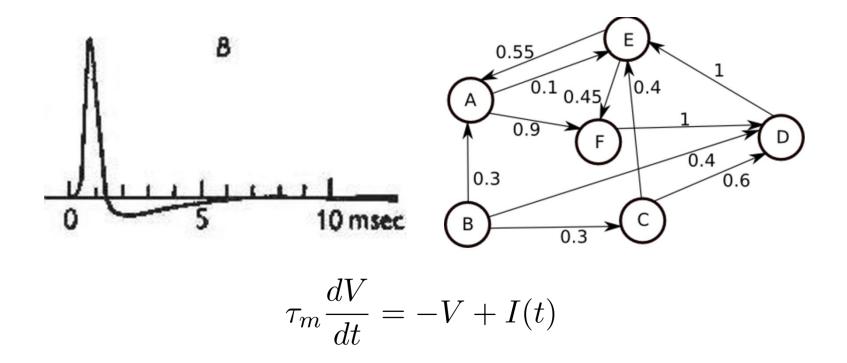
Plastic Neural Network Simulations

Justin Skycak

Neuron Dynamics

Excitatory Refractory Leaky Integrate & Fire



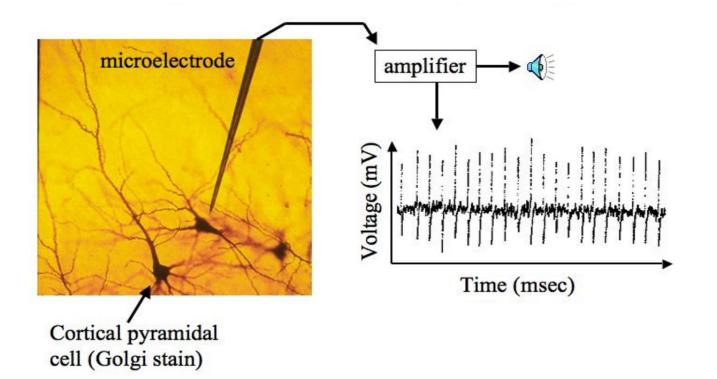
I(t) = random noise + weighted input from spiking neighbors

When V = Vth (spike), set V \rightarrow 0 and hold for duration R (refractory period)

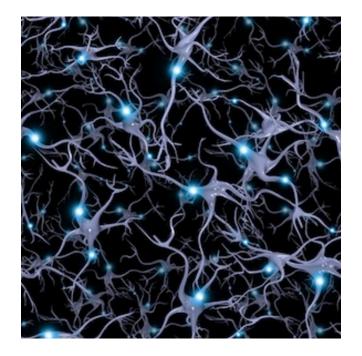
Weights between 0 and 1; weighted input = weight * Vth

Stimulus

Each stimulated neuron spikes every p milliseconds (assuming not in refractory period)

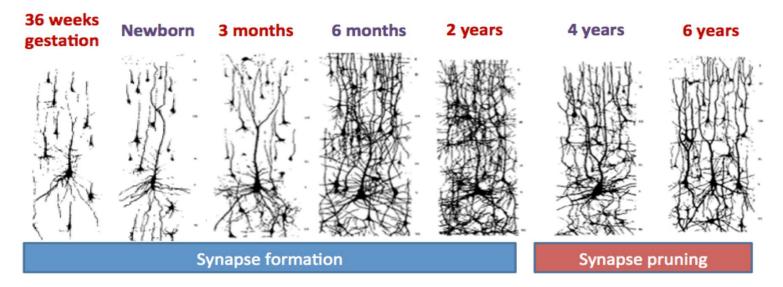


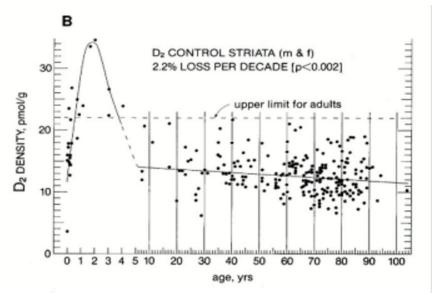
We have a biologically-plausible spiking network.



But it's boring because its connectivity is fixed.

Developmental Plasticity



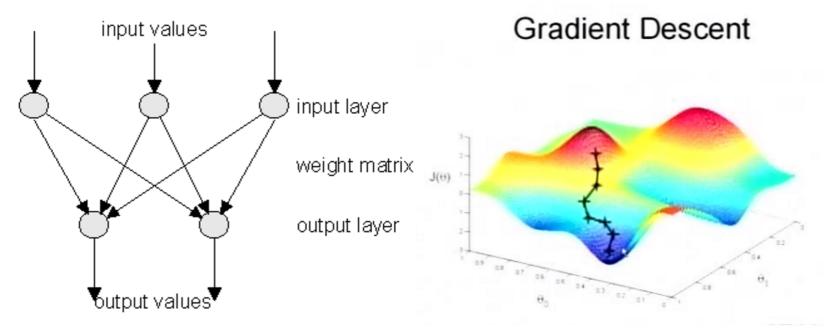


Rate of synapse formation skyrockets during infancy.

