $s_1$ , controls one phase of the oscillation while the second slow variable,  $s_2$ , controls the other phase. This can be achieved by adjusting  $g_{s2}$  so that the dominance curves become separated as in Fig. 9C. In Fig. 9C, for low values of  $g_{s2}$ ,  $DF_{AP} \approx -1$ , while  $DF_{SP} \approx 1$ . This means that  $s_1$  is in control of the SP, while  $s_2$  controls the AP. For high values of  $g_{s2}$ ,  $DF_{AP} \approx -1$ , so  $s_1$  is in control of the AP and  $s_2$  controls the SP.

When  $s_2$  is in control of the active phase,  $s_1$  reaches its maximum value very early in the AP, while  $s_2$  increases monotonically. If the model cell is reset before the end of the active phase, and if  $s_1$  controls the silent phase duration (as in Fig. 9C, for  $g_{s2} = 27$  pS), then the silent phase will have a full duration. That is,  $s_1$  will be starting at the same value almost regardless of when the reset occurs in the active phase. This is shown in Fig. 11A. However, when resetting half-way through the silent phase, the induced active phase is reduced (Fig. 11B). This occurs because  $s_2$  is in control of the active phase;  $s_2$  is between its minimum and maximum values when the reset occurs, so during the subsequent AP it need only travel a portion of the distance required to terminate the AP. In this case, silent-active resetting is phase dependent (Fig. 11C); the duration of the induced AP is closer to the unperturbed AP duration the longer the system is in the SP before the reset. On the other hand, active-silent resetting is approximately phase independent, if resetting occurs after  $s_1$  reaches its maximum value (Fig. 11D). That is, a reset very early in the active phase does not result in a full-length silent phase, but resets applied at most points during the AP do produce nearly full-length silent phases as in Fig. 11C.

When  $s_2$  is in control of the silent phase, and  $s_1$  controls the active phase (e.g., for  $g_{s2} = 97$  pS, Fig 9C),  $s_1$  reaches its minimum value early in the silent phase, while  $s_2$  decreases monotonically. If the model cell is reset before the end of the silent phase, then the active phase will have a full duration. That is,  $s_1$  will be starting at the same value almost regardless of when the reset occurs in the silent phase. This is shown in Fig. 12B. However, when resetting half-way through the active phase, the induced silent phase is reduced (Fig. 12A). This occurs because  $s_2$  is in control of the silent phase;  $s_2$  is between its minimum and maximum values when the reset occurs, so during the subsequent SP it need only travel part of the distance required to terminate the SP. In this case, silent-active resetting is nearly phase independent (Fig. 12C) and the active-silent resetting is phase dependent (Fig. 12D).

#### **5 Bidirectional Full-Length Resets**

So far we have described full-length resets in one direction. By changing model parameters, we can get either active-silent or silent-active full-length resets. Such full-length resets are consistent with experimental data from pancreatic islets (Cook *et al.*, 1981). However, in (Cook *et al.*, 1981) there was an example in which full-length resets occurred in both directions (*bidirectional full-length resetting*) in an islet, which cannot be accounted for with the model in its current form. However, with a few modifications the model can reproduce this data. The idea is to design the system so that  $s_1$  controls the active phase and  $s_2$  controls the silent phase. Then, make the  $s_1$  and  $s_2$  time scales voltage dependent, so that  $s_1$  is slow during the active phase and fast during the silent phase, and vice-versa for  $s_2$ . Time scales that achieve this are

$$\tau_{s1}(V) = \tau_{s1,min} + \tau_{s1,max}\phi_1(V), \tau_{s2}(V) = \tau_{s2,min} + \tau_{s2,max}\phi_2(V)$$
(18)

with

$$\phi_1(V) = \frac{1}{1 + \exp[-(40 + V)/3]}, \\ \phi_2(V) = \frac{1}{1 + \exp[(42 + V)/3]}$$
(19)

and  $\tau_{s1,min} = 100$  msec,  $\tau_{s2,min} = 100$  msec,  $\tau_{s1,max} = 10$  sec, and  $\tau_{s2,max} = 10$  sec.

