

Question of the Day

Fill in the blank:

“All models are _____, but some are _____.”

“If you have a hammer, then everything looks like _____”

Your alien friend sees that people use umbrellas in the rain, and infer that umbrellas are a rain-blessing artifact. He is committing the fallacy of _____

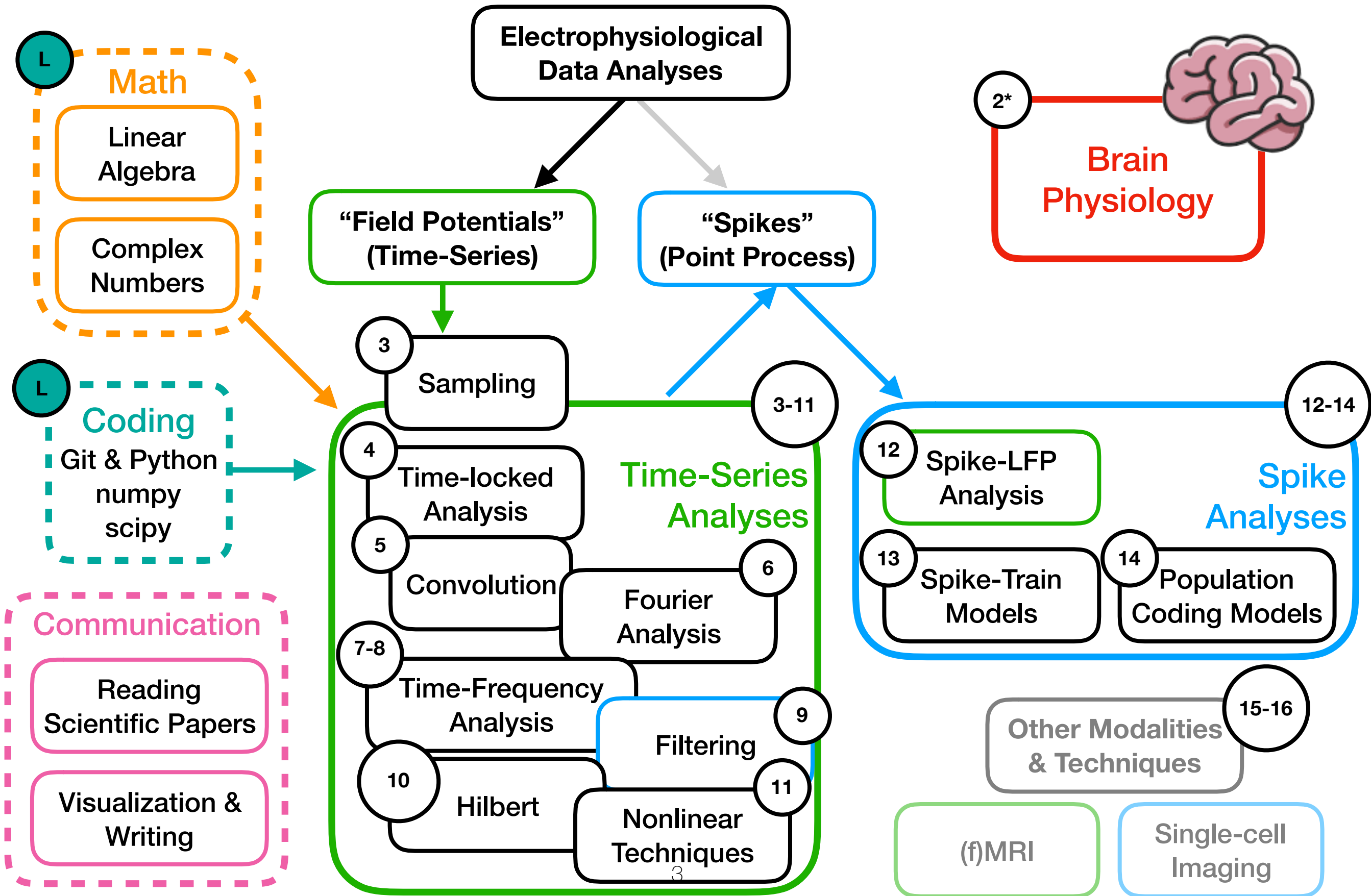


Spikes: Rate Models & Rate-LFP Analyses

Lecture 13
July 24, 2019



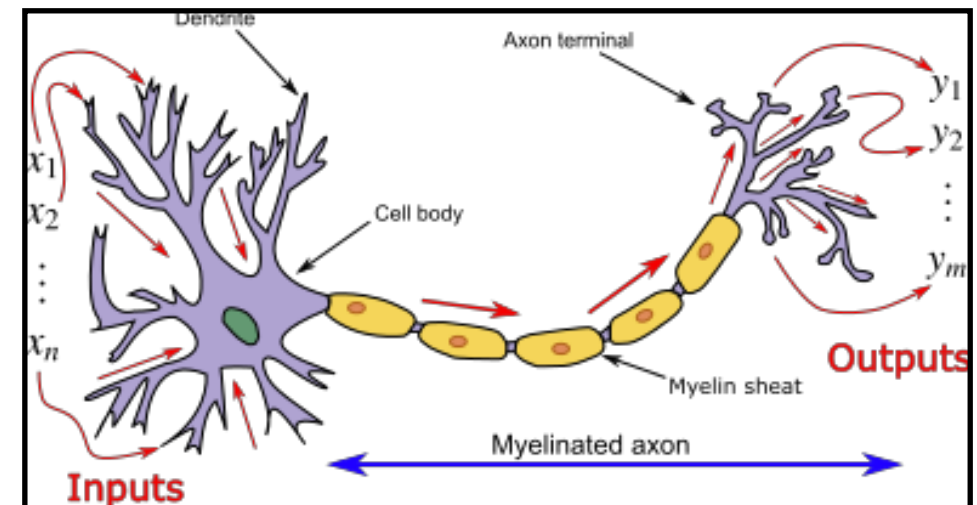
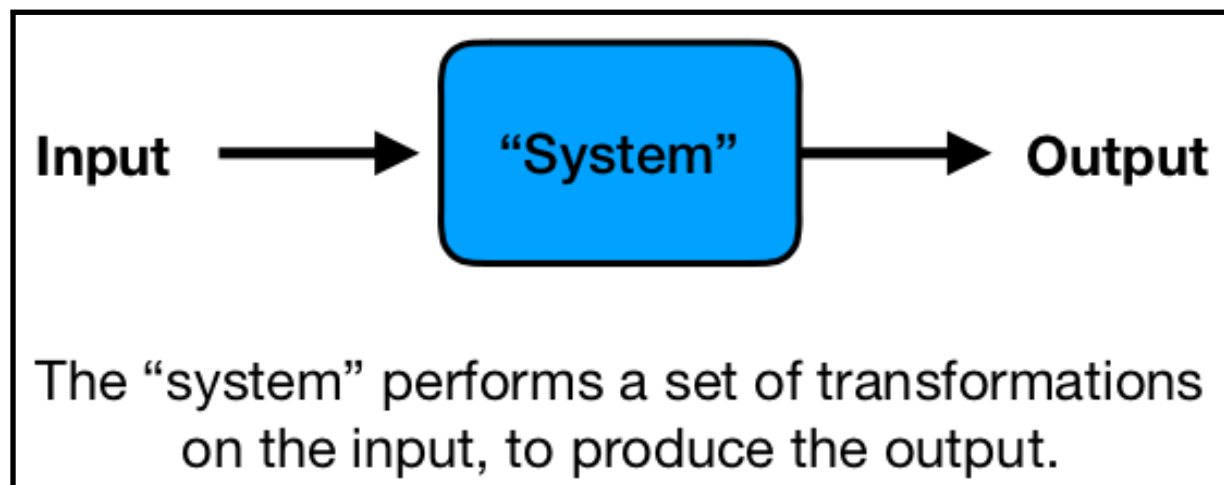
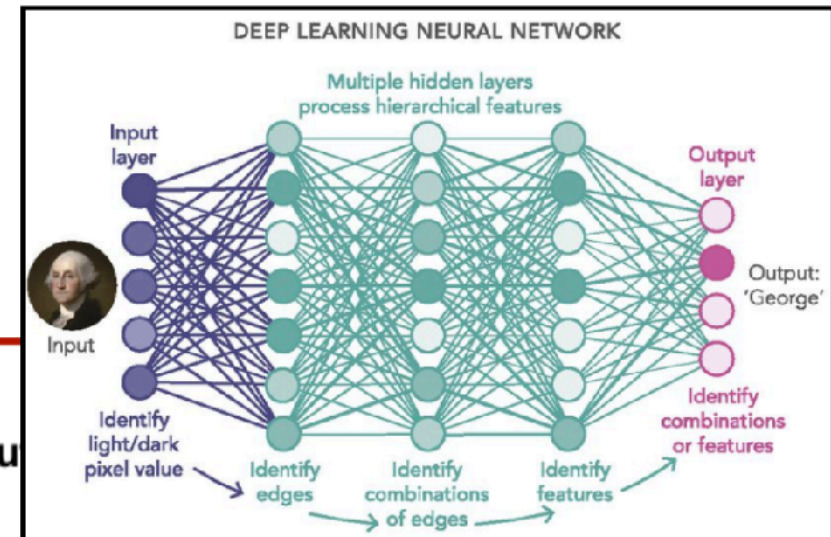
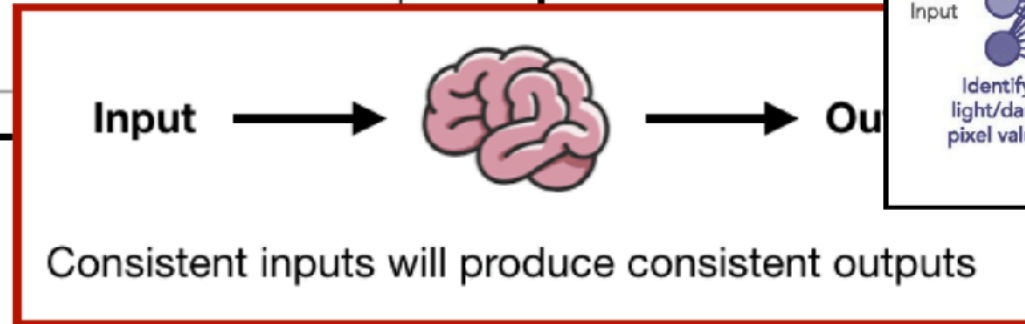
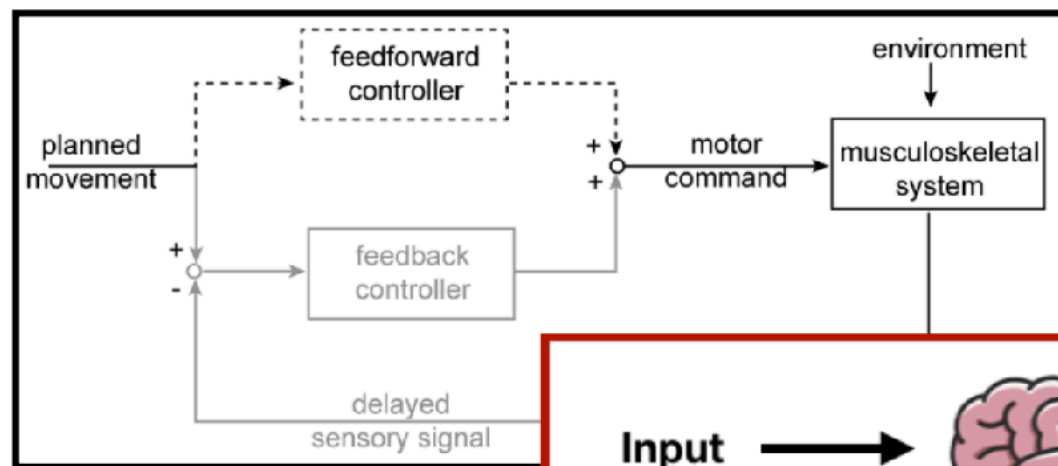
Course Outline: Road Map



