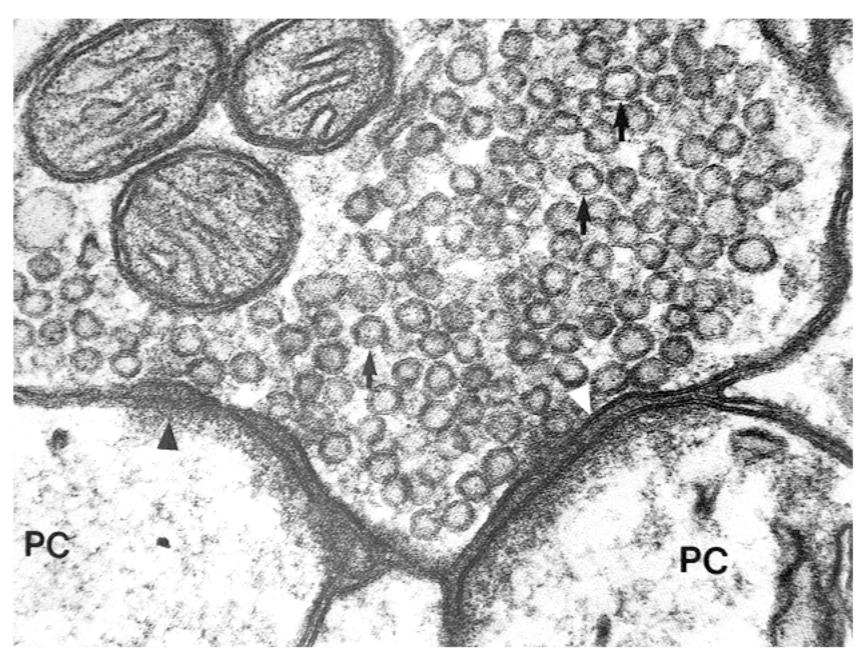
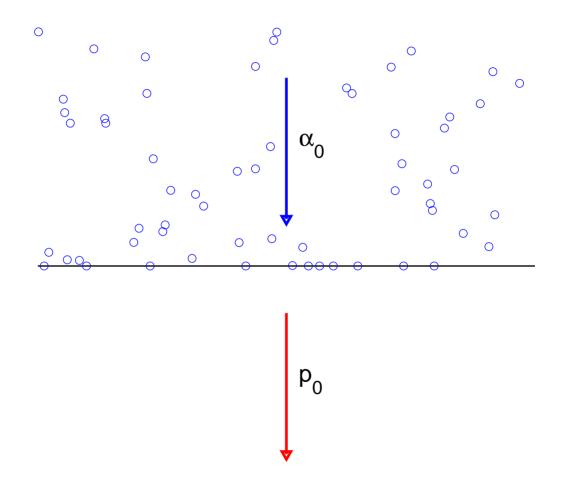
### Is There a Benefit of Randomness in Synaptic Vesicle Release?

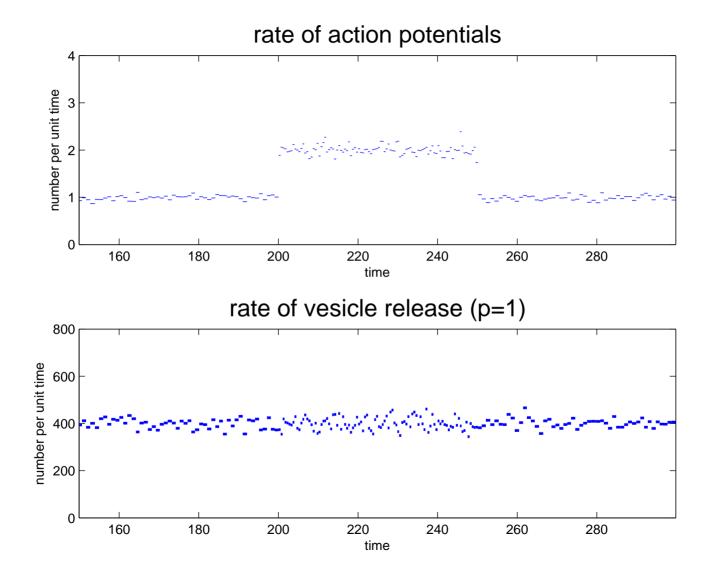
Calvin Zhang-Molina (1) and Charles S. Peskin (2)

