

# Combining synaptic and cellular resonance in a feed-forward neuronal network

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CAMS Report 0506-37, Summer 2006

Center for Applied Mathematics and Statistics

**NJIT**

# Combining synaptic and cellular resonance in a feed-forward neuronal network

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June 1, 2006

## Abstract

We derive a mathematical theory to explain the subthreshold resonance response of a neuron to synaptic input. The theory shows how a neuron combines information from its intrinsic resonant properties with those of the synapse to determine the neuron's generalized resonance response. Our results show that the maximal response of a postsynaptic neuron can lie between the preferred intrinsic frequency of the neuron and the synaptic resonance frequency. We compare our theoretical results to parallel findings on experiments of the crab pyloric central pattern generator.

## 1 Introduction

Coherent rhythmic activity in the brain has been proposed to rely on the intrinsic tendency of neurons to produce a preferred subthreshold response to signals of a fixed frequency [2]. This property of neurons, known as membrane or cellular resonance, has been extensively studied in relationship to subthreshold network oscillations and signal processing [2][4]. Recently, it was demonstrated that synapses with short-term dynamics can also respond preferentially to signals arriving at a given frequency and have an attenuated response to signals at lower or higher frequencies [3]. This property is termed synaptic resonance [3] and can occur, for example, if the synapse demonstrates both short-term facilitation and depression. At this time little is known about the interaction between cellular and synaptic resonant properties in producing network activity.

Our analysis of neurons and synapses of a rhythmically active network, the pyloric central pattern generator of the crab *Cancer borealis*, shows that both neurons and synapses in this network demonstrate resonance properties. We focus on the connection between pacemaker and follower neurons and show that the follower neuron LP has a cellular resonance at a peak frequency that is distinct from the resonant frequency of the synapse it receives from

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the pacemaker neurons AB and PD. These distinct resonant frequencies point to a more general question: if cellular and synaptic resonances occur at different frequencies, do they interact to produce a net output resonance and, if so, at what frequency does this combined resonance occur?

In this study we examine how cellular and synaptic resonance properties interact to determine the resonant properties in a feed-forward network. Specifically, we examine the response of a postsynaptic cell that exhibits cellular resonance when it receives input from a resonant synapse. The network is driven by a fixed-frequency input that mimics input from a presynaptic cell. Of particular interest is the stimulus frequency that results in the largest subthreshold postsynaptic amplitude. We refer to this frequency as the generalized resonant frequency of the network. In this work we do not consider supra-threshold behavior of the postsynaptic cell.

## 2 The Model

We consider the response of an excitable cell to periodic synaptic input. The cell model used is the Hodgkin-Huxley (HH) model [1]. The motivation for our choice is that when the HH model is in its resting state, it exhibits subthreshold resonance. Sufficiently large synaptic inputs can result in supra-threshold behavior; however, we shall not consider such inputs in this study.

The Hodgkin Huxley model with synaptic input is given by

$$\begin{aligned} V' &= I_0 - m^3 h g_{Na}(V - E_{Na}) - n^4 g_k(V - E_k) - g_l(V - E_l) - g_{syn} s(V - E_{syn}) \\ x' &= \alpha_x(V)(1 - x) - \beta_x(V)x \end{aligned}$$

where  $x = h, m, n$ . The gating functions,  $\alpha$  and  $\beta$  and associated parameter values are standard [1]. The variable  $s$  represents the ratio of the total possible neurotransmitter being used by the synapse.

We use two different models for synaptic input  $s$ . In Model 1, we assume that synaptic strength is a Gaussian function of input frequency and use this theoretical construct to examine the effects of such a synapse on resonance properties of the postsynaptic neuron. In Model 2, we implement the frequency-dependent changes in synaptic strength by assuming that the synapse exhibits short-term facilitation and depression. This more realistic synapse is used to demonstrate the biological application of our theoretical construct.