With these changes to the model, we get bursting as shown in Fig. 13A. During the active phase  $\tau_{s1} \approx 10$  sec, while during the silent phase  $\tau_{s1} \approx 100$  msec (Fig. 13B). The  $s_2$  time scale,  $\tau_{s2}$ , is the opposite (Fig. 13B). Thus,  $s_1$  quickly resets to its minimum value during the silent phase, while  $s_2$  quickly achieves its maximum value during the active phase (Fig. 13C). During the active phase,  $s_2$  almost instantaneously achieves its maximum value, while  $s_1$  rises monotonically. Likewise, during the silent phase,  $s_1$  almost achieves its minimum value instantaneously, while  $s_2$  decreases monotonically (Fig. 13C).

bidirectional full-length resetting is now possible. Since  $s_1$  quickly reaches its minimum value during the silent phase, when reset to the active phase  $s_1$  has to rise the normal amount to reach its maximum value, yielding a full-length silent-active reset (Fig. 14A). Since  $s_2$  quickly reaches its maximum value during the active phase, when reset to the silent phase  $s_2$  must decrease the normal amount to end the silent phase, producing a full-length active-silent reset (Fig. 14B).

#### 6 Discussion

We developed a measure, the dominance factor, to quantify the contributions of two slow variables to a phantom relaxation oscillation and phantom bursting. This is useful for determining which slow variable controls each phase of the oscillation or whether the two work together. The dominance factor also allows us to quantitatively categorize bursting into fast, medium, and slow (Fig. 7). In the past, it has been difficult to distinguish between medium and slow bursting in a quantitative way. With the dominance factor, it is also easy to see when the control of the active and silent phases is shifted from one slow variable to the other as a parameter is varied (Figs. 6, 7, 9).

The method used here was previously developed in the context of a neural relaxation oscillator with two types of negative feedback variables, one divisive and one subtractive (Tabak *et al.*, submitted). Here,  $s_1$  and  $s_2$  are both subtractive, so the analysis developed in the earlier paper predicts that the contribution of each variable should be the same for the active and silent phase. Also, the contributions of  $s_1$  and  $s_2$  to both the active and silent phase should depend on the inverse ratio of their time constants. Given the difference in time scales between  $s_1$  and  $s_2$ ,  $s_1$  should control both the active and silent phase. We show here that this is true only for fast bursting. As  $g_{s1}$  is lowered, the contributions vary quantitatively, but at some point a qualitative change occurs due to the phantom effect. That is, the system becomes stuck in the active or silent phase and has to wait for  $s_2$  controls the duration of that phase. Also, unless parameters are tuned precisely, as  $g_{s1}$  or  $g_{s2}$  is varied, the system will first be stuck in one phase, but not the other. Thus, the contribution of each variable to the active and silent phases so  $s_2$  will control the duration of both phases.

The method assumes that both slow negative feedback variables are the only variables responsible for burst termination and that they vary monotonically during each phase of the burst. In that case, all the C values should be positive and below 1. Also,  $C^{s1} + C^{s2} = 1$ . We have good agreement to this rule in the relaxation case (Fig. 6), but not in the bursting case (Fig. 7). This is because during medium bursting  $s_1$  does not vary monotonically. Instead, it

quickly reaches a high value during the active phase, then slowly decreases since the spikes can push the V-s<sub>1</sub> trajectory below the s<sub>1</sub>-nullcline. This decrease of s<sub>1</sub> slows down the termination of the active phase, instead of s<sub>1</sub> steadily contributing to its termination. This is why for some  $g_{s1}$  values we get  $C_{AP}^{s1} < 0$  or  $C_{AP}^{s2} > 1$ . Nevertheless, the results are qualitatively similar to those obtained in the relaxation case. Finally, the method should in principle be used with small  $\delta \tau$ . Unfortunately, the spikes make the active phase duration discontinuous, so the  $C_{AP}$  values become very variable when  $\delta \tau$  is small. To avoid this problem, we have used  $\delta \tau = \tau$  after having checked that in the relaxation case we obtain quantitatively similar results with  $\delta \tau = 0.05\tau$  and  $\delta \tau = \tau$ . Thus, the method that was originally developed for a relaxation oscillation can be extended to bursting where a fast oscillation (the spikes) is superimposed on a slow relaxation rhythm.

