#### **SYDE 556/750**

#### Simulating Neurobiological Systems Lecture 4: Temporal Representations

Chris Eliasmith

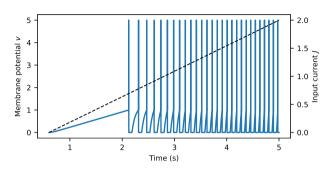
Sept 18 & 23, 2024

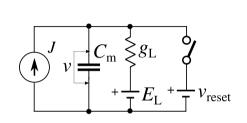
- ► Slide design: Andreas Stöckel
- ► Content: Terry Stewart, Andreas Stöckel, Chris Eliasmith





#### Reminder: The LIF Neuron





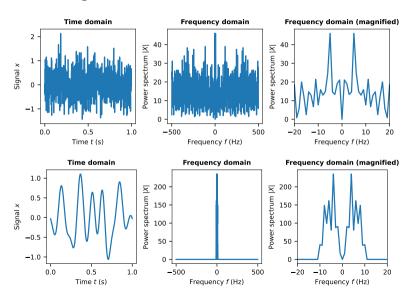
$$egin{aligned} rac{\mathrm{d}}{\mathrm{d}t} v(t) &= -rac{1}{ au_{\mathrm{RC}}} ig( v(t) - J ig) \,, \\ v(t) &\leftarrow \delta(t-t_{\mathrm{th}}) \,, \\ v(t) &\leftarrow 0 \,, \end{aligned}$$

if 
$$v(t) < 1 \, ,$$
 if  $t = t_{
m th} \, ,$  if  $t > t_{
m th}$  and  $t \geq t_{
m th} + au_{
m ref} \, ,$ 

#### Temporal Decoding

- For population decoders, we needed to integrate their responses, a(x), over the represented variable, x.
- For temporal decoders, we will likely want to integrate their responses,  $\mathbf{a}(t)$ , over the represented variable,  $\mathbf{x}(t)$ .
- What space do we want to sample to estimate the integrals?

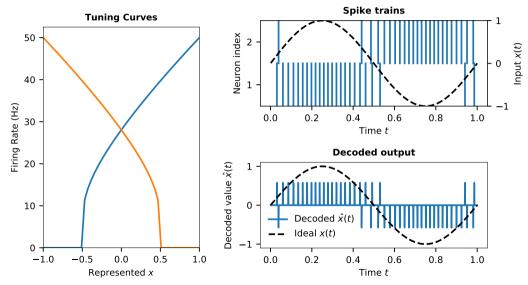
# Random Signals



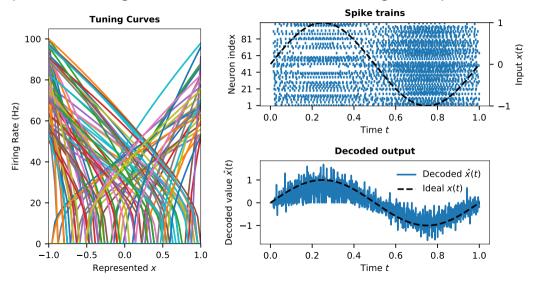
White Noise (zero mean)

Bandlimited
White Noise
(zero mean,
10 Hz bandwidth)

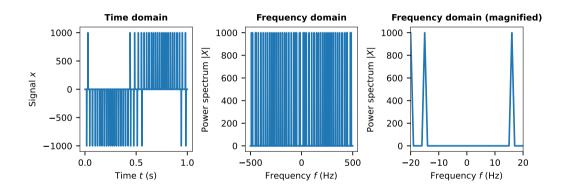
# Temporal Decoding of Two Neurons - Weighted Spikes



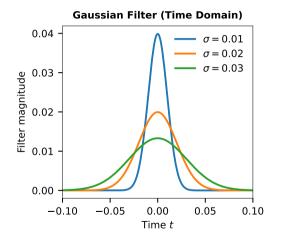
#### Temporal Decoding of One Hundred Neurons - Weighted Spikes



#### Frequency Response of Two Neurons



#### Filtering by Convolution



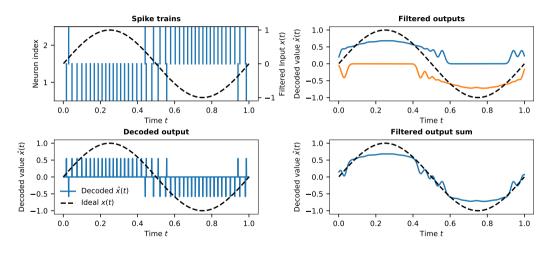
Gaussian Filter

$$h(t)=c\exp\left(rac{-t^2}{\sigma^2}
ight)$$
 where  $c$  chosen s.t.  $\int_{-\infty}^{\infty}h(t)\,\mathrm{d}t=1$ 