Model 1 is given by the equation

$$s' = \alpha_{syn}(V_{pre})(\kappa(f) - s) - \beta_{syn}(V_{pre})s \quad (1)$$

where  $\kappa(f)$  denotes maximum synaptic transmission, determined by short-term synaptic dynamics.  $V_{pre}$  is the presynaptic voltage. The functions  $\alpha_{syn}$  and  $\beta_{syn}$  are Heaviside functions that satisfy  $\alpha_{syn}(V) = Heav(V - V_\theta)$  and  $\beta_{syn} = Heav(V_\theta - V)$ , where  $V_\theta$  is the synaptic release threshold. We assume that  $\kappa(f)$  is a Gaussian function:

$$\kappa(f) = e^{(1/f - 1/f_s)^2 / \sigma} \quad (2)$$

This is a dimensionless variable so that  $\sigma$  has units  $ms^2$ .

In Model 2, we use utilize the model of a depressing/ facilitating synapse given in [8]. The equations for this synapse are

$$\begin{aligned} x' &= \frac{z}{\tau_{rec}} - ux\delta(t - t_{sp}) \\ s' &= -\frac{s}{\tau_1} + ux\delta(t - t_{sp}) \\ z' &= \frac{s}{\tau_1} - \frac{z}{\tau_{rec}} \end{aligned} \tag{3}$$

where  $x$  is fraction of total synaptic resources available,  $s$  is the fraction of resources currently in use, and  $z$  is the fraction in the recovery state. The parameter  $\tau_1$  is the synaptic decay time and  $\tau_{rec}$  is the recovery time.  $t_{sp}$  is the time that a presynaptic spike occurs. To incorporate facilitation we let  $u$  evolve according to

$$u' = -\frac{u}{\tau_{facil}} + U(1 - u)\delta(t - t_{sp}) \tag{4}$$

where  $\tau_{facil}$  is the time it takes for the facilitation to wear off and  $U$  represents the contribution of each presynaptic spike to facilitation.

We quantitatively describe the synaptic and cellular resonance properties using a pair of functions of the presynaptic firing rate. The synaptic response curve (SRC) is denoted by  $\kappa(f)$ . This function determines the synaptic strength, relative to a maximum, for a given frequency. In Model 1 this function is explicitly given by (2). For Model 2,  $\kappa(f)$  is computed numerically. In particular, a presynaptic neuron activates the synapse given by equations (3) and (4) at different assigned frequency. The maximal steady-state value of  $s$  is then plotted for each value of  $f$  to obtain the SRC. The cellular resonance curve (CRC), denoted by  $A(K, f)$ , measures the amplitude of subthreshold oscillations.  $K$  is a frequency-independent stimulus strength and  $f$  is the input frequency. To obtain the CRC for the HH equations we determine the amplitude for a stimulus of fixed frequency and repeat the stimulus at different frequencies. The CRC is defined as the trough-to-peak amplitude of the response.

We define the generalized response curve (GRC) to be the amplitude of subthreshold oscillations in the postsynaptic cell as a function of presynaptic frequency. The function that describes this curve can be written as  $A(\kappa(f), f)$  where  $A(K, f)$  and  $\kappa(f)$  are the CRC and SRC, respectively.

**Biological Methods** Experiments were carried on adult male crabs (*C. Borealis*) purchased from local distributors (Newark, NJ). Details of experimental measurements are identical to those described in [5]. Spontaneous activity in the pyloric network was blocked by superfusion with saline containing 0.1  $\mu$ M TTX (Biotium, CA). (TTX does not block graded synaptic release in this system.) For measurements of synaptic resonance the presynaptic neuron was voltage clamped with two electrodes (TEVC) and a ZAP function waveform was used as the voltage command. The ZAP waveforms were injected with the software Scope on a PCI-6070-E board (National Instruments, Austin, TX). Data were acquired using the same software and board as well as a Digidata 1332A board with the PClamp 9.2 software (Molecular Devices, Union City, CA). Acquired data were saved as binary files and were analyzed with Readscope and PClamp software. Scope and Readscope are software developed in the Nadim lab (<http://stg.rutgers.edu/software/index.htm>). Peaks of the IRC, SRC and GRC in the biological recordings were obtained using FFT filter smoothing.