It is unlikely that the approach used here to calculate dominance factors can be applied experimentally. This fact argues for the utility of mathematical models for biological systems. It may be known from experiments that two or more slow processes are involved in the burst generation, but without a model it is difficult or impossible to know how much each variable contributes to bursting. This is the case with pancreatic  $\beta$ -cells, where slow variables such as the cytosolic  $Ca^{2+}$  concentration, the  $Ca^{2+}$  in the endoplasmic reticulum, the ATP/ADP ratio, and slow inactivation of  $Ca^{2+}$  currents have all been postulated to contribute to bursting (Bertram & Sherman, 2000). With the development of models containing some or all of the slow processes, and the technique that we describe here to quantify slow variable dominance, it becomes possible to identify the key processes driving the bursting. In fact, we are currently applying this technique to a  $\beta$ -cell model that contains many of the slow variables listed above, with the challenge that there are more than two slow variables.

Another extension of the simple  $\beta$ -cell model would be the inclusion of channel noise. This would add a stochastic element to the voltage differential equation (1). It has been shown previously that phantom bursting is sensitive to noise, particularly for the case of slow bursting (Pedersen, 2007). In this case, active and silent phases can be significantly shorter than predicted by the deterministic model, since now the noise, rather than the slower of the two slow variables, is what terminates the active/silent phase. Therefore, the contribution to bursting of the  $s_2$  variable predicted by the dominance factor analysis of the deterministic model would be overstated for the stochastic model. In other words, we expect that dominance factors for the stochastic model would be closer to 1 than those for the deterministic model in the case of single  $\beta$ -cells, which can be very noisy. However, this effect would depend on noise amplitude, which is small in islets.

One application of the dominance factor is in the determination of parameter values that allow phase-independent resetting. Such resetting was documented in islets nearly 30 years ago (Cook *et al.*, 1981). An earlier model, similar to our model for bidirectional full-length resets, was able to account for this (Smolen & Sherman, 1994). However, that model was not a phantom bursting model and thus the burst period was constrained to a relatively narrow range of values. The present model possesses both the desired (but unidirectional) full-length reset properties (Figs. 11, 12), and can produce the wide range of oscillation periods that is characteristic of pancreatic  $\beta$ -cells (Bertram *et al.*, 2000).

bidirectional full-length resets cannot be produced with the phantom bursting model in its current form. We showed how the model can be changed to produce this type of resetting (Fig. 14), but with the changes the model is no longer a phantom bursting model. Since bidirectional full-length resets have been documented in at least one pancreatic islet (Cook *et al.*, 1981), this calls into question the validity of the phantom bursting model as a description of islet electrical activity. One possibility is that the slow processes (such as

 $Ca^{2+}$  dynamics in the cytosol and the endoplasmic reticulum, and the ATP/ADP ratio) work together to produce phantom bursting in some islets, but not others. It is also possible that with a more sophisticated phantom bursting model (Bertram & Sherman, 2004) bidirectional full-length resets can be achieved, although we have not yet found this to be true.

Our analysis predicts unidirectional full-length resets for medium bursting islets, but not for slow or fast bursting islets. Active-silent full-length resets should be found in islets with a long active phase and short silent phase (Fig. 11), while, silent-active full-length resets should be found in islets with a short active phase, but a long silent phase (Fig. 12). For fast and slow bursting only one variable controls the duration of both phases, so full-length resets should not occur. At the time that Cook et al. performed their islet resetting experiments (Cook *et al.*, 1981) slow islet oscillations had not appeared in the literature, and resetting of only medium-period (15–30 sec) oscillations were attempted. Since the long period of the now commonly reported slow islet oscillations would make resetting data easier to interpret than in the case of fast bursting, we encourage investigators to continue the work of Cook and colleagues by examining islet resetting of slow bursters.