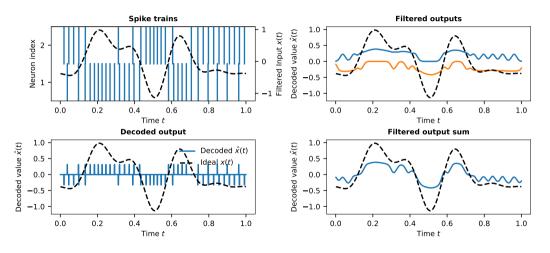
Convolution

$$(f*h)(t) = \int_{-\infty}^{\infty} f(t-t')h(t') dt'$$

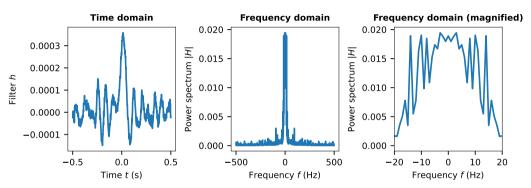
#### Filtering a Spike Train



# Filtering a Spike Train for a Random Signal

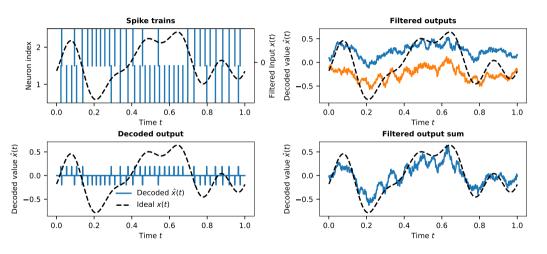


#### Optimal Filter

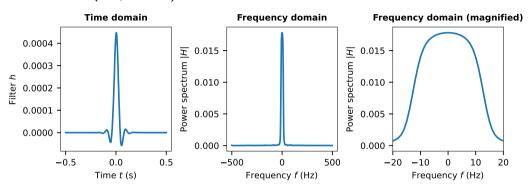


$$H(\omega) = \frac{X(\omega)\overline{R}(\omega)}{|R(\omega)|^2}$$

# Filtering a Spike Train for a Random Signal (Optimal Filter)

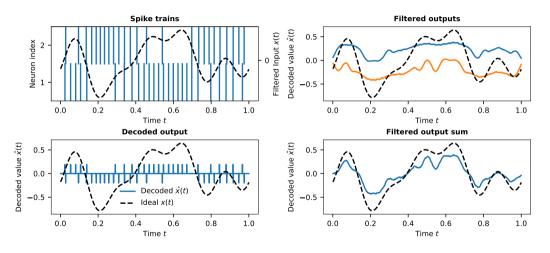


#### Optimal Filter (Improved)

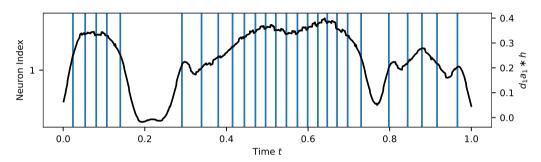


$$H(\omega) = \frac{X(\omega)\overline{R}(\omega) * W(\omega)}{|R(\omega)|^2 * W(\omega)}$$

# Filtering a Spike Train for a Random Signal (Improved Optimal Filter)



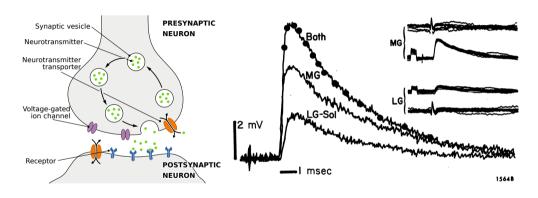
#### Pros and Cons of the Optimal Filter



Precise Good for analysing data after the fact Non-causal
 Does not describe a biological process

We need to find a mechanism that low-pass filters spikes over time!

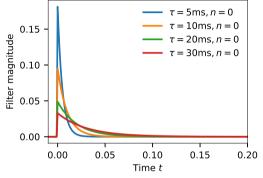
# Synapses as Filters



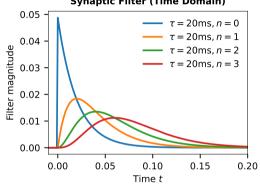
Post-synaptic currents (EPSCs, IPSCs) are low-pass filtered spike trains!

# Exponential Low-Pass Filter (I)

#### Synaptic Filter (Time Domain)



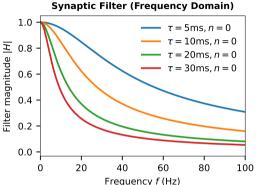
#### Synaptic Filter (Time Domain)



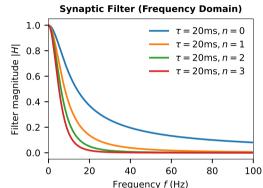
$$h(t) = egin{cases} c^{-1}t^n \exp^{-t/ au} & ext{if } t \geq 0\,, \ 0 & ext{otherwise}\,, \end{cases}$$

where 
$$c = \int_0^\infty t^n \exp^{-t/\tau} dt$$
.

#### Exponential Low-Pass Filter (II)

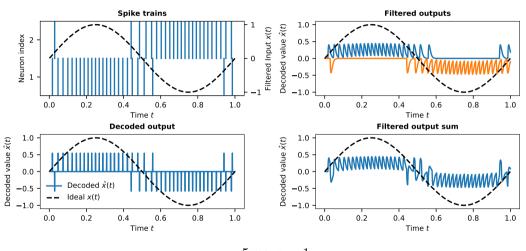


$$\mathit{h}(t) = egin{cases} c^{-1}t^n \exp^{-t/ au} & ext{if } t \geq 0\,, \ 0 & ext{otherwise}\,, \end{cases}$$



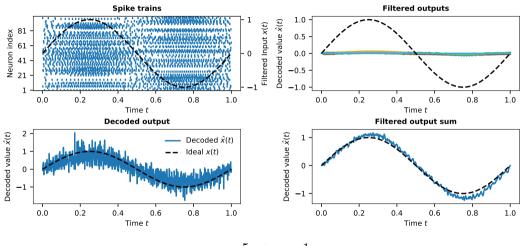
where 
$$c = \int_0^\infty t^n \exp^{-t/\tau} dt$$
 .

#### Example: Synaptic Filter for Two Neurons



$$\tau=5\,\mathrm{ms}, \textit{n}=1$$

#### Example: Synaptic Filter for One Hundred Neurons



#### Image sources

#### \_\_\_\_\_

From Wikimedia.

Title slide
"Continue halloon with clock face and hall floating above the Fiffal Tower Paris France"

"Captive balloon with clock face and bell, floating above the Eiffel Tower, Paris, France." Author: Camille Grávis, between 1889 and 1900.