### 3 Results

**Experimental** The PD and LP neurons are members of the pyloric central pattern generator network in the crab *Cancer borealis*. The PD neurons are members of the pyloric pacemaker group and produce inhibitory synapses to the LP neuron. We measured the CRC in the LP neuron by injecting a ZAP current waveform and recording of the membrane voltage response. The CRC of the LP neuron peaked at 1.5 Hz indicating the membrane resonance frequency Fig. 1. For measurement of the SRC, the presynaptic PD neuron was voltage clamped with a ZAP function voltage waveform while the synaptic current was measured in the postsynaptic LP neuron which was voltage clamped at a holding potential of -60 mV. The SRC indicated a peak synaptic response at 0.68 Hz Fig. 1. The GRC was measured by voltage clamping the PD neuron using a ZAP waveform (as in the SRC measurement) but recording the postsynaptic response in the LP neuron in current clamp mode. Thus, the voltage response in the postsynaptic LP neuron was determined by both the synaptic input it received from the PD neuron and its own intrinsic dynamics. The GRC showed a peak response at 0.87 Hz (Fig. 1). These results showed distinct peak frequencies of the SRC and CRC and also that the GRC peak frequency fell between these two peaks.

**Modeling** We now determine how the location of the local maximum of the GRC depends on the properties of the CRC and SRC. The frequency of the presynaptic cell that elicits the largest postsynaptic response. This will be a local maximum for  $A(\kappa(f), f)$  which can be found by solving

$$\frac{d}{df}A(\kappa(f), f) = 0.$$

The local maximum of the SRC occurs at the so-called resonant frequency of the synapse which we denote by  $f_s$ . Similarly, the local maximum of the CRC occurs at resonant frequency of the cell, denoted by  $f_c$ . We assume, without loss of generality, that  $f_c < f_s$ . We also assume that  $\kappa(f)$  and  $A(K, f)$  are monotone functions on the interval  $I = [f_c, f_s]$  and that they are each differentiable functions of  $f$  and  $K$ ; see Fig 2.

A generic property of most neurons is that the amplitude of subthreshold oscillations increases with stimulus strength. synaptic cell. Making this assumption, here, yields

$$\frac{\partial A}{\partial K} > 0. \tag{5}$$

As a function of frequency, however, the CRC  $A(K, f)$  is monotone decreasing on  $I$  (Fig. 2a) since it attains its local maximum at  $f = f_c$ . Therefore for  $f \in [f_c, f_s]$ ,

$$\frac{\partial A}{\partial f} \leq 0. \tag{6}$$

Similarly, the assumptions on the synaptic profile imply that on the interval  $I$ , the SRC obeys

$$\kappa'(f) \geq 0. \tag{7}$$

Note that equality holds in (6) when  $f = f_c$  and in (7) when  $f = f_s$ . Next differentiate to obtain

$$\frac{d}{df}A(\kappa(f), f) = \kappa'(f)\frac{\partial A}{\partial K} + \frac{\partial A}{\partial f} \tag{8}$$