#### Acknowledgments

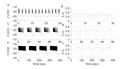
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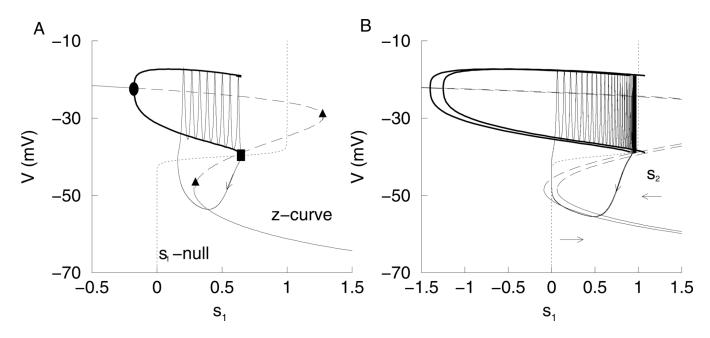
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#### Figure 1.

(A) Fast bursting,  $g_{s_1} = 20$  pS. (B) Bursting is driven by  $s_1$  (dashed), while  $s_2$  is nearly constant (solid). (C,D) Medium bursting driven by both  $s_1$  and  $s_2$ ,  $g_{s_1} = 7$  pS. (E, F) Slow bursting, is driven by  $s_2$ ,  $g_{s_1} = 2$  pS.

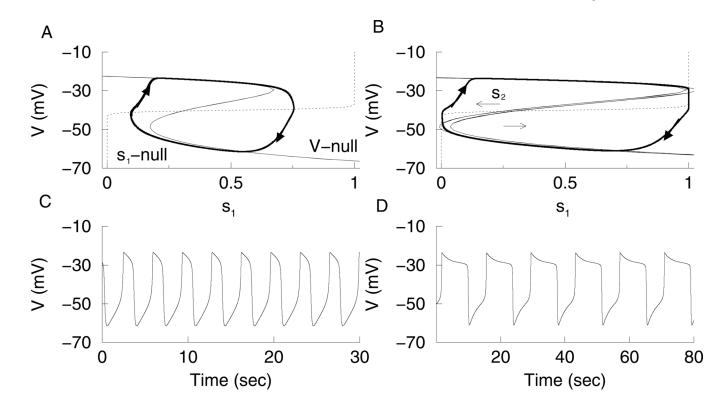
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#### Figure 2.

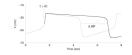
(A) Fast subsystem bifurcation diagram of fast bursting ( $g_{s1} = 20 \text{ pS}$ ) with  $s_1$  as the bifurcation parameter and  $s_2 = 0.436$ . The  $s_1$ -nullcline and burst trajectory are superimposed on the bifurcation diagram. The circle represents a Hopf bifurcation, the square represents a homoclinic bifurcation, and the triangles represent saddle node bifurcations. (B) Fast/slow analysis of medium bursting ( $g_{s1} = 7 \text{ pS}$ ). There are two bifurcation diagrams with  $s_1$  as the bifurcation parameter. The curve on the left has  $s_2$  fixed at its maximum value (0.633) achieved during the bursting, while the curve on the right has  $s_2$  fixed at is minimum value (0.600). The burst trajectory is superimposed on the diagram. Arrows indicate direction of movement of the *z*-curve driven by variations in  $s_2$ .

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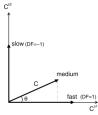
#### Figure 3.

Phase plane analysis of fast and medium relaxation oscillations. (A) The V-nullcline (*z*-shaped curve) and  $s_1$ -nullcline (dotted curve) for fast oscillations with  $g_{s1} = 40$  pS and  $s_2$  fixed at 0.436. The trajectory (heavy solid curve) follows the upper and lower branches of the V-nullcline. (B) The V-nullcline and  $s_1$ -nullcline for medium oscillations with  $g_{s1} = 20$  pS. The V-nullcline on the left has  $s_2$  at it maximum value (0.619), while the V-nullcline on the right has  $s_2$  at its minimum value (0.591). (C) Fast relaxation oscillations driven by  $s_1$ . (D) Medium relaxation oscillations driven by  $s_1$  and  $s_2$ .