HOWEVER, synapses continue to form across entire lifespan.

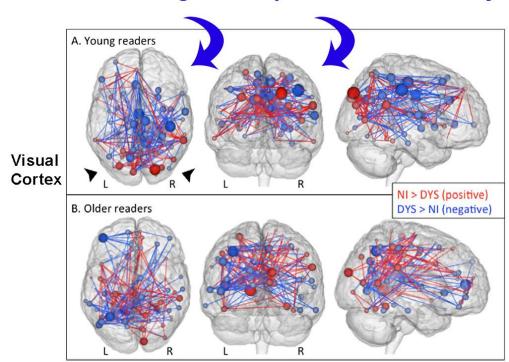
Learning = modified connectivity

Weight modification can implement gradient descent on feed-forward artificial neural networks



Common artificial learning rule: backpropagation (can be extended to recurrent ANNs as well – "backpropagation through time")

Connectivity often implicated in brain disorders



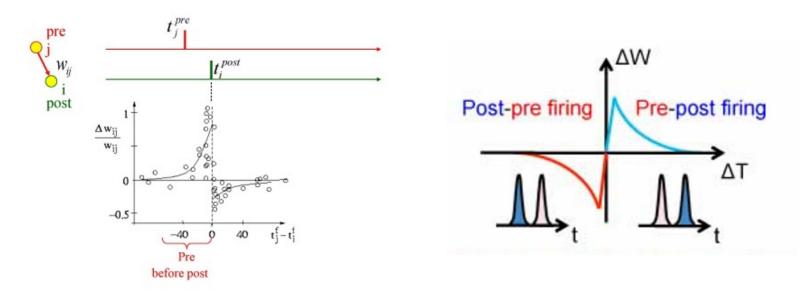
Right Hemisphere Connections - Dyslexia

Biological Synapse Modification Spike-Timing Dependent Plasticity

Hebbian learning: "Neurons that fire together wire together"

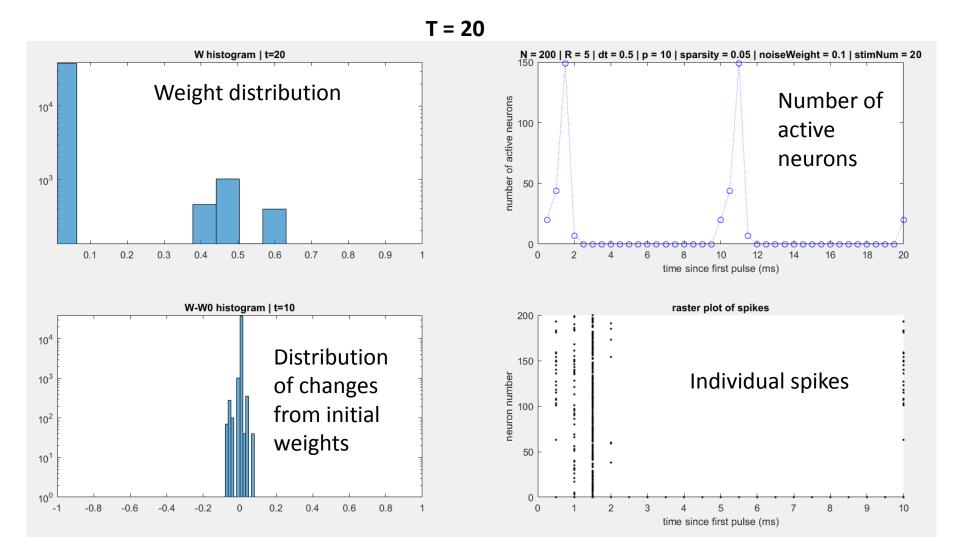
STDP learning: like Hebbian, but depends on order of spiking