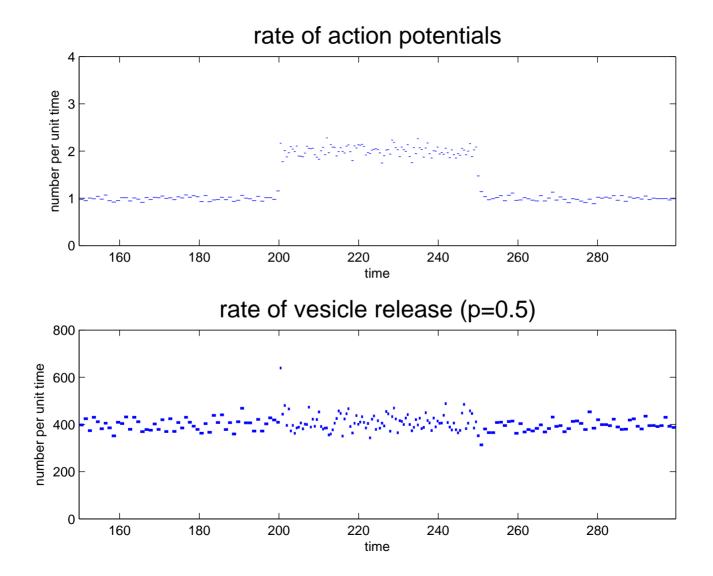
- 1. Department of Mathematics University of Arizona Tucson, AZ 85721 USA calvinz@math.arizona.edu
- Courant Institute of Mathematical Sciences New York University New York, NY 10012 USA peskin@cims.nyu.edu

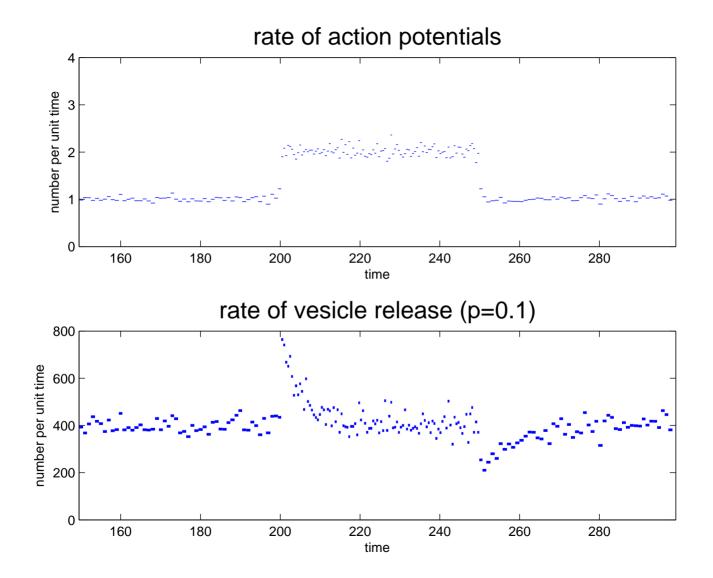


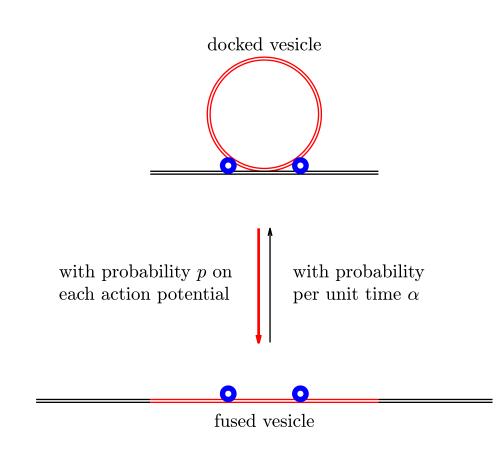
## Idealized Model with an Unlimited Number of Docking Sites