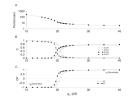
#### Figure 4.

Measuring the effect of a slow variable on the duration of the active phase. The time constant of the slow variable,  $\tau$ , is increased by  $\delta \tau$  at the beginning of the active phase (arrow). This causes the slow variable to slow down and the active phase duration to increase by  $\delta AP$  (bold curve).



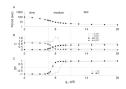
#### Figure 5.

Interpretation of the dominance factor,  $DF = \cos\theta - \sin\theta$ . When  $\theta = 0$ , DF = 1 and the oscillation is fast. When  $\theta = \frac{\pi}{2}$ , DF = -1 and the oscillation is slow. Medium frequency oscillations occur when  $\theta \in (0, \frac{\pi}{2})$  and  $DF \in (-1, 1)$ .



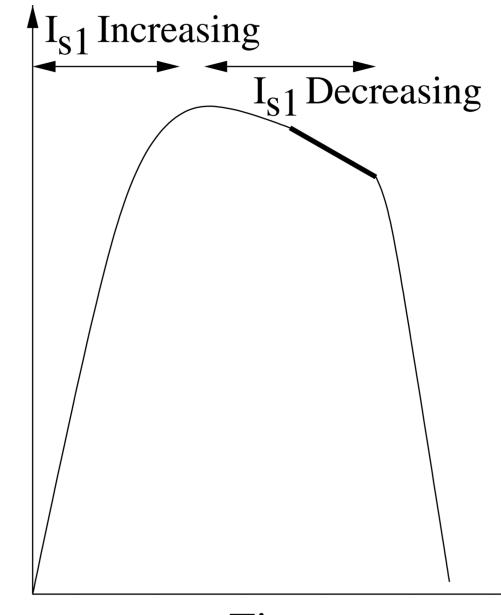
#### Figure 6.

Results of the quantification method on the phantom relaxation oscillator.  $\delta \tau = \tau$  here and in other figures that follow. The results obtained using  $\delta \tau = 0.05\tau$  are similar for the relaxation case. (A) Oscillation period decreases with  $g_{s1}$ . (B) C values for active and silent phases and for  $s_1$  and  $s_2$ . (C) For low values of  $g_{s1}$  the DF is close to -1 indicating that  $s_2$  is the variable driving slow oscillations, while for high values of  $g_{s1}$  DF is close to 1 indicating that  $s_1$  is the variable driving slow oscillations.



#### Figure 7.

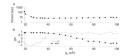
Results for phantom bursting. (A) Burst period decreases with  $g_{s1}$ . (B) C values for active and silent phases and for  $s_1$  and  $s_2$ . (C) For low values of  $g_{s1}$ , DF is close to -1 indicating that  $s_2$  is the variable driving slow bursting, while for high values of  $g_{s1}$ , DF is close to 1 indicating that  $s_1$  is the variable driving fast bursting. The type of bursting can be defined in terms of the dominance factors. **NIH-PA Author Manuscript** 



## Time

#### Figure 8.

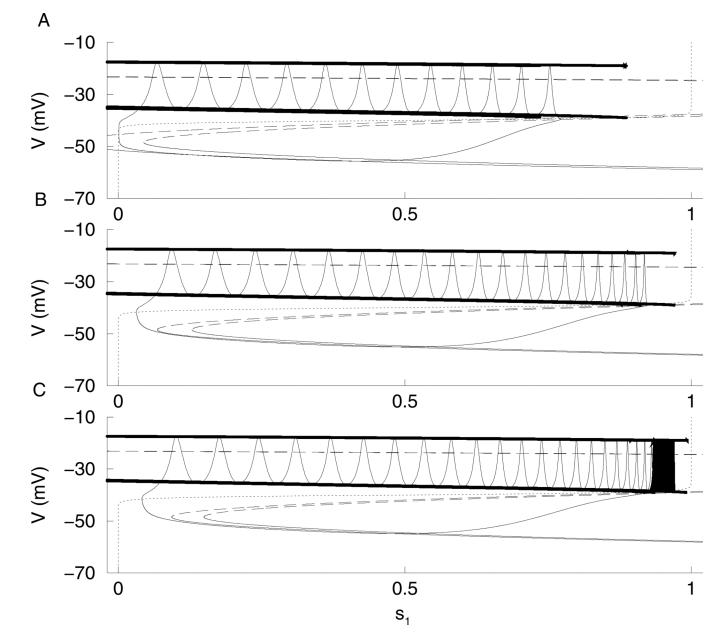
As  $s_1$  rises during an active phase,  $I_{s1}$  increases, which promotes the termination of the AP. However,  $I_{s1}$  starts to decline toward the end of the burst, leading to burst prolongation. Therefore, an increase in the time constant for  $s_2$  ( $\tau_{s2}$ ), leads to a longer decline in  $s_1$  (bold part of curve), which acts to increase AP duration.