1. Conceptualize neuron as computational device
2. Compute spike counts, smoothing & firing rate
3. Understand correlation and rate-LFP analyses



Single Neuron as a Computational Device

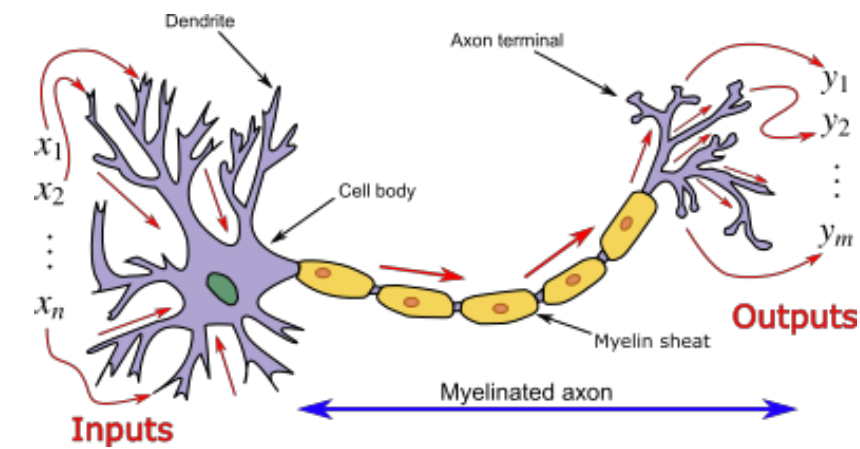


Systems engineering view: neuron receives (sensory or synaptic) input, performs a “computation”, and sends the result as an output.

This is a **model**, or abstraction, of the biological cell!



Single Neuron as a Computational Device



$$f\left(\sum_i w_i x_i + b\right)$$

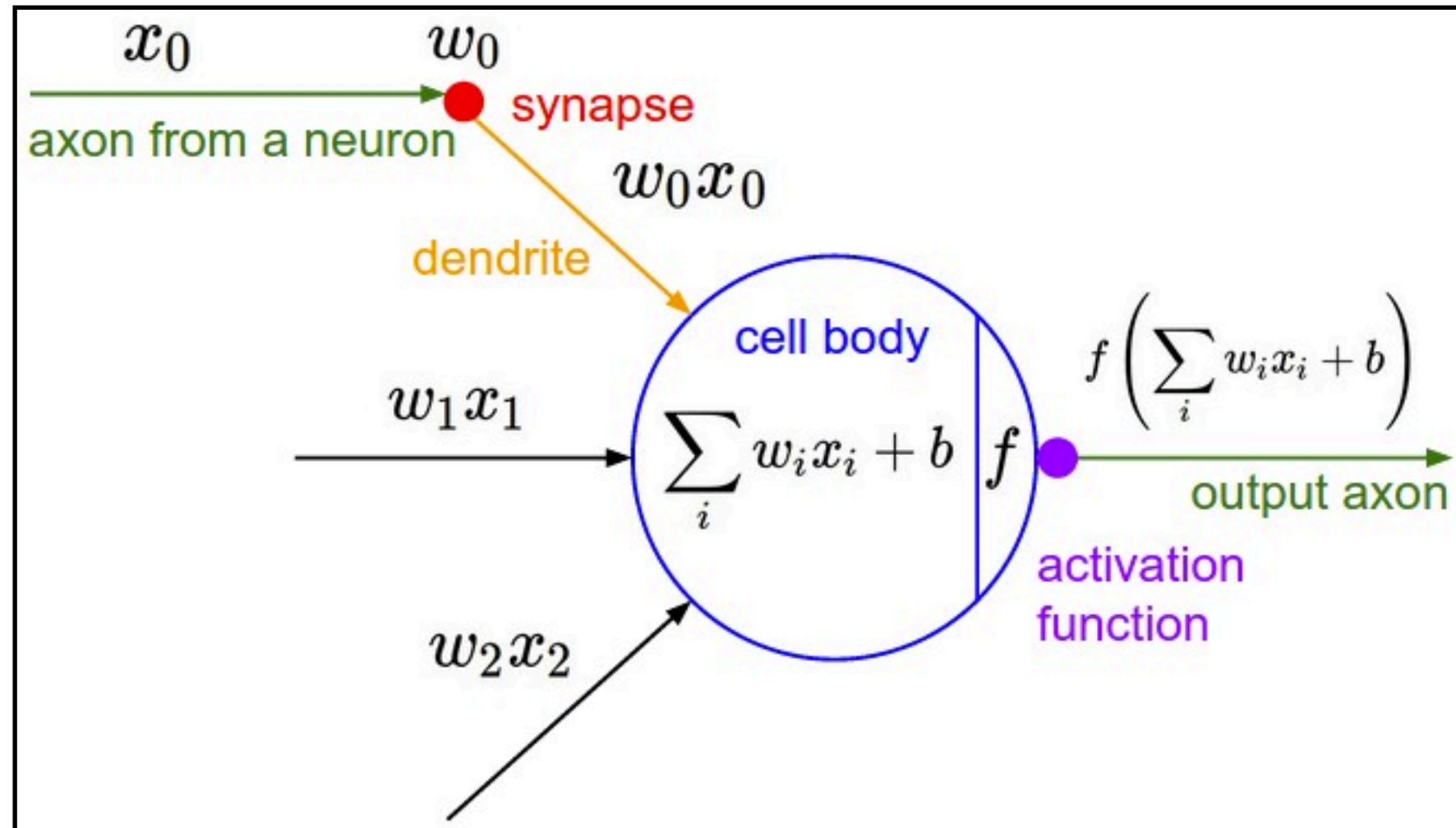
dot product

w: “synaptic” weights

x: inputs

b: bias

$f()$: nonlinearity



But what is x, physically?



Information Encoding

The biological neuron only receives and emits discrete **action potentials**.

What aspect of the action potential “encodes information”, that we can manipulate computationally?

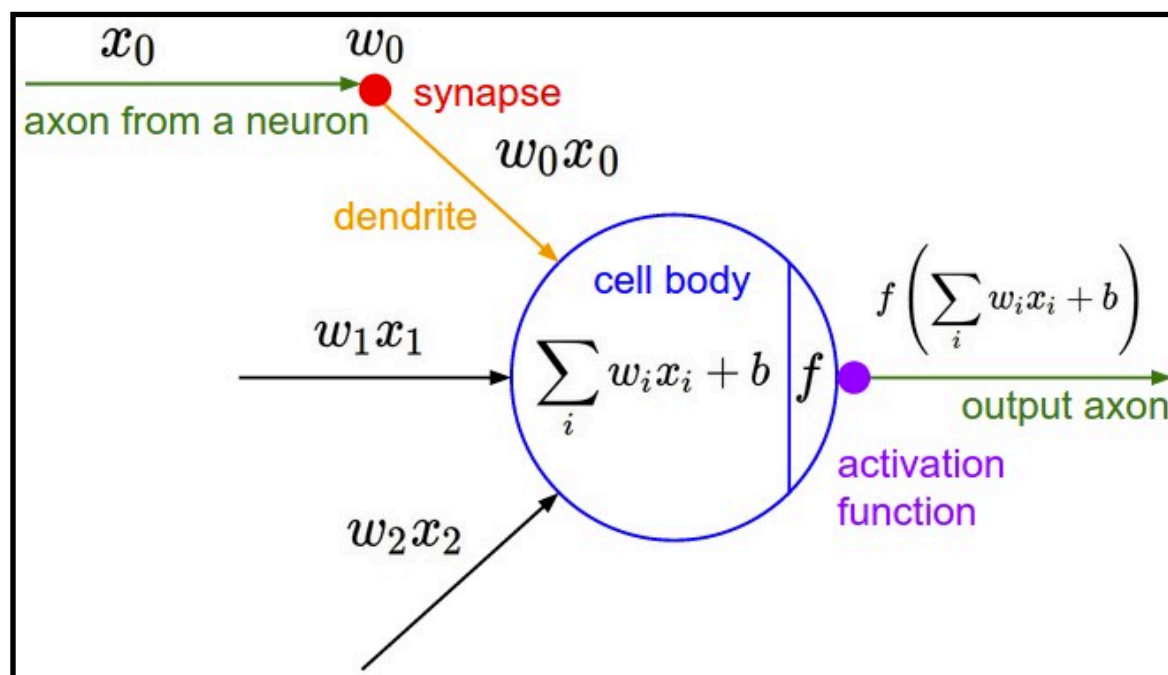
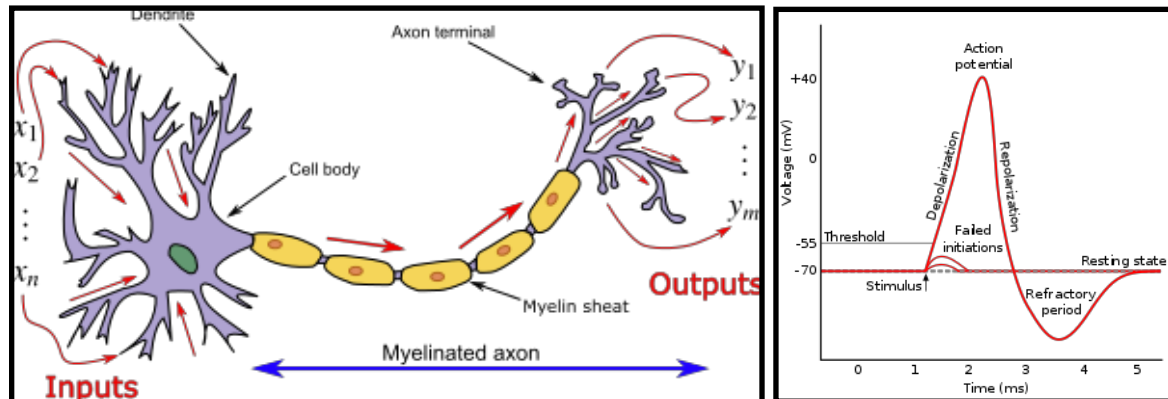
Action potentials (more or less) have constant amplitudes and widths...