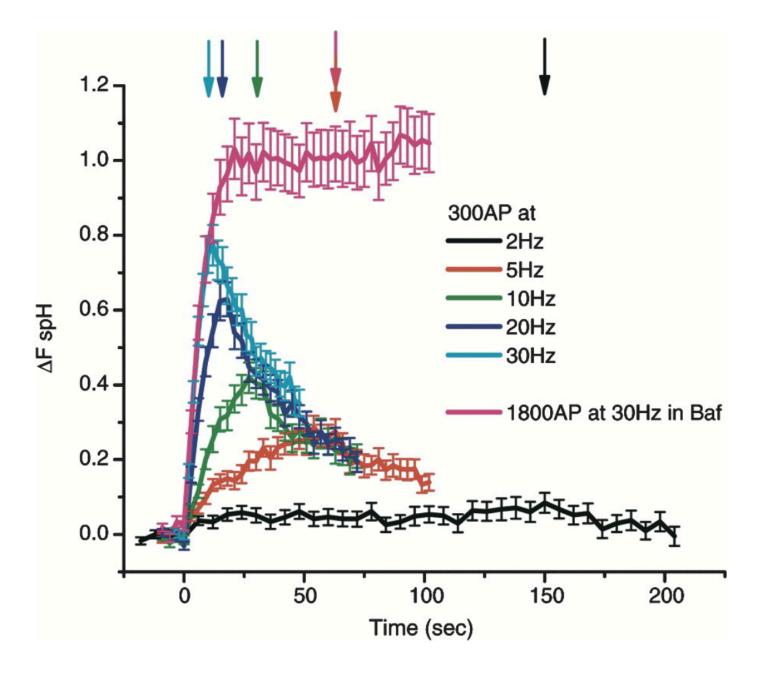








Fernandez-Alfonso T and Ryan TA: The Kinetics of Synaptic Vesicle Pool Depletion at CNS Synaptic Terminals. Neuron 41: 943-953, 2004



#### Modeling the Experiment of Fernandez-Alfonso & Ryan

#### Summary of the experiment

- Start with a synapse that has been at rest for some time
- Apply a sequence of 300 equally spaced action potentials
- Observe the fraction of fused vesicles as a function of time during and after the spike train
- Repeat for a variety of interspike intervals

#### Model of the experiment

- All docking sites are occupied at t = 0.
- Action potential arrival times are  $t_k = kT$  for k = 0, ..., n-1 with n = 300 and T = interspike interval of the experiment.
- Let *p* be the probability of fusion of any docked vesicle upon arrival of an action potential. We assume that *p* is constant during any one experiment, but that it depends (because of *facilitation*) on the rate of arrival of action potentials, and therefore varies from one experiment to another.
- Let α be the rate constant for reformation of a docked vesicle from a fused vesicle. We find that α is constant across all of the experiments.

Let f(t) be the expected fraction of fused vesicles at time t. Then

$$f(t_0^-) = 0 (1)$$

$$f(t_k^+) = f(t_k^-) + p\left(1 - f(t_k^-)\right), \quad k = 0, \dots, n-1$$
(2)

Between the arrival times of action potentials, and also after the arrival of the last action potential, we have the following differential equation for f(t):

$$\frac{df}{dt} = -\alpha f \tag{3}$$

From the decay of f(t) following the end of the train of action potentials, we can check whether the decay is indeed exponential and identify the parameter  $\alpha$ . We get a good fit to all of the individual experiments with