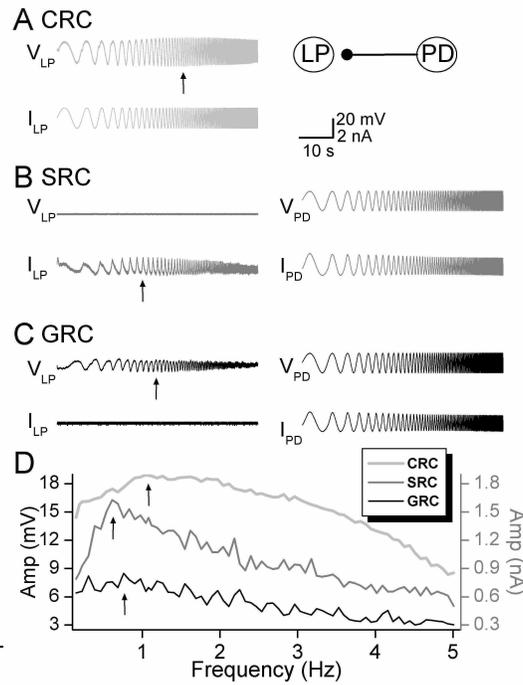


Figure 1: Measurement of CRC, SRC and GRC in the crab pyloric network. A. The CRC of the LP neuron was measured by injection a ZAP current of amplitude  $\pm 2$  nA sweeping frequencies between 0.1 and 5 Hz and indicated a resonance frequency at 1.5 Hz. Inset shows schematic diagram indicating the synaptic connection from the PD neuron to the LP neuron. B. The SRC was measured by voltage-clamping the PD neuron with a ZAP function waveform and measuring the synaptic current in the LP neuron. The LP neuron was voltage clamped at a holding potential of -60 mV. The SRC showed a peak value at 0.68 Hz. C. The GRC was measured by injecting a ZAP function waveform into the voltage-clamped PD neuron while the postsynaptic response of the LP neuron was measured in current clamp. The GRC showed a peak value at 0.87 Hz. D. The CRC, SRC and GRC plotted as a function of frequency indicate the distinct peaks of the CRC and the SRC and that the GRC peak falls between these two values.

We can show that the location of the local maximum of the GRC lies between  $f_c$  and  $f_s$  by showing that the derivative  $dA/df = 0$  for some  $f \in (f_c, f_s)$ . At  $f = f_c$ , from (6),  $\frac{\partial A}{\partial f} = 0$ . Equations (5) and (7) then imply that the right-hand side of (8) is positive when  $f = f_c$ . Thus near  $f = f_c$  as frequency increases, the amplitude of subthreshold oscillations increases. Similarly, at  $f = f_s$ ,  $\kappa'(f_s) = 0$  from (7) and from (6),  $\partial A/\partial f < 0$ . Thus the right-hand side is negative at  $f = f_s$  and  $A(\kappa(f), f)$  is decreasing near this value. Since  $A(\kappa(f), f)$  is increasing at  $f = f_c$  ( $dA/df < 0$ ) and decreasing at  $f = f_s$  ( $dA/df > 0$ ), there must be an intermediate frequency at which  $dA/df = 0$ , corresponding to a local maximum of  $A(\kappa(f), f)$ , the GRC.

This theoretical argument shows that the presynaptic frequency corresponding to the maximum response occurs for neither the resonant frequency of the synapse nor the resonant frequency for the cell, but between them. If we assume a fixed CRC, then the location, within  $I$ , of the GRC maximum depends on the steepness of the synaptic profile and the location of the peak ( $f_s$ ). For very steep synaptic profiles ( $\kappa'(f)$  large on  $I$ ), the preferred frequency of the network will be very close to that of the synapse. For flat synaptic profiles ( $\kappa'(f) \approx 0$  on  $I$ ), the preferred network frequency will be near the that of the cell.

We now turn our attention to the synaptic models described earlier. To demonstrate the dependence of the GRC on the shape of the SRC curve, we consider Model 1 where the synapse evolves according to equation (1) and  $\kappa(f)$  is simply a Gaussian. In Figure 2 the GRC is shown along with the numerically computed CRC and the Gaussian  $\kappa(f)$  for  $\sigma = 10 \text{ sec}^{-1}$ . Most important to note is the location of the maximum of the GRC which clearly lies between the local maxima of the CRC and SRC. In Fig. 2, we demonstrate how the steepness of the synaptic profile influences the preferred network frequency. In this case, we compare the SRC with  $\sigma = 40 \text{ ms}^{-2}$  to a SRC with  $\sigma = 10 \text{ ms}^{-2}$  in equation (2). The smaller value of  $\sigma$  results in a much flatter SRC. The resonant frequency of the GRC is shifted toward the peak of the CRC, as predicted by the theory.