One view (rate coding):

Number of spikes within a time window.

Spike Count: over a longer window

Firing Rate: “instantaneous” quantity

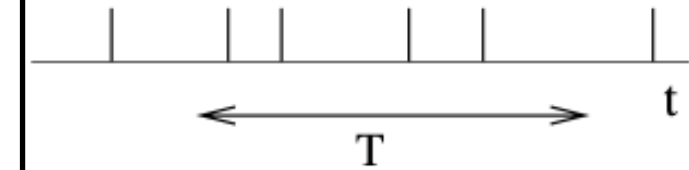


Spikes, Spike Counts, & Firing Rate

Rate = average over time
(single neuron, single run)

Spike count

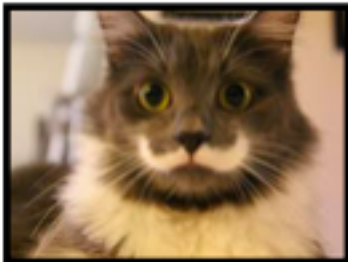
$$v = \frac{n_{sp}}{T}$$



Stimulus

Response

Trial 1



10 spikes -> 100Hz

Trial 2



2 spikes \rightarrow 20Hz

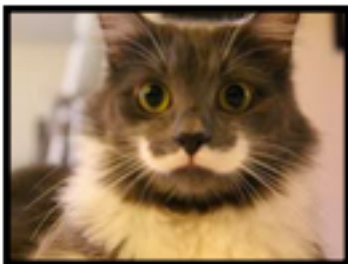
Trial 3




4 spikes \rightarrow 40Hz

■ ■ ■

Trial N



12 spikes -> 120Hz

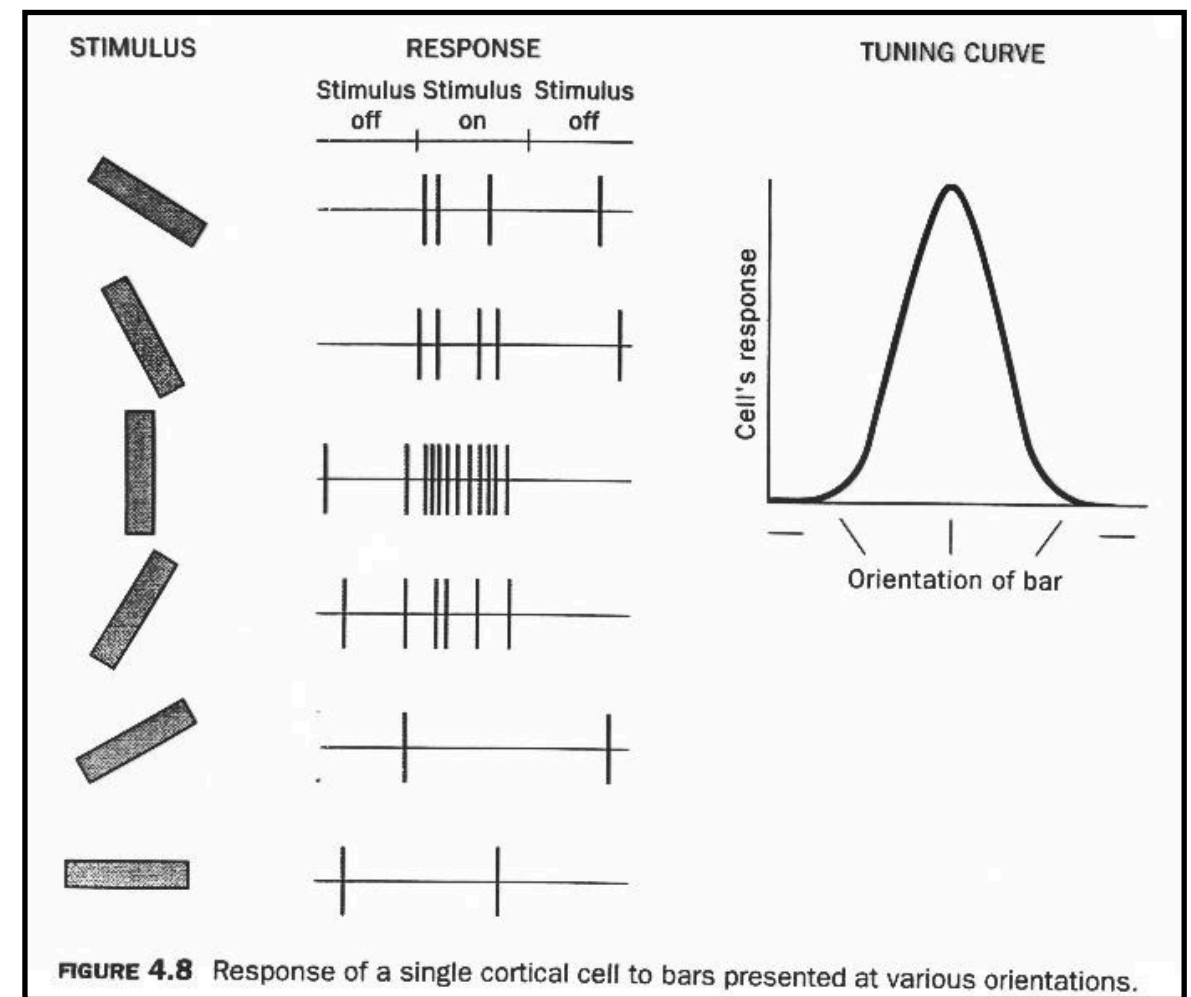
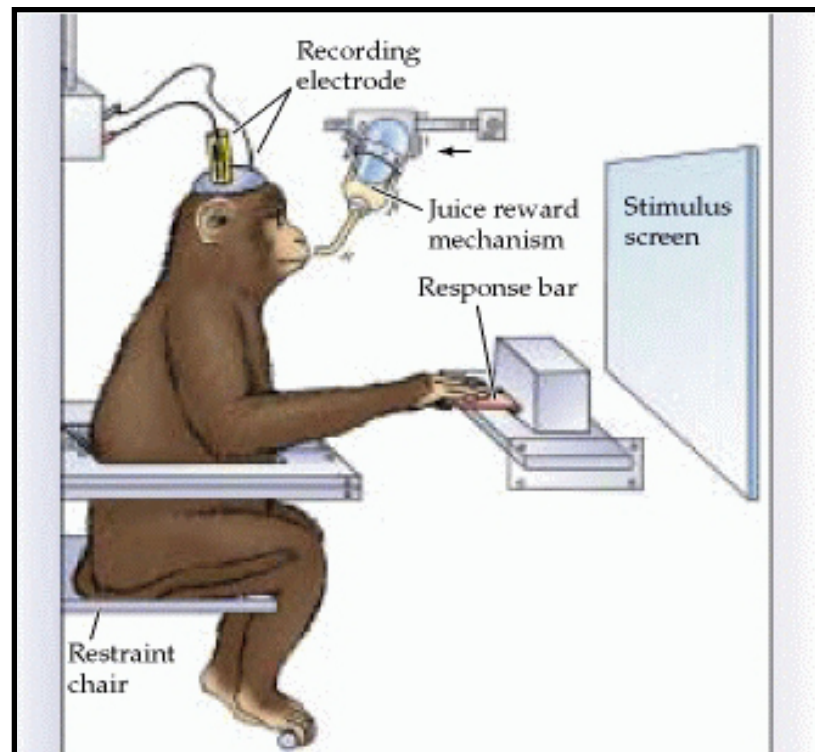


100 ms

This neuron is a
“cat” neuron.



Receptive Fields & Stimulus-Tuned Neurons



Stimulus with **varying aspects** are presented, e.g.:

- location
- orientation (angle)
- color
- sound frequency
- etc...

“This neuron has an orientation preference.”