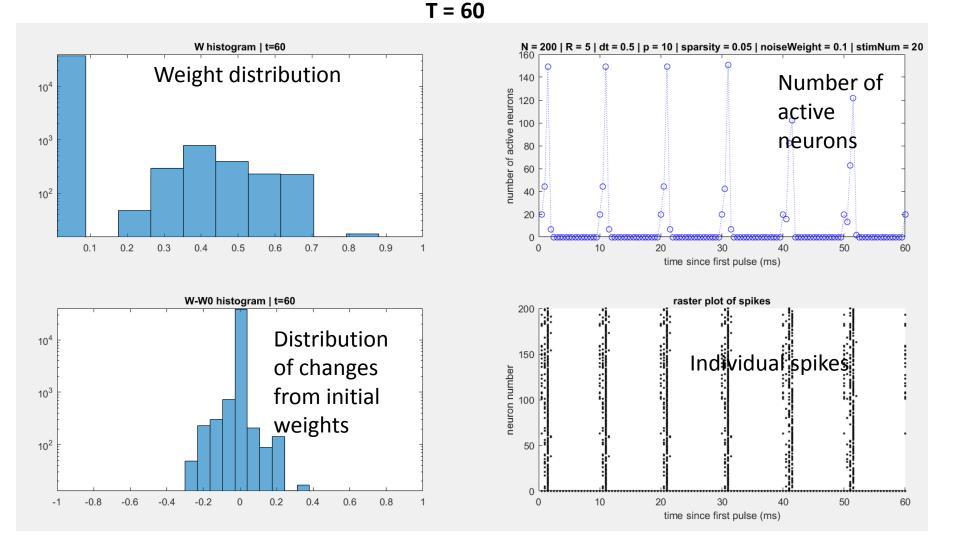
If I predict you, I'll strengthen my weight to you. But if you predict me, I'll weaken my weight to you.

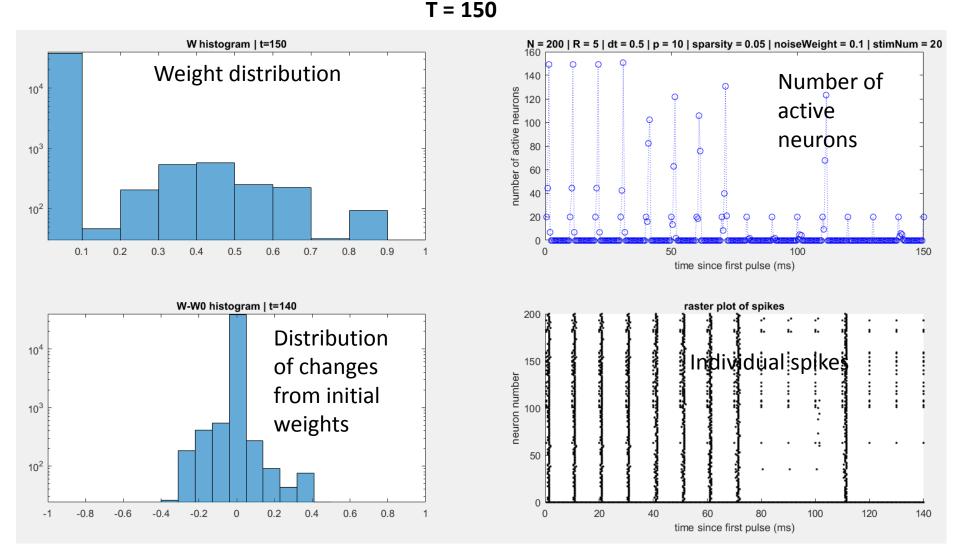


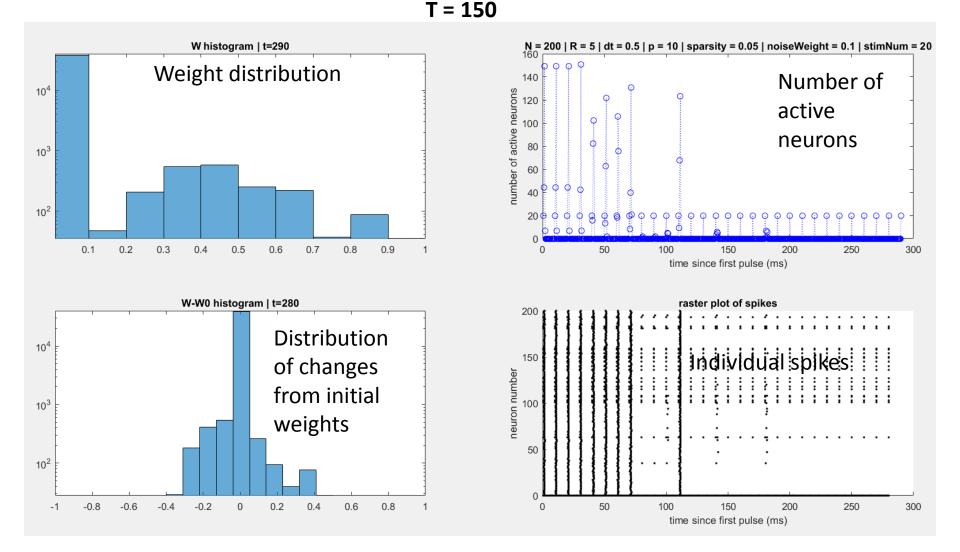
My simulations

- 1. Randomly select *sparsity*% of connections to exist (set their weight to 0.5)
- 2. Set all other weight to epsilon (~0.01)
- 3. Simulate network and see what happens to
 - Weight distribution
 - Network activity









Seizure?