#### Figure 9.

Results for phantom bursting with  $g_{s2}$  as the varying parameter and  $g_{s1} = 8.5$  pS. (A) Burst period decreases with  $g_{s2}$ . For  $g_{s2} < 19$  pS, the system spikes continuously. (B) C values for active and silent phases and for  $s_1$  and  $s_2$ . (C) For low values of  $g_{s2}$ ,  $DF_{AP}$  is close to -1 and  $DF_{SP}$  is close to 1 indicating that  $s_2$  drives the active phase, while  $s_1$  drives the silent phase. However, for high values of  $g_{s2} DF_{AP}$  is close to 1 and  $DF_{SP}$  is close to -1 indicating that  $s_1$  drives the SP.

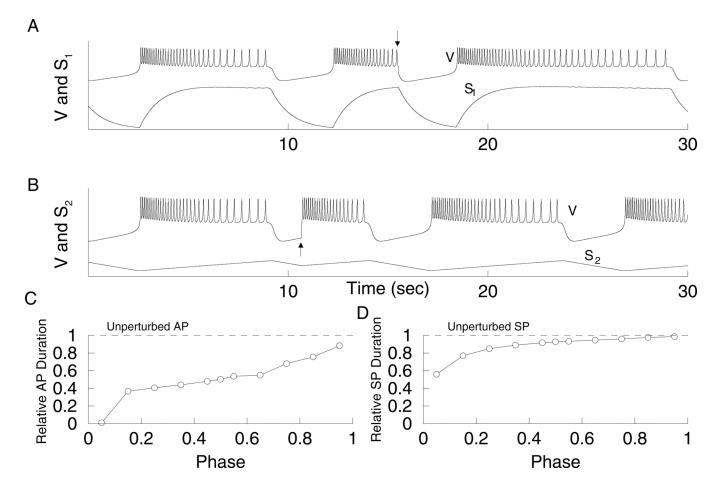
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#### Figure 10.

Bifurcation diagrams with  $g_{s1} = 8.5$  pS. The two dashed curves are the bifurcation diagrams for the extreme values of  $s_2$ . (A)For  $g_{s2} = 100$  pS, the phase point gets stuck in the SP. (B)For  $g_{s2} = 40$  pS, the phase point does not get stuck. (C)For  $g_{s2} = 20$  pS, the phase point gets stuck in the AP.

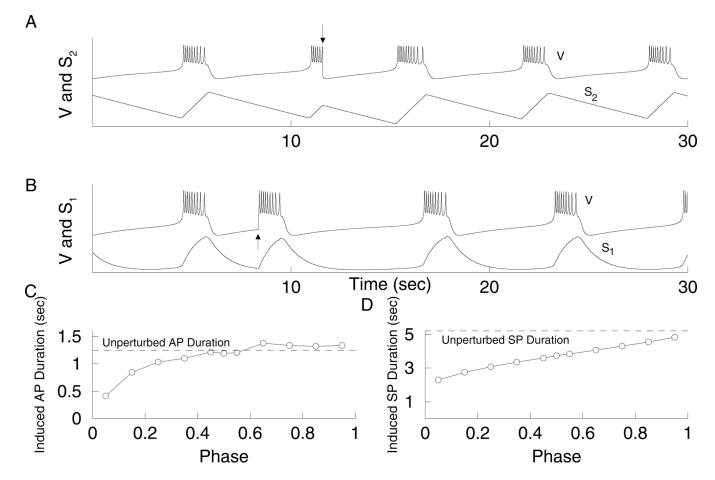
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#### Figure 11.