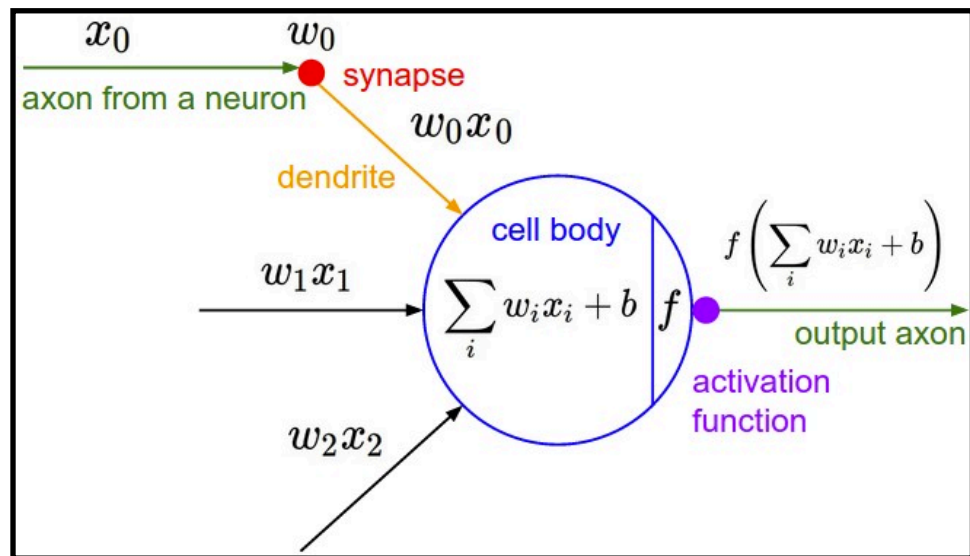
Receptive Fields & Stimulus-Tuned Neurons

Tuning	Location in the Brain	Special Name?
place	hippocampus	place cells & grid cells
motion	retina, or V5/MT	direction cells/motion cells
grandmother	IT/hippocampus	gnostic/grandmother/Jennifer aniston cells
numerosity	PFC	numérons
biological motions	premotor/SMA	mirror neuron
phallic images		

Google: “__ tuned neuron” or “neuron responsive to ____” or “__ cell neuroscience”



Single Neuron as a Computational Device



$$f\left(\sum_i w_i x_i + b\right)$$

dot product

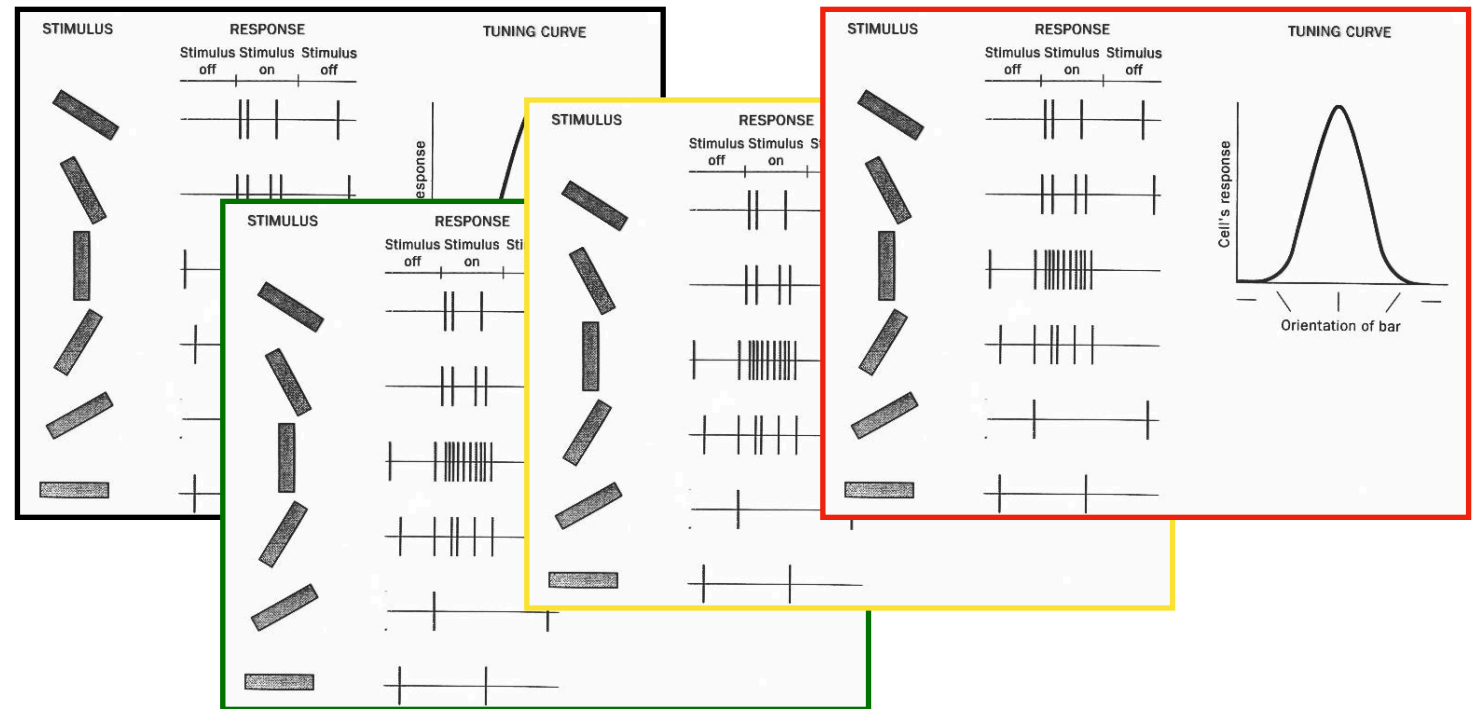
w: “synaptic” weights

x: inputs

b: bias

$f(\cdot)$: nonlinearity

Different neurons have different “tuning curves”, i.e., sensitive to different values of the variable.

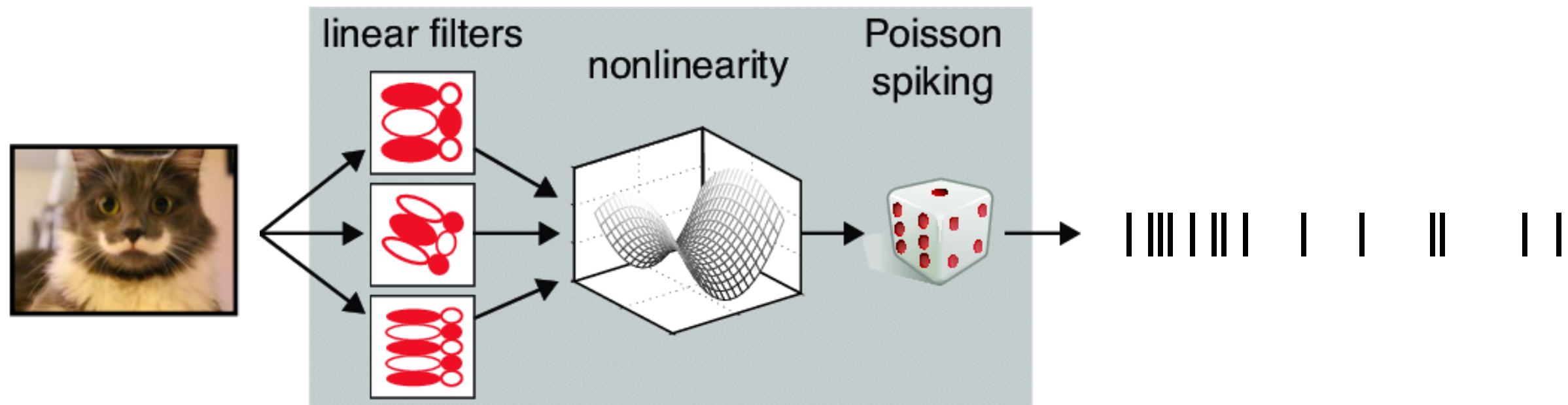


downstream neuron combines them linearly via dot product (linear filter)

Naive model of a neuron's function!

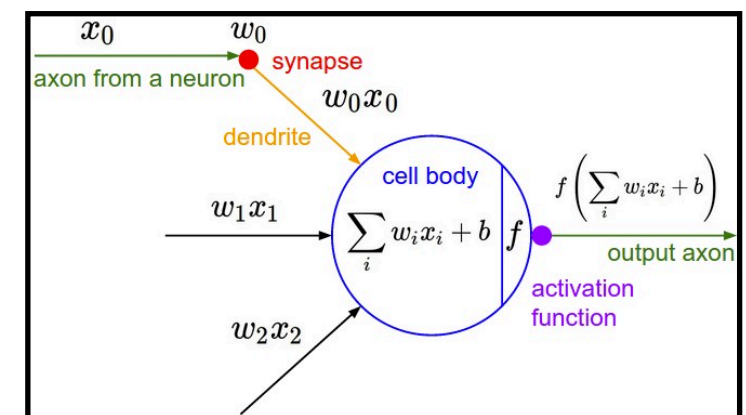


Linear-Nonlinear Poisson Model



linear: dot product / filter of stimulus tuning

nonlinear: activation function, e.g., sigmoid, ReLu



Poisson: spikes are randomly emitted as a Poisson process, given the **average rate parameter**

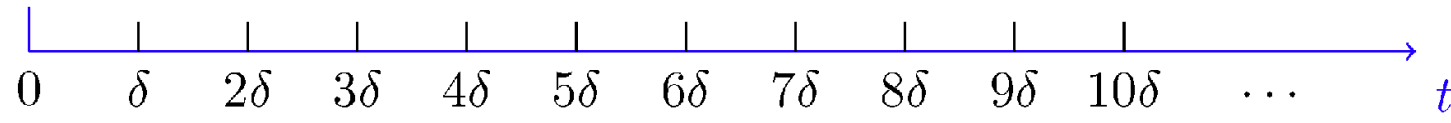
$$P\{N(t) = n\} = \frac{(\lambda t)^n}{n!} e^{-\lambda t}.$$



Brief Intro to Poisson Spikes

Poisson: spikes are randomly emitted as a Poisson process, given the **average rate parameter**

$$P\{N(t) = n\} = \frac{(\lambda t)^n}{n!} e^{-\lambda t}.$$



Generate a random number between 0-1 at every time step.

If that number is greater than **(rate * interval length)** -> no spike.

Otherwise -> spike

Exercise: rate = 5Hz, interval length = 0.1s, total time = 2 seconds

nice property:

for some duration T , average count = rate $\times T$ = variance

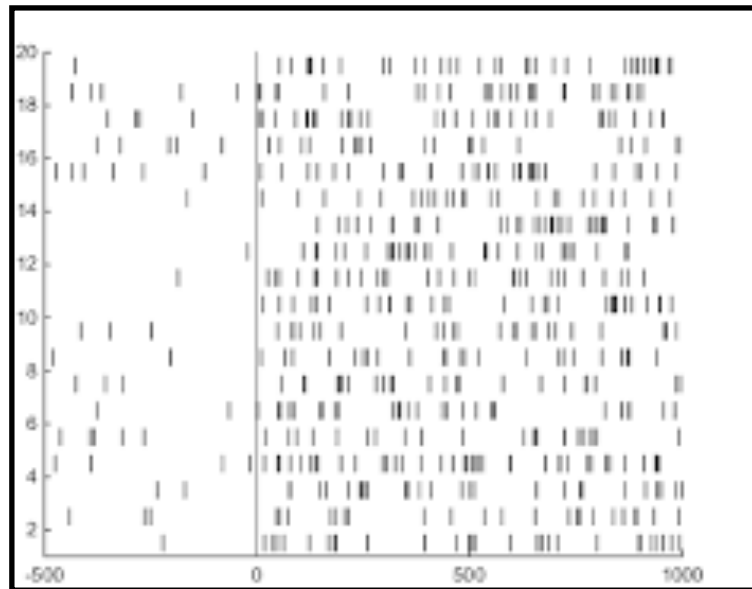


Rate Encoding Model

Model of how a neuron encodes information via spike trains.