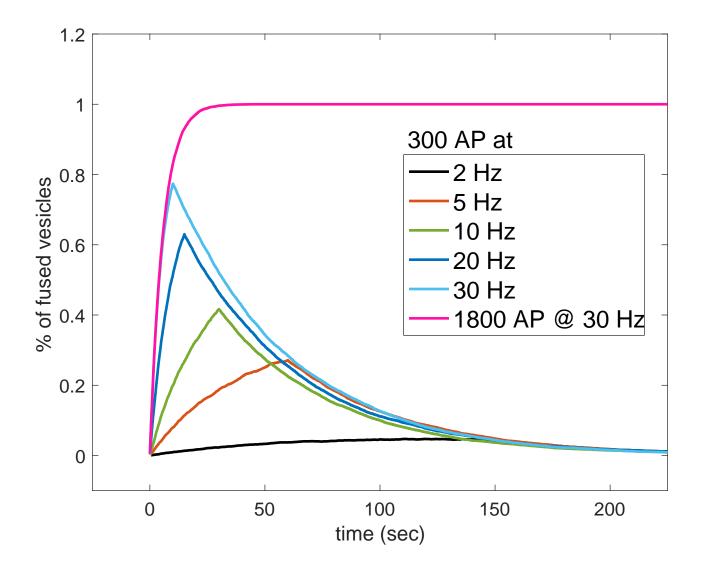
$$\alpha = 0.02/s \tag{4}$$

To evaluate p, which we expect to be different in each experiment, we use the fraction of fused vesicles immediately following the last action potential in the spike train. This is given by

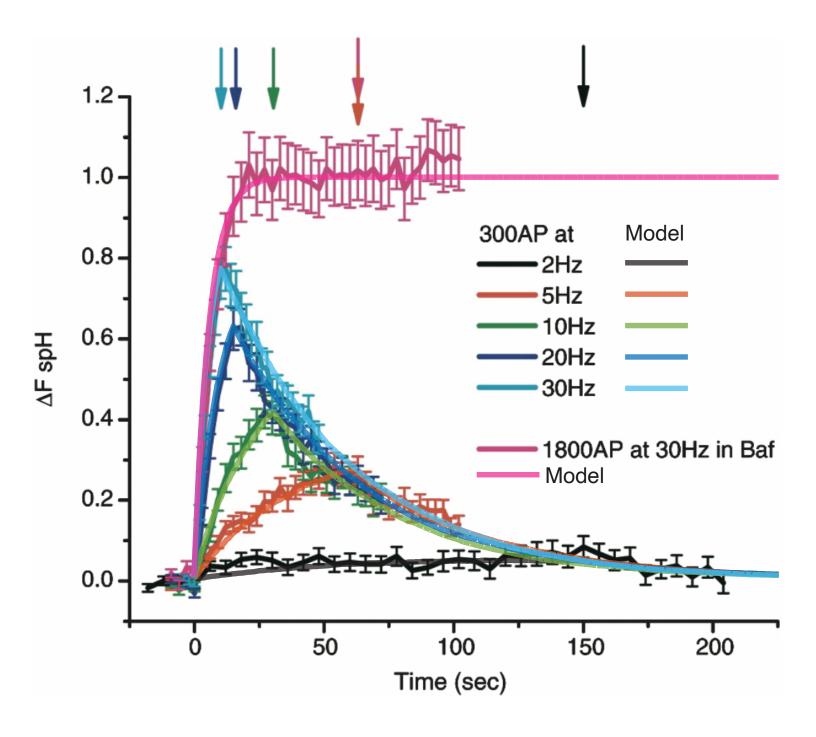
$$f(t_{n-1}^{+}) = p \frac{1 - ((1-p)\exp(-\alpha T))^n}{1 - (1-p)\exp(-\alpha T)}$$
(5)