We now demonstrate that the prediction of the location of the GRC peak is also valid for more realistic synaptic inputs using Model 2. In Model 2, the synapse is governed by equations (3) and (4) and the SRC has a frequency dependent peak due to the combination of short-term facilitation and short-term depression. that a biologically realistic synapse reproduces the theoretical predictions. In Fig. 3 the SRC, CRC and GRC are shown, with parameter values given in the caption.

As predicted, the peak of the GRC lies slightly to the right of the peak of the CRC and to the left of the peak of the SRC. Because the SRC is rather flat, resonant network frequency is much closer to that of the cell than the synapse.

## 4 Discussion

Many neurons and synapses respond optimally at a preferred resonant frequency. How resonance properties of neurons and synapses interact in a network setting remains an interesting and challenging question. We have shown that in the crab pyloric network, cellular and synaptic resonances interact to produce a maximum-amplitude subthreshold response at an intermediate frequency. This led us to develop a mathematical framework to examine whether this interaction between cellular (CRC) and synaptic (SRC) response curves holds

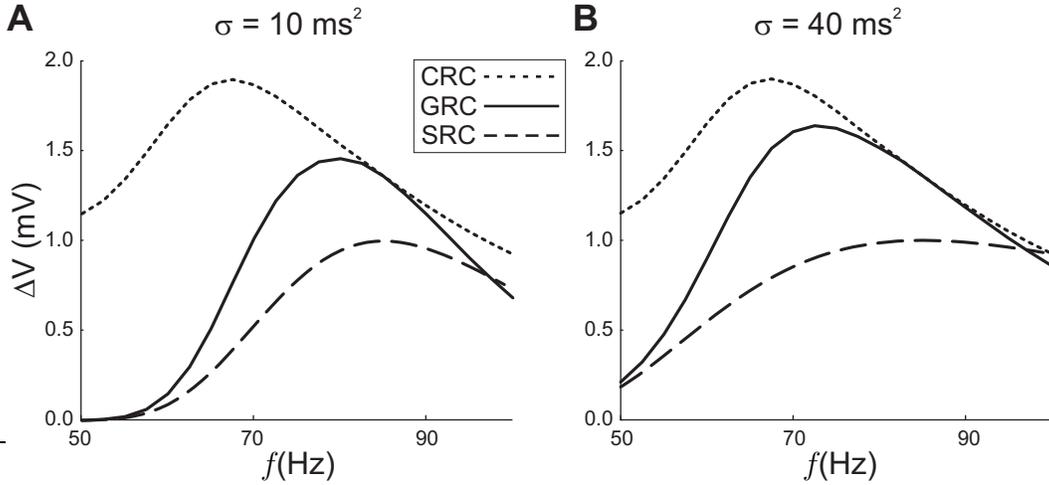


Figure 2: The response curves corresponding to equations (1, 2). The dotted curves (CRCs) are the amplitude of oscillations for the given frequency (horizontal axis). The solid curves (GRCs) are the oscillation amplitude of the postsynaptic response when the signal is passed via a resonant synapse. The SRCs (dashed) are unitless curves superimposed so that the preferred resonant frequency can be seen. From this figure it is easy to see that the maximum network response occurs for a frequency that is neither the resonant frequency for the cell nor the preferred frequency for the synapse. In the left panel  $\sigma = 10 \text{ ms}^2$ , resulting in a steeper profile for the SRC than in the right panel. The peak of the GRC is located closer to the peak of the SRC than in the right panel ( $\sigma = 40 \text{ ms}^2$ ).