Assumption: information (e.g., stimulus intensity) is encoded via a neuron's firing rate.

Assumes precise spike timing on the millisecond scale does not matter (in contrast to **spike timing encoding models**).



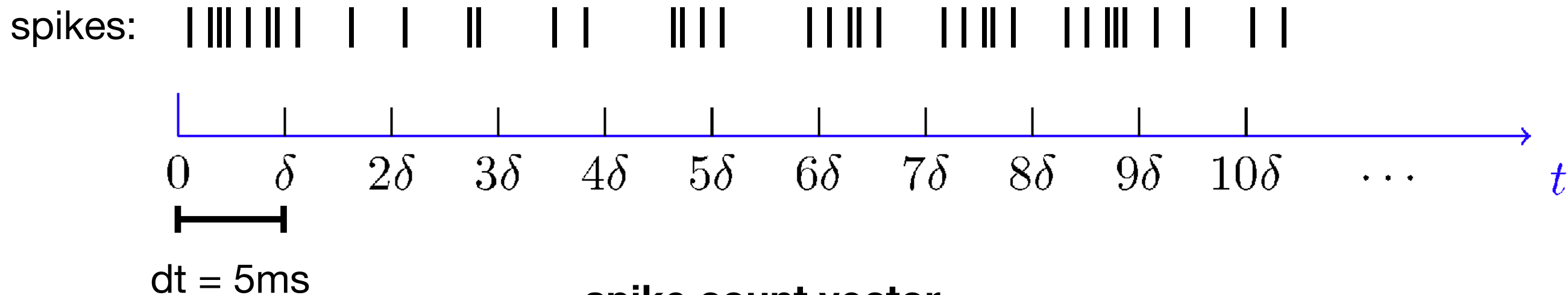
Given some spike timestamps, we want to estimate firing rate.



1. Conceptualize neuron as computational device
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Spike Times to Spike Counts



spike count vector

6	2	3	2	3	2	4	5	5	2	...
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or

8	5	5	9	7	...
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or

34

Time bin width is a parameter choice!

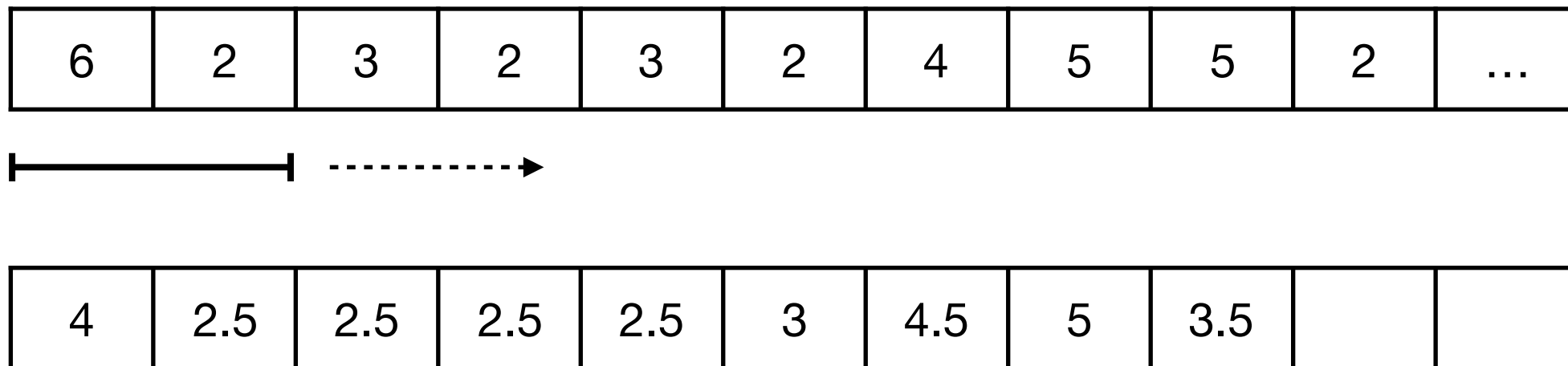
Smaller bins = better temporal resolution, but noisier estimate
Bigger bins = worse temporal resolution, more accurate estimate



Spike Counts to Firing Rate

Spike count vector is usually very noisy, we want a more continuous & smooth firing rate estimate.

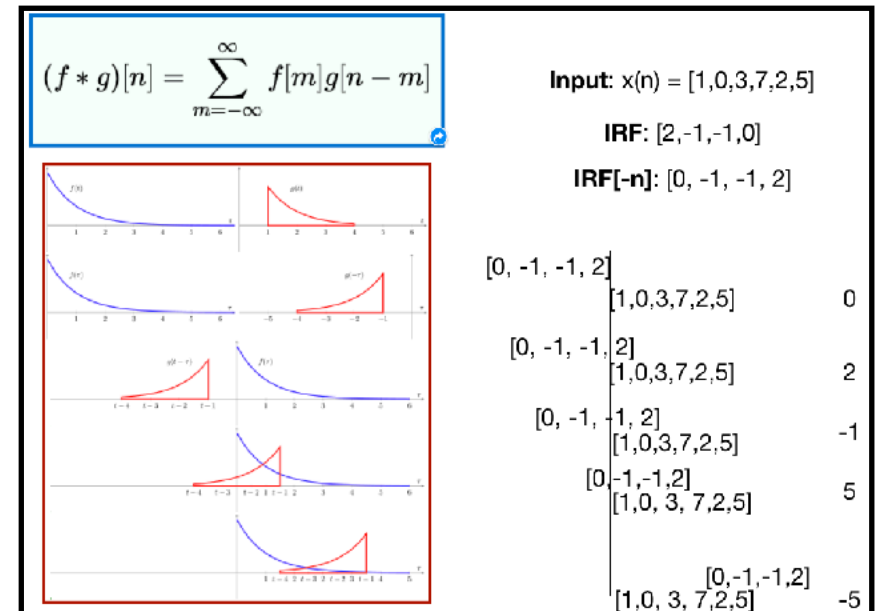
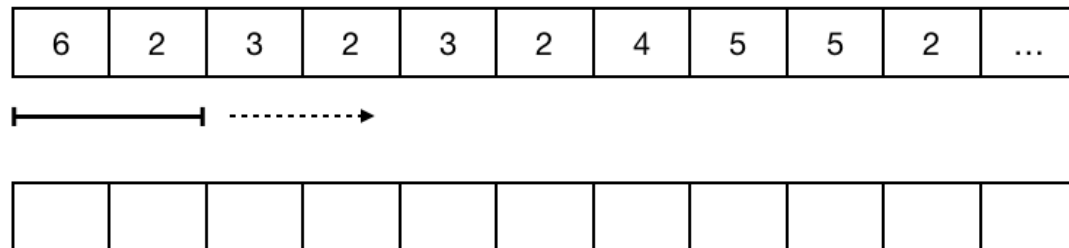
Solution: take “moving average”.



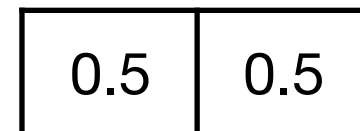
Moving Average Smoothing as Filter

Spike count vector is usually very noisy, we want a more continuous & smooth firing rate estimate.

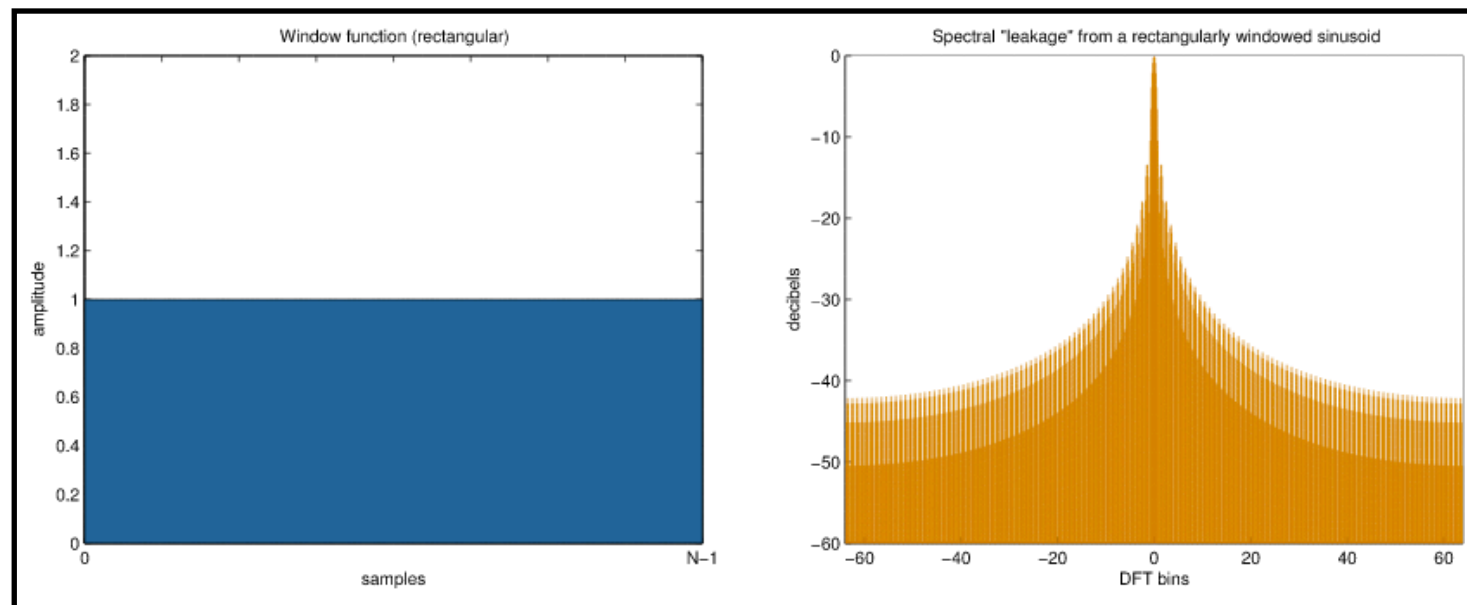
Solution: take “moving average”.



