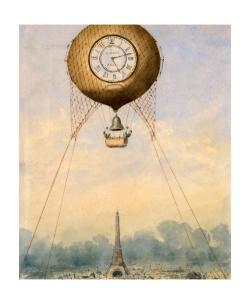
SYDE 556/750

Simulating Neurobiological Systems Lecture 4: Temporal Representations

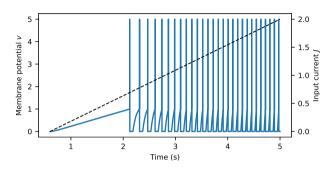
Andreas Stöckel

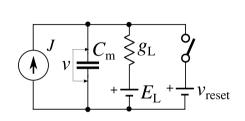
January 22 & 28, 2020





Reminder: The LIF Neuron

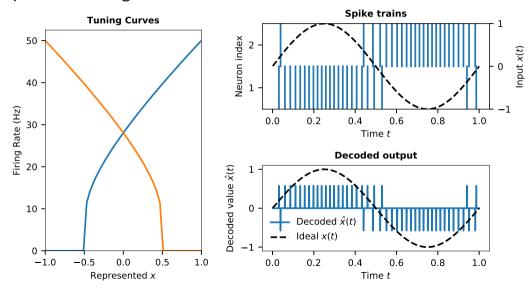




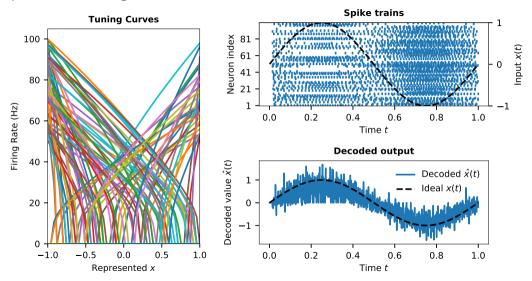
$$egin{aligned} rac{\mathrm{d}}{\mathrm{d}t} v(t) &= -rac{1}{ au_{\mathrm{RC}}} ig(v(t) - J ig) \,, \ v(t) &\leftarrow \delta(t - t_{\mathrm{th}}) \,, \ v(t) &\leftarrow 0 \,, \end{aligned}$$

$$\begin{array}{l} \text{if } \textit{v}(t) < 1 \,, \\ \\ \text{if } \textit{t} = \textit{t}_{\rm th} \,, \\ \\ \text{if } \textit{t} > \textit{t}_{\rm th} \text{ and } \textit{t} \geq \textit{t}_{\rm th} + \tau_{\rm ref} \,, \end{array}$$

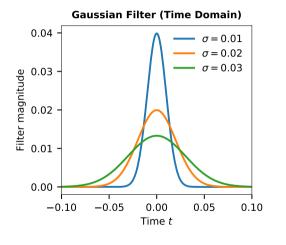
Temporal Decoding of Two Neurons



Temporal Decoding of One Hundred Neurons



Filtering by Convolution



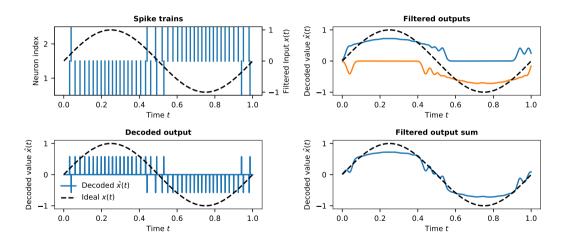
Gaussian Filter

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

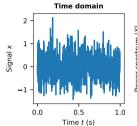
Convolution

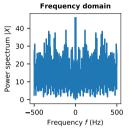
$$(f * g)(t) = \int_{-\infty}^{\infty} f(t - \tau)g(\tau) d\tau$$

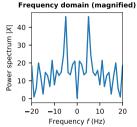
Filtering a Spike Train



Random Signals



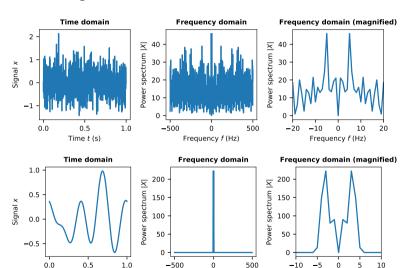




White Noise (zero mean)

Random Signals

Time t (s)



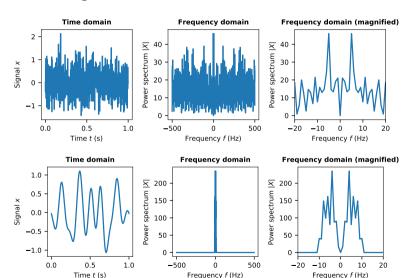
Frequency f (Hz)

Frequency f(Hz)

White Noise (zero mean)

Bandlimited
White Noise
(zero mean,
5 Hz bandwidth)