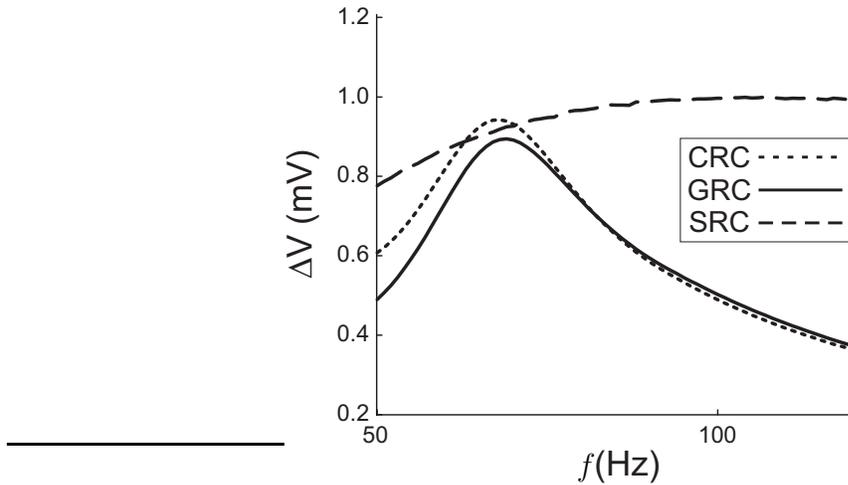


Figure 3: The SRC, CRC, and GRC when the synapse described by (3) (4) is used. The Hodgkin-Huxley parameters are the same as in figure 2. Parameter values for the synapse in this figure are  $\tau_{rec} = 125$ ,  $\tau_1 = 3$ ,  $\tau_{facil} = 100$  and  $U = 0.01$ . The peak of the SRC is near  $130 \text{ Hz}$ .

in general. Our results show that the resonant frequency of a feed-forward network indeed lies between the resonant frequencies of the synapse and the postsynaptic neuron. We also demonstrate that by changing the shape of the CRC or SRC, while keeping the location of the resonant frequencies associated with these curves fixed, the shape of the combined response (GRC) can be changed and its resonant frequency can be shifted. Although we analyzed the GRC shape with the assumption that  $f_c < f_s$ , it should be noted that the opposite assumption produces similar results. In general, these results suggest that resonance properties can interact in non-intuitive ways.

A future step to be undertaken is to use mathematical analysis to understand how sub-threshold resonance relates to the spiking behavior of neurons [7]. We are currently studying this problem in the context of feedback networks modeled on the PD-LP neuron loop of the crab.

## Acknowledgements

Supported by an NJIT Strategic Initiative Grant to the Department of Mathematical Sciences (JDD), NIH grant MH-60605 (FN) and NSF grant DMS-0315862(AB).

## References

- [1] A. L. Hodgkin and A. F. Huxley, A quantitative description of membrane current and its application to conduction and excitation in nerve, *J Physiol* 117 (1952) 500-44.
- [2] B. Hutcheon and Y. Yarom, Resonance, oscillation and the intrinsic frequency preferences of neurons, *Trends Neurosci* 23 (2000) 216-22.
- [3] E. M. Izhikevich, N. S. Desai, E. C. Walcott and F. C. Hoppensteadt, Bursts as a unit of neural information: selective communication via resonance, *Trends Neurosci* 26 (2003) 161-7.
- [4] I. Lampl and Y. Yarom, Subthreshold oscillations of the membrane potential: a functional synchronizing and timing device, *J Neurophysiol* 70 (1993) 2181-6.
- [5] Y. Manor and F. Nadim, Synaptic depression mediates bistability in neuronal networks with recurrent inhibitory connectivity, *J Neurosci* 21 (2001) 9460-70.
- [6] E. Puil, H. Meiri and Y. Yarom, Resonant behavior and frequency preferences of thalamic neurons, *J Neurophysiol* 71 (1994) 575-82.
- [7] M. J. Richardson, N. Brunel and V. Hakim, From subthreshold to firing-rate resonance, *J Neurophysiol* 89 (2003) 2538-54.
- [8] M. Tsodyks, A. Uziel and H. Markram, Synchrony generation in recurrent networks with frequency-dependent synapses, *J. Neurosci.* 20 (2000) 1-5.