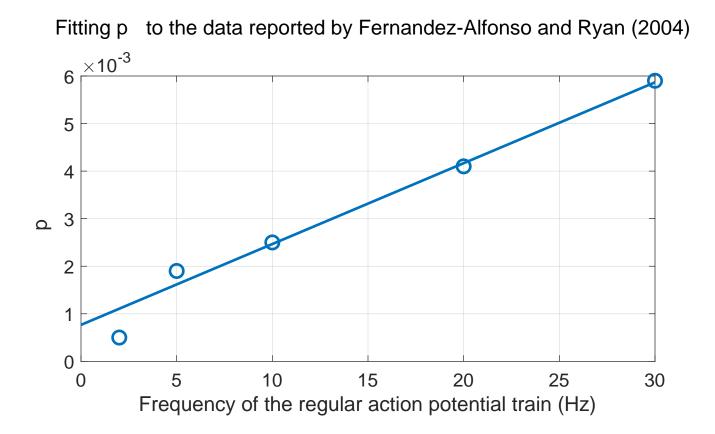
The left-hand side is a number in (0,1) that is experimentally determined, and everything on the right-hand side is known except for p. It is straightforward to show that there is a unique  $p \in (0,1)$  that satisfies (5) and we find that value numerically. With  $\alpha$  and p determined, we can evaluate f(t) for any t by solving (1-3) and compare to the experimental results. The comparison is shown in the following figures.

# **Model Results**



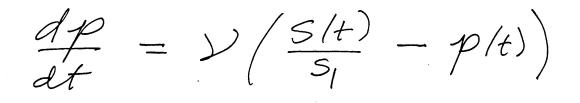
Model Results & Experimental Results Superimposed





Interaction of facilitation/depression

 $\frac{dn}{dt} = \chi(n_s - n/t) - r/t)$ 



r(t) = 5(t) p(t) n(t)

S(t) = rate farrival of action potentials p/t) = probability of release for each do cred ve sicle when an actim potenticl arrives n(t) = number of docked vesicles r/t) = rate of vesicle release

Parameters

ns = # of docking sites

X = rate constant for refill of empty site (0.02/second)

2 = rate constant for adaptation of facilitation

S1 = Extrapolated rate of arrival of action potatials

that would make p=1

(6000/second)

Sensitivity output to imput:  $\Delta r/r$ As/s

Sensitivity to sudden changes is equal to 1, since n and p cannot change abruptly Sensitivity to slow changes is egnal to

 $2\left(\frac{n^*}{n_s}\right) = 2\left(\frac{1}{1+\frac{(S^*)^2}{\sqrt{S_1}}}\right)$ 

where  $\sqrt{XS_1} \cong 10/\text{second}$ 

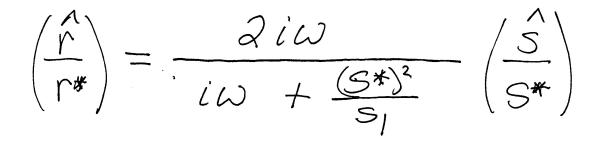
Frequency response of the synapse to small-amplitude modulation of a regular spike drain :  $\left(\frac{\hat{r}}{r^*}\right) = \frac{1 + \frac{1}{1 + i\omega/\nu}}{\frac{(5^*)^2}{1 + \frac{(5^*)^2}{\alpha s_1}}}{\frac{1 + \frac{(5^*)^2}{\alpha s_1}}{1 + i\omega/\alpha}}$ <u>S</u> \$\*

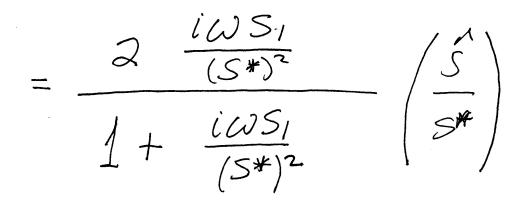
High-Brequency gain = 1

Low - frequency gam =  

$$2\left(\frac{1}{1+\frac{(S^*)^2}{\sqrt{S_1}}}\right)$$

If a << w << 2, then





and this is the transfer function Rapsendo-differentictor.

## References on Stochastic Aspects of Synaptic Vesicle Dynamics

Zhang C and Peskin CS: Improved signaling as a result of randomness in synaptic vesicle release. Proceedings of the National Academy of Sciences USA 112(48): 14954-14959, 2015

Zhang C and Peskin CS: Analysis, simulation, and optimization of stochastic vesicle dynamics in synaptic transmission. Communications on Pure and Applied Mathematics 73(1): 3-62, 2020