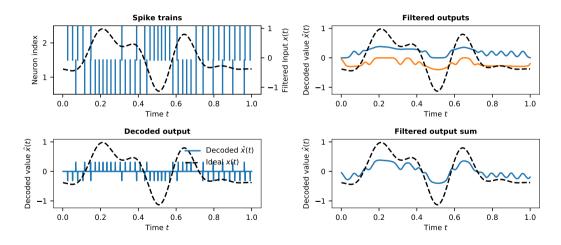
Random Signals



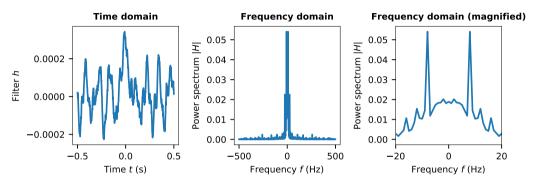
White Noise (zero mean)

Bandlimited White Noise (zero mean, 10 Hz bandwidth)

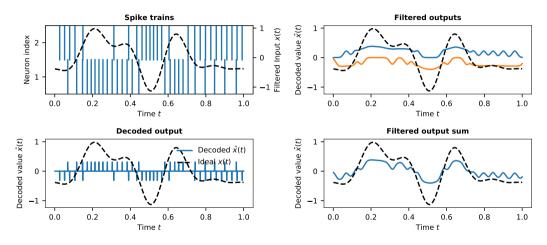
Filtering a Spike Train for a Random Signal

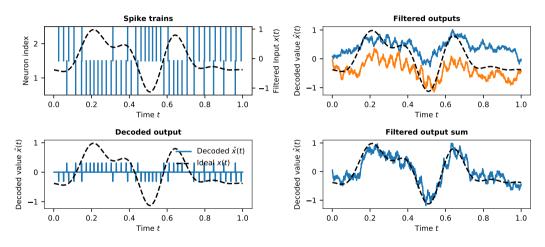


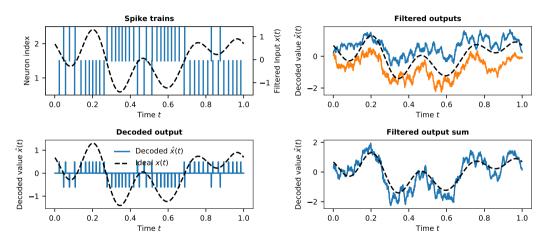
Optimal Filter

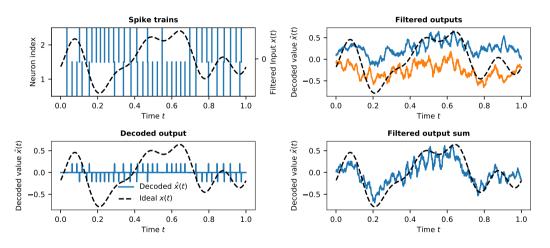


$$H(\omega) = \frac{X(\omega)\overline{R}(\omega)}{|R(\omega)|^2}$$

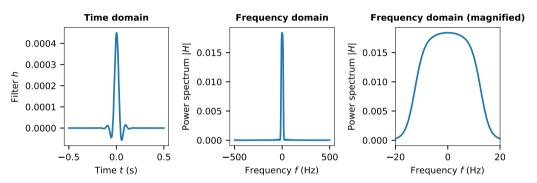




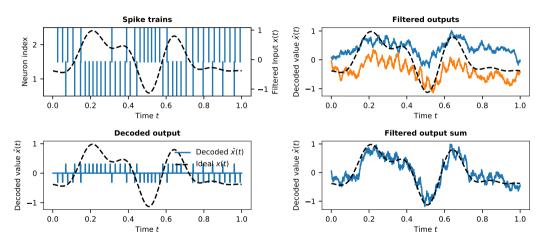


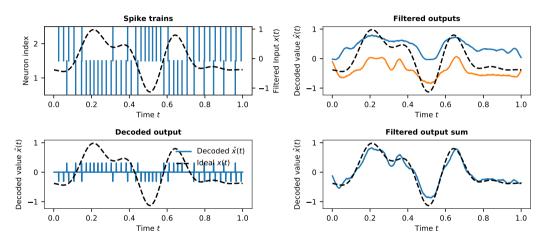


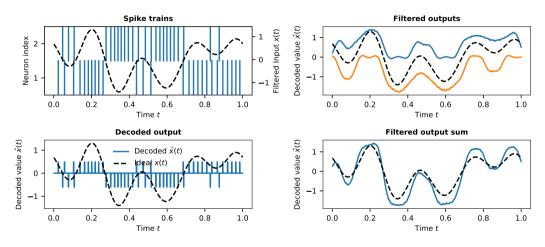
Optimal Filter (Improved)