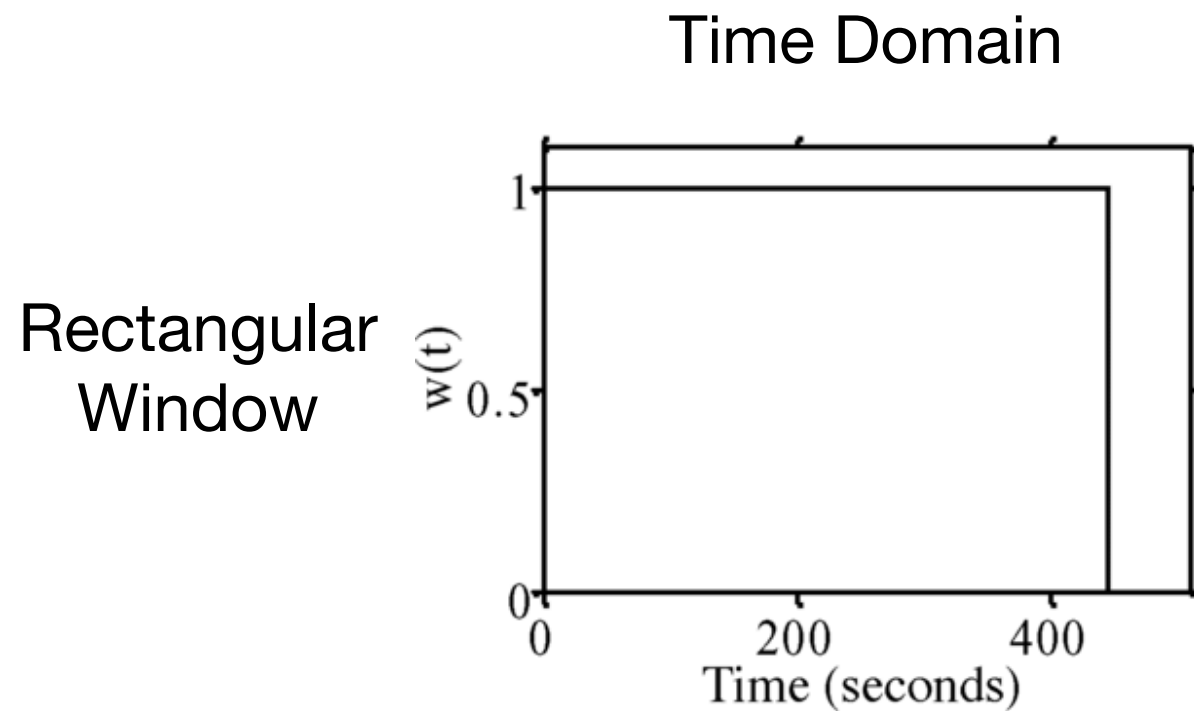
Equivalent to
convolution with:



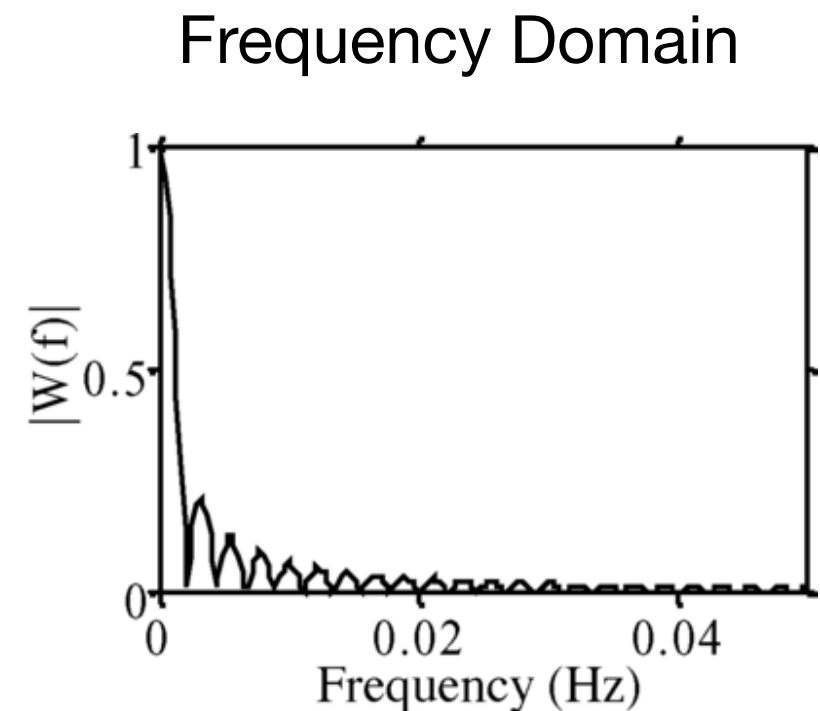
2-point boxcar or rectangular window



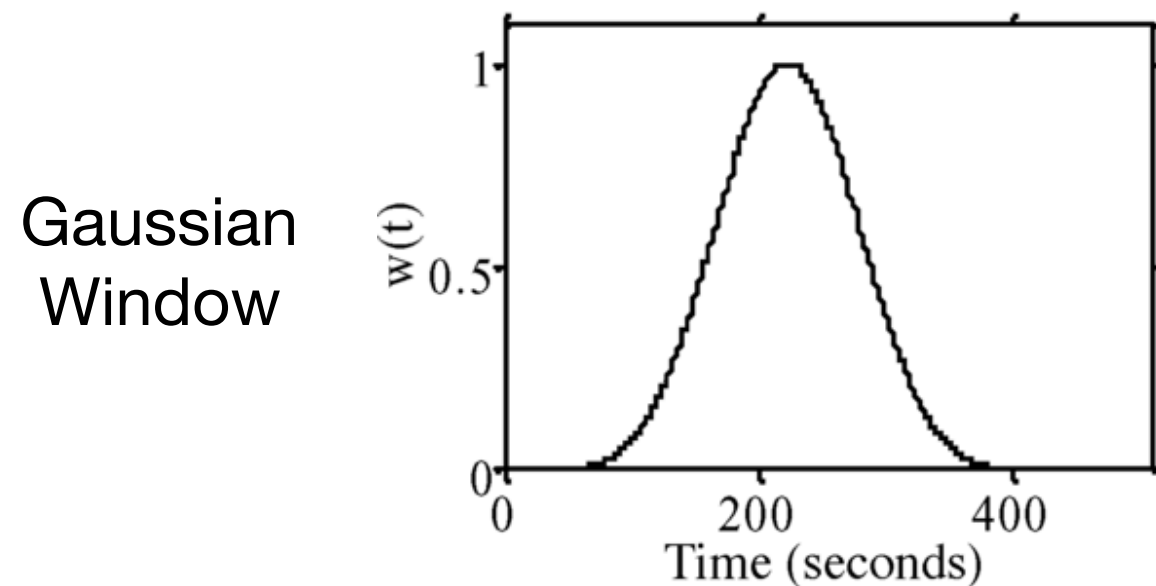
Firing Rate: Smoothed Spike Count



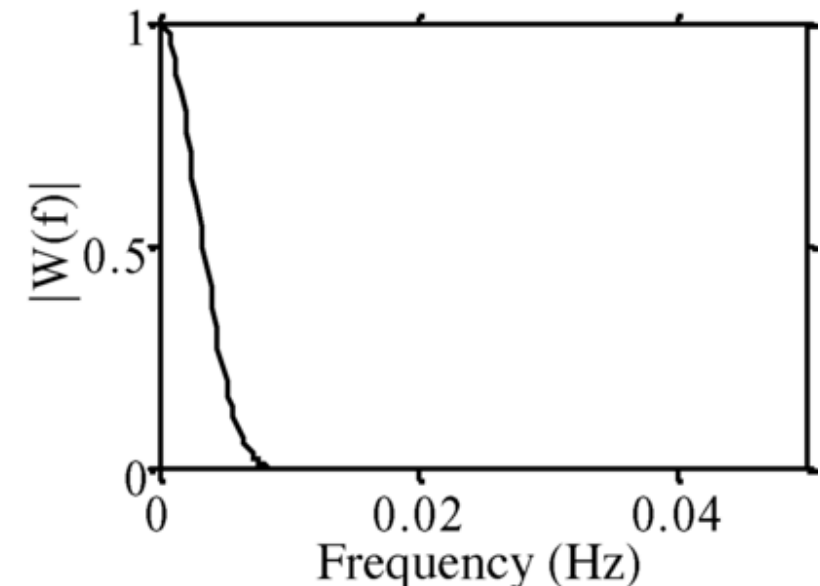
(a)



(b)



(c)



(d)

Smoothing is applying a low-pass filter!

