## *COMS30017* **Computational Neuroscience**

Week 4: Synapses and Synaptic Plasticity

Dr. Beatriz E. P. Mizusaki fv18192@bristol.ac.uk

# Video 2: COMPUTATIONAL MODELLING OF A SYNAPSE



## Intended learning outcomes

- Understand different abstract representations of synapses; • Be able to write down a simple model for a synaptic input;

# Synapses are complex



Sheng & Hoogenraad, Annu Rev Neurosci (2007)

# Computational model of a synapse



### Many possible levels of abstraction:

 $\nu_j = \sum W_i \nu_i$ 

## MCell simulation of synaptic release



Bartol et al, Frontiers Syn Neuro (2015)





AMPAR

С



L-type and R-type VDCC

NMDAR-GluN2A/2B

TO TI T1

**GluT GLT-1/GLAST** 

#### Simulations at molecular level are possible: detailed but computationally expensive



- ·We could simulate the dynamics of each molecule involved in the signalling process (like the MCell simulation).
- But since that is (very) computationally expensive, we might instead go for a reduced mass-action chemical-kinetics model.
- However for many purposes that is still too expensive and parameter heavy, so instead we use even simpler phenomenological models that black-box the synapse as a simple input-output system.

## Model of a synapse

# What goes is model of a synapse?

- $\cdot$  Synapses are unreliable  $\rightarrow$  transmission is probabilistic;
- ·Weighted transformation of input amplitudes, non-linear;
- ·Can be excitatory or inhibitory (depending on presynaptic neuron);

# Simple model of a synapse

 $\cdot$  The most common way to phenomenologically model a synapse is as a time-dependent conductor in series with a battery.  $^{\circ}$ 

$$I_s(t) = \bar{g_s}s(t)[E_s - V(t)]$$

s = exc, inh

- The value of  $E_s$  determines whether the synapse is excitatory or inhibitory: • for excitatory synapses  $E_s$  usually = 0 mV • for inhibitory synapses  $E_s$  usually = V<sub>rest</sub>
- But how should we model s(t)?



# Simple model of a synapse

#### Single exponential

$$s(t) \rightarrow s(t) + 1$$
  
 $s(t) = e^{-t/\tau_s}$ 

# Alpha function $s(t) = te^{-t/\tau_s}$

### Difference of two exponentials $s(t) = e^{-t/\tau_{decay}} - e^{-t/\tau_{rise}}$



# Summary

- Real synapses are complicated molecular machines.
- We can model them at multiple levels of granularity, as appropriate for the task at hand.
- A simple way of modelling a PSP on the postsynaptic membrane potential is with the addition of a single impulse to the current.