Resetting with  $g_{s2} = 27$  pS and  $g_{s1} = 8.5$  pS. In this case,  $s_1$  is in control of the SP (DF = 0.85), while  $s_2$  is in control of the AP (DF = -0.99). (A) Half-way through the AP the system was reset to the SP (arrow), which has full length.  $s_1$  has reached its maximum at the time of resetting (bottom curve). The V and  $s_1$  time courses have been scaled to facilitate comparison. (B) Half-way through the SP the system was reset to the AP (arrow), which is reduced.  $s_2$  is in the middle of decreasing to its minimum value at the time of resetting (bottom curve). (C) The duration of the induced AP is phase dependent. (D) The duration of the induced SP is close to the duration of the unperturbed SP if the resetting occurs after  $s_1$  reaches its maximum value.

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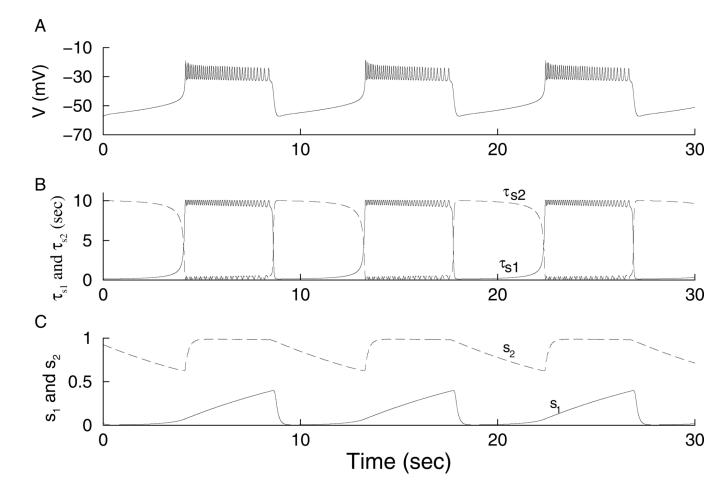


#### Figure 12.

Resetting with  $g_{s2} = 97$  pS and  $g_{s1} = 8.5$  pS. In this case,  $s_1$  is in control of the AP (DF = 0.79), while  $s_2$  is in control of the SP (DF = -0.68). (A) Half-way through the AP the system was reset to the SP (arrow), which is reduced.  $s_2$  is midway to its maximum value at the time of resetting (bottom curve). (B) Half-way through the SP the system was reset to the AP (arrow), which has full length.  $s_1$  has reached its minimum value at the time of resetting (bottom curve). (C) The duration of the induced AP is close to the width of the unperturbed AP if the resetting occurs after  $s_1$  reaches its minimum value. (D) The width of the induced SP is phase dependent.

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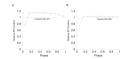
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#### Figure 13.

Bursting produced by the model with V-dependent  $s_1$  and  $s_2$  time scales (Eqs. 18, 19),  $g_{s1} = 16 \text{ pS}$  and  $g_{s2} = 30 \text{ pS}$ . (A) Voltage time course. (B)  $\tau_{s1} \approx 10$  sec during the active phase and  $\approx 100$  msec during the silent phase.  $s_1$  is in control of the active phase. (C)  $\tau_{s2} \approx 100$  msec during the active phase and  $\approx 10$  sec during the silent phase.  $s_2$  is in control of the silent phase.

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#### Figure 14.

Bidirectional resetting produced by the bursting model with V-dependent  $s_1$  and  $s_2$  time scales,  $g_{s1} = 16$  pS, and  $g_{s2} = 30$  pS. (A) Silent-active phase-independent resetting. (B) Active-silent phase-independent resetting.