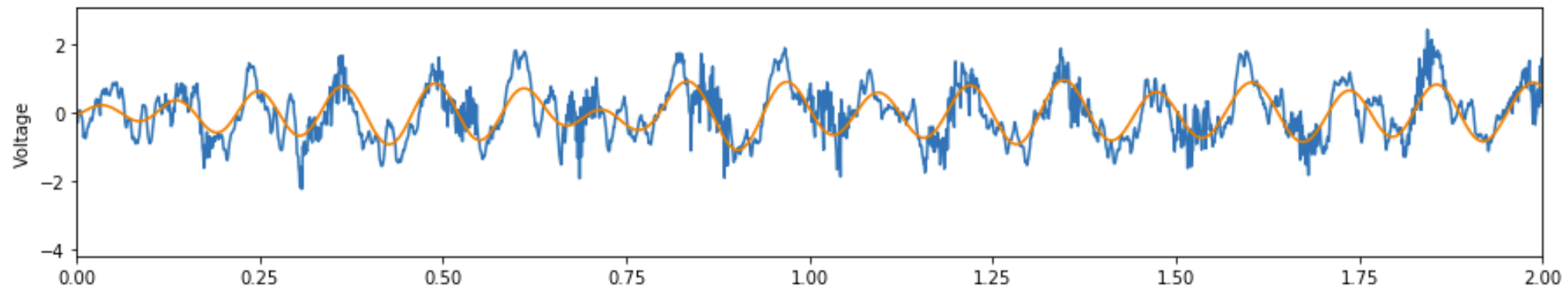


1. Conceptualize neuron as computational device
2. Compute spike counts, smoothing & firing rate
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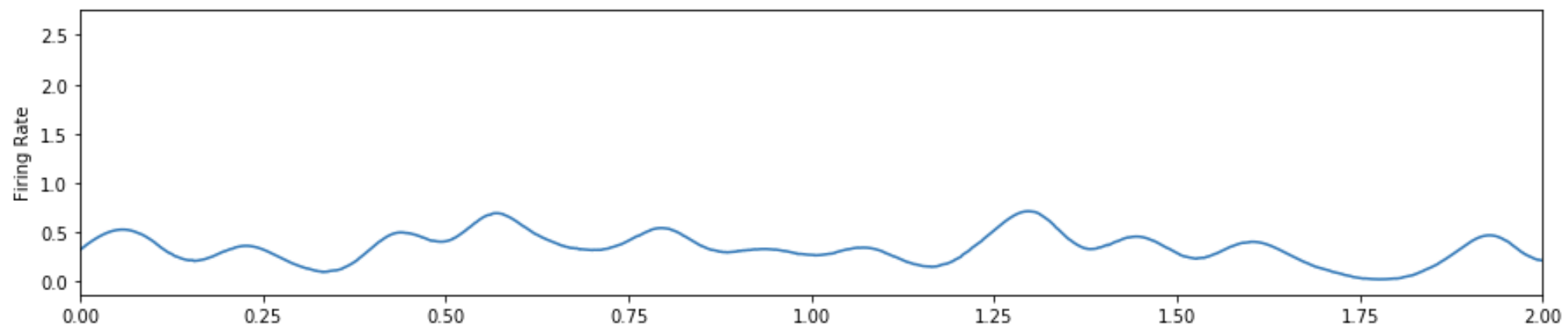
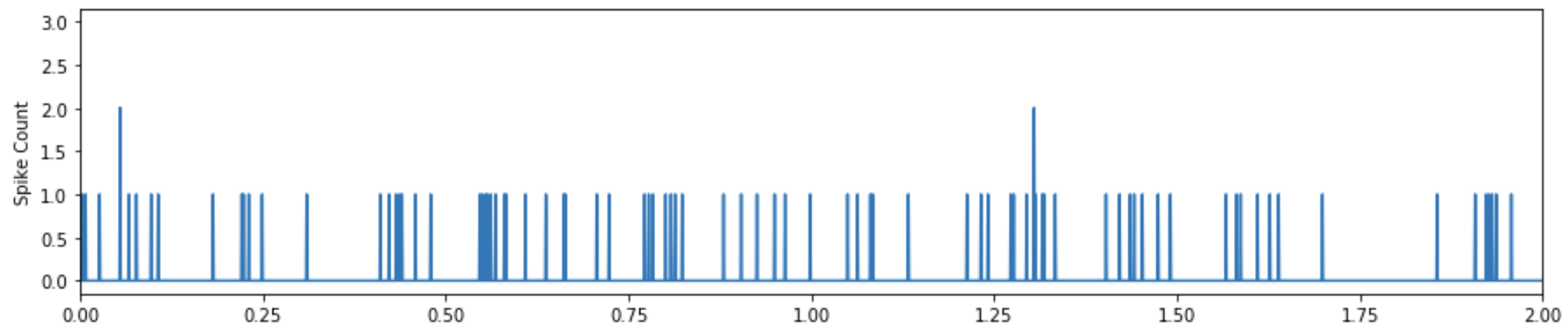
Firing Rate & LFP Analysis

LFP



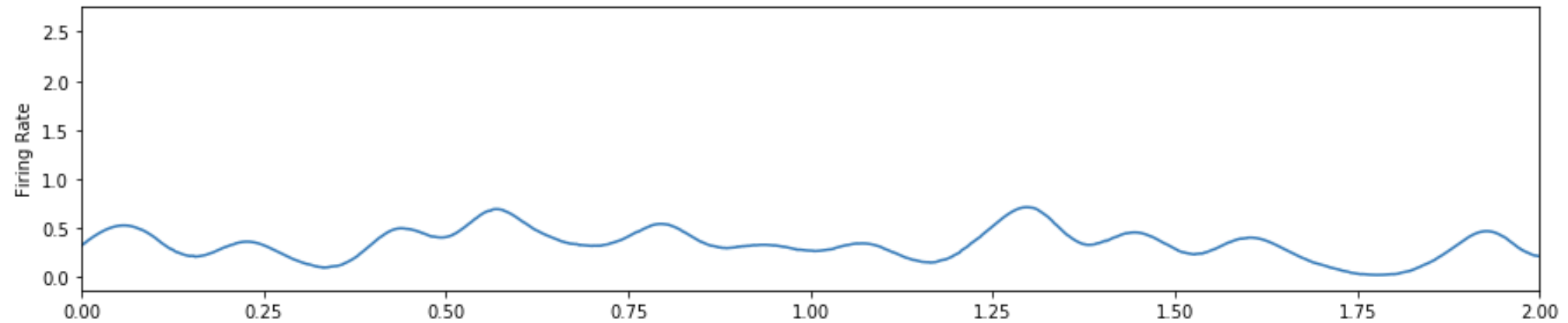
spikes: | ||| | ||| | | || | | ||| | ||| | ||| | ||| | | |

Use same time bin width (dt) to bin spike counts

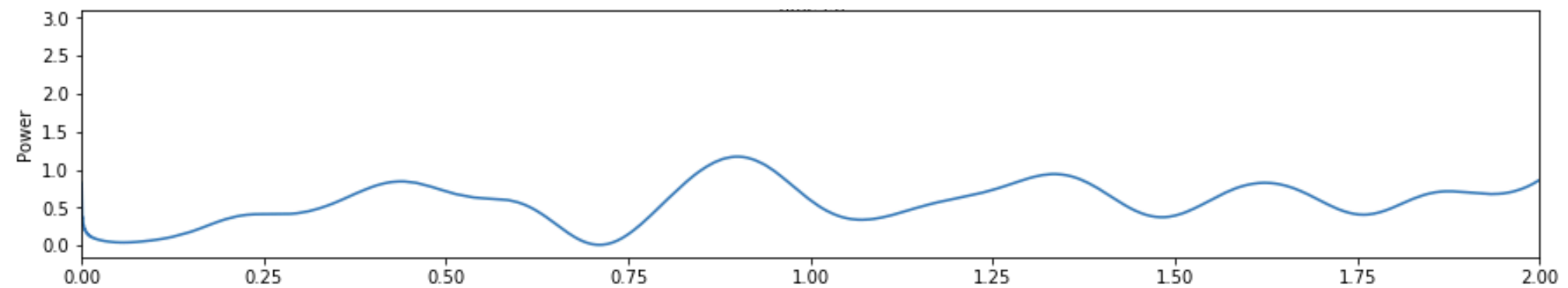


Firing Rate & LFP Analysis

Firing Rate



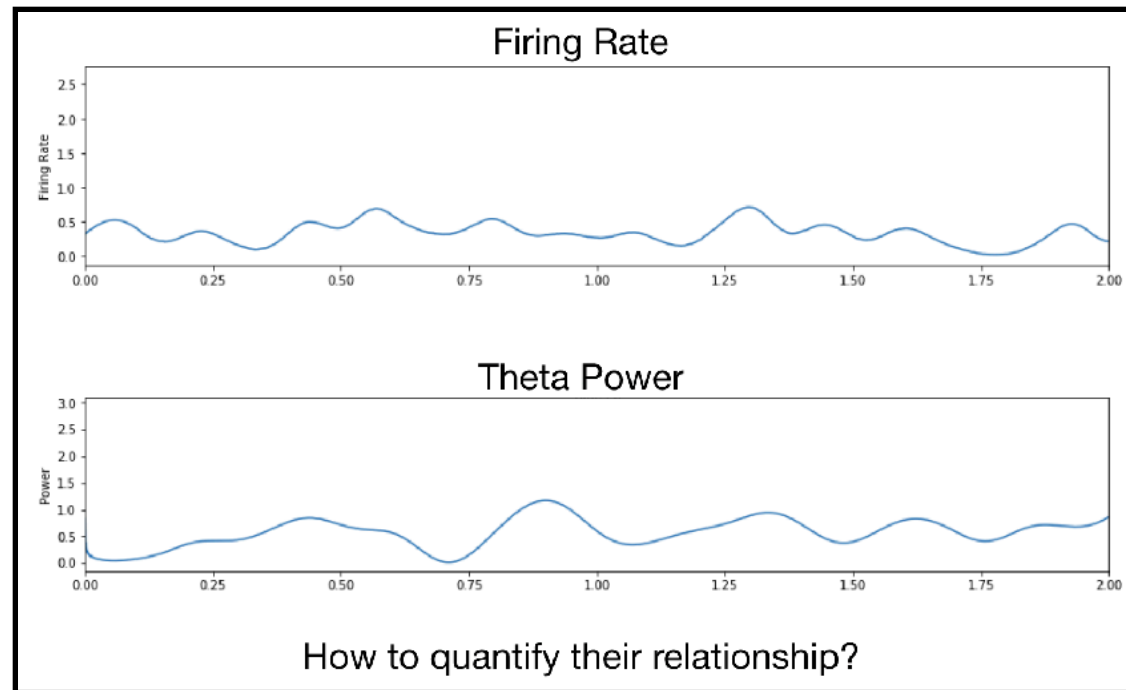
Theta Power



How to quantify their relationship?