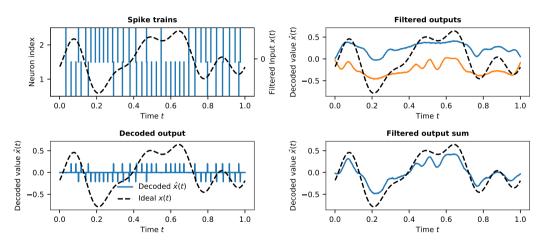


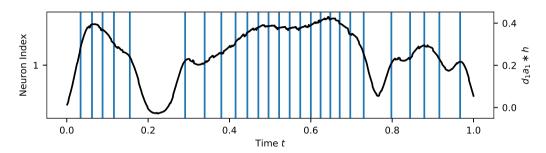
$$H(\omega) = \frac{X(\omega)\overline{R}(\omega) * W(\omega)}{|R(\omega)|^2 * W(\omega)}$$

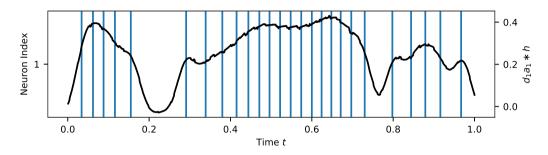




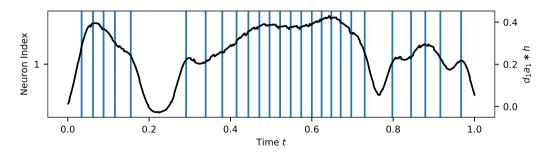






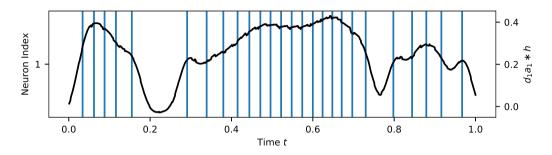


Precise
Good for analysing data after the fact



Precise
Good for analysing data after the fact

Non-causal
 Does not describe a biological process



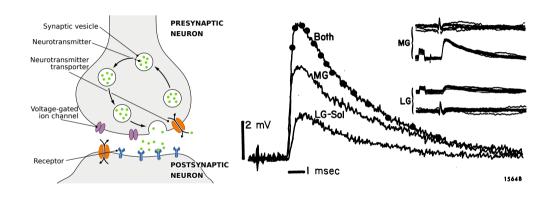
Precise
Good for analysing data after the fact

Non-causal

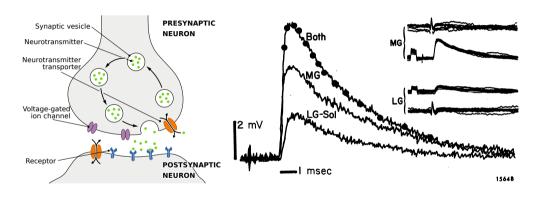
Does not describe a biological process

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

Synapses as Filters



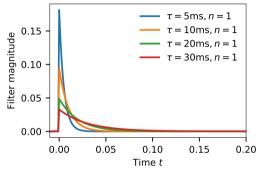
Synapses as Filters



Post-synaptic currents (EPSPs, IPSPs) are low-pass filtered spike trains!

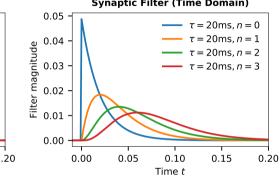
Exponential Low-Pass Filter (I)

Synaptic Filter (Time Domain)



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

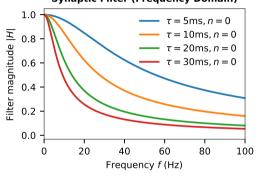
Synaptic Filter (Time Domain)



where
$$c=\int_0^\infty t^n \exp^{-t/ au}\,\mathrm{d}t$$
 .

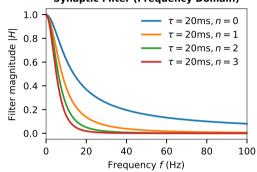
Exponential Low-Pass Filter (II)





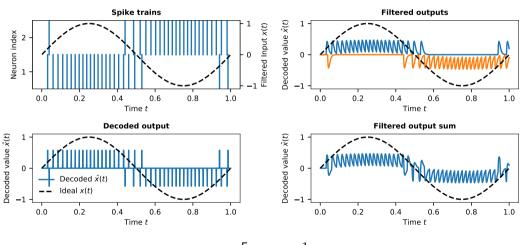
$$h(t) = egin{cases} c^{-1}t^n \exp^{-t/ au} & ext{if } t \geq 0\,, \ 0 & ext{otherwise}\,, \end{cases}$$

Synaptic Filter (Frequency Domain)



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

Example: Synaptic Filter for Two Neurons



Example: Synaptic Filter for One Hundred Neurons

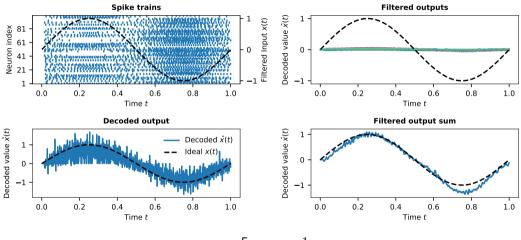


Image sources

Title slide

"Captive balloon with clock face and bell, floating above the Eiffel Tower, Paris, France."

Author: Camille Grávis, between 1889 and 1900.

From Wikimedia.