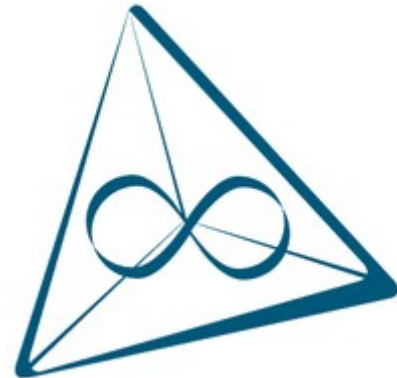


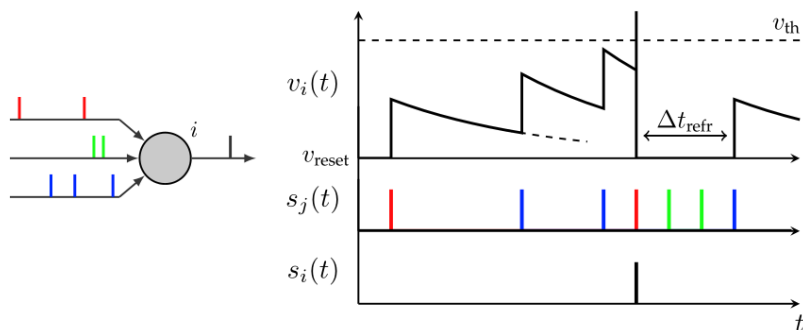
# STDP leads to efficient coding of predictions

Pau Vilimelis Aceituno  
Masud Ehsani  
Jürgen Jost



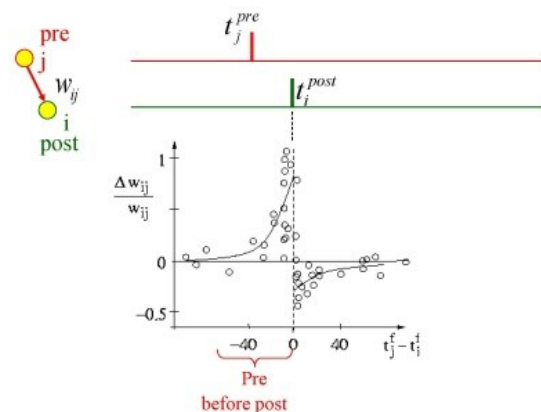
# Models

- Leaky Integrate and Fire Neurons



Paredes-Vallés et al., *IEEE Trans. In Pattern Analysis and Machine Intelligence*

- Spike-Time Dependent Plasticity

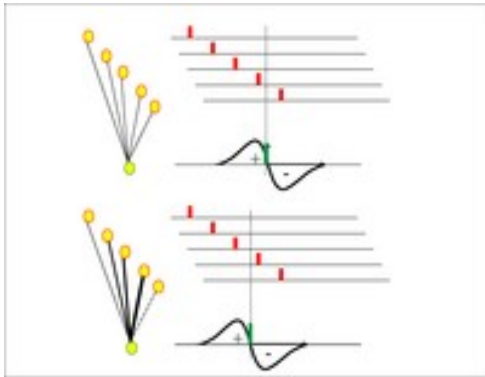


Sjöström & Gerstner, *Scholarpedia*