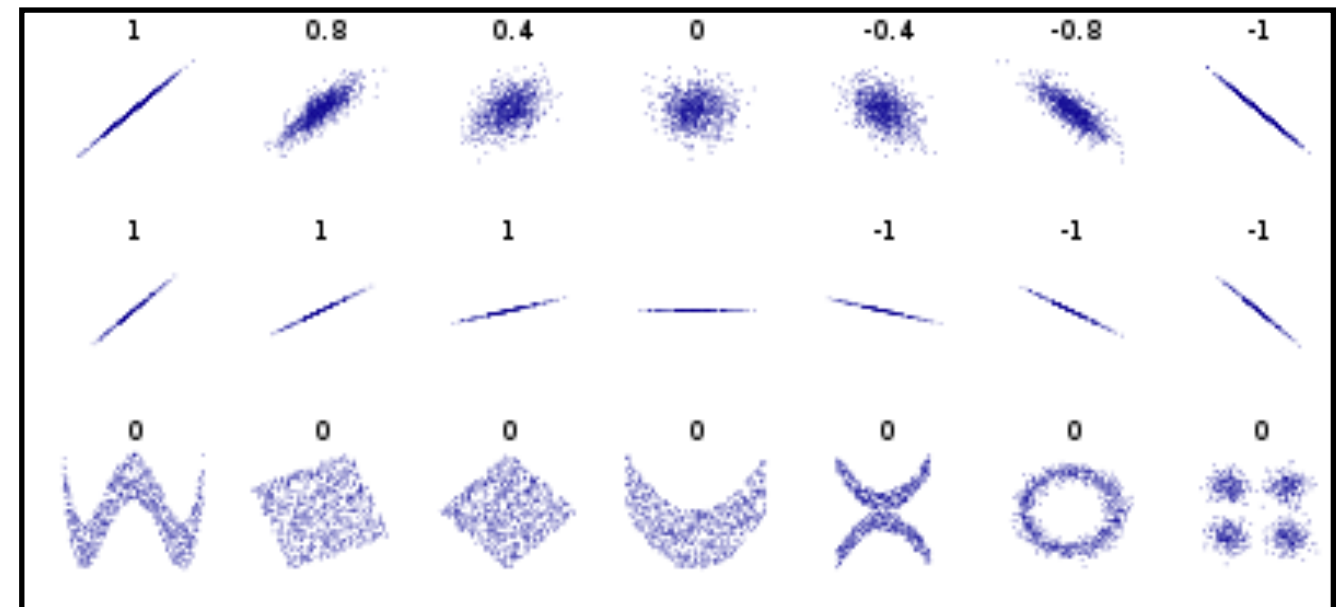


Correlation



Pearson Correlation Coefficient

$$\rho_{X,Y} = \text{corr}(X, Y) = \frac{\text{cov}(X, Y)}{\sigma_X \sigma_Y} = \frac{E[(X - \mu_X)(Y - \mu_Y)]}{\sigma_X \sigma_Y}$$



For two discrete signals x and y :

$$r_{xy} \stackrel{\text{def}}{=} \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{(n-1)s_x s_y} = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{\sqrt{\sum_{i=1}^n (x_i - \bar{x})^2 \sum_{i=1}^n (y_i - \bar{y})^2}}$$



For two discrete signals x and y :

$$r_{xy} \stackrel{\text{def}}{=} \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{(n-1)s_x s_y} = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{\sqrt{\sum_{i=1}^n (x_i - \bar{x})^2 \sum_{i=1}^n (y_i - \bar{y})^2}},$$

↑
standard deviation
of x and y

covariance

$$\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})$$

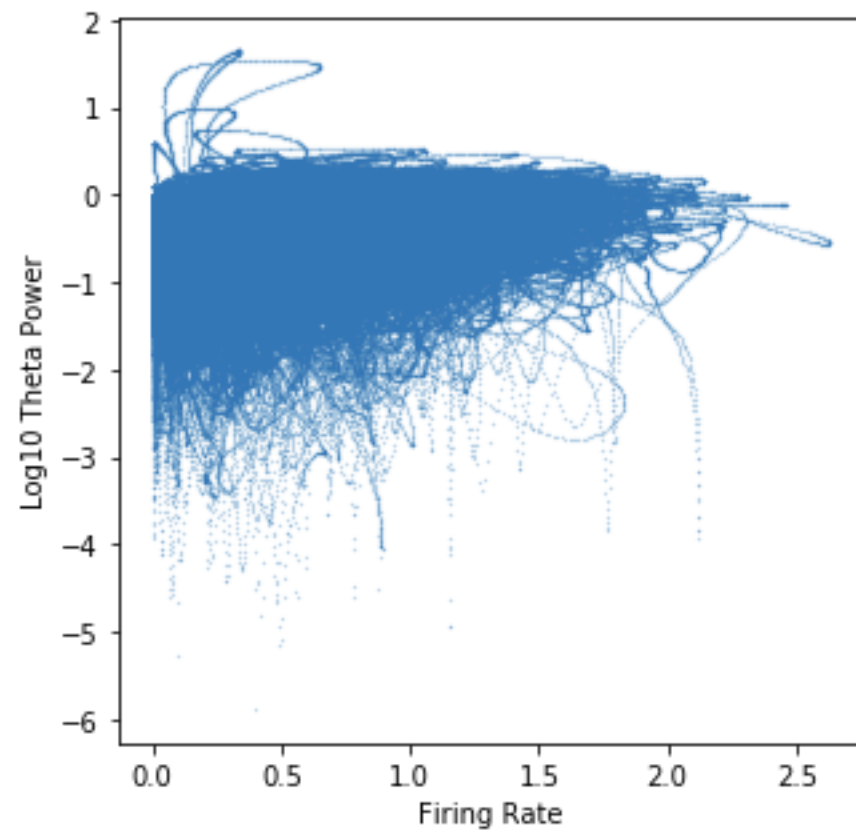
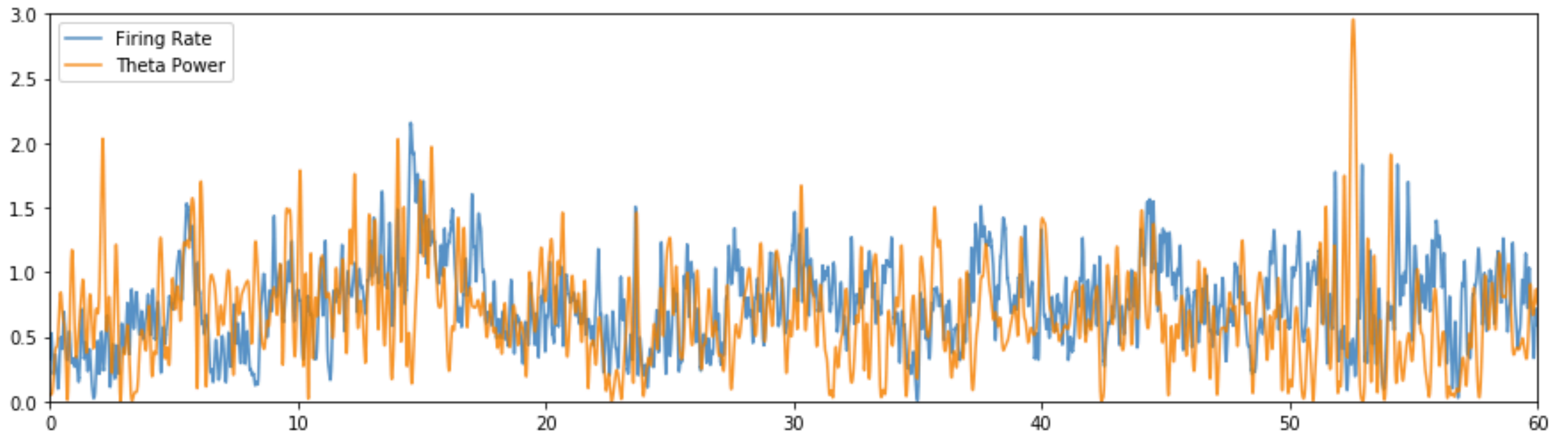
what is this operation if $\bar{x} = 0, \bar{y} = 0$

dot product!

Pearson correlation coefficient is the dot product between 2 mean-subtracted signals, normalized by their standard deviations



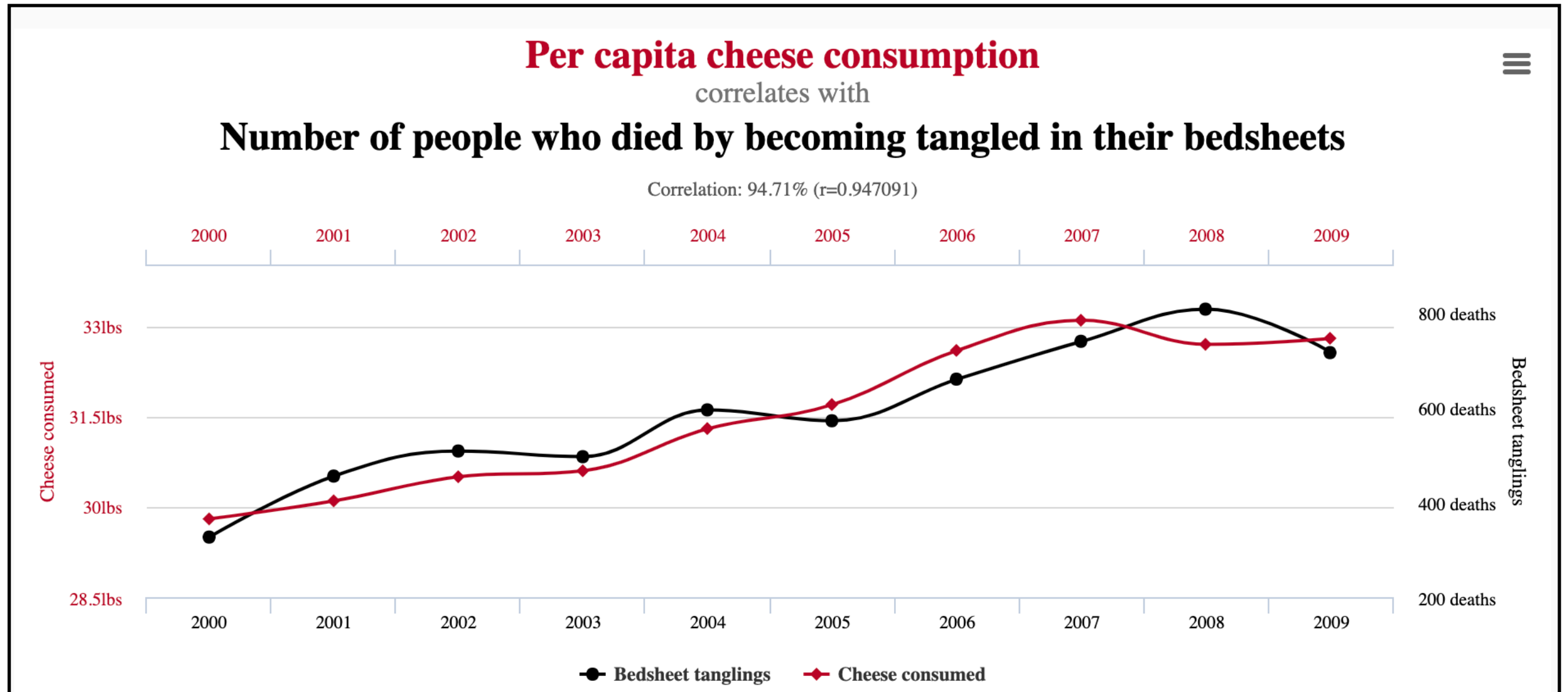
Rate-Power Correlation



np.corrcoef()



Correlation Does Not Imply Causation



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<https://tinyurl.com/cogs118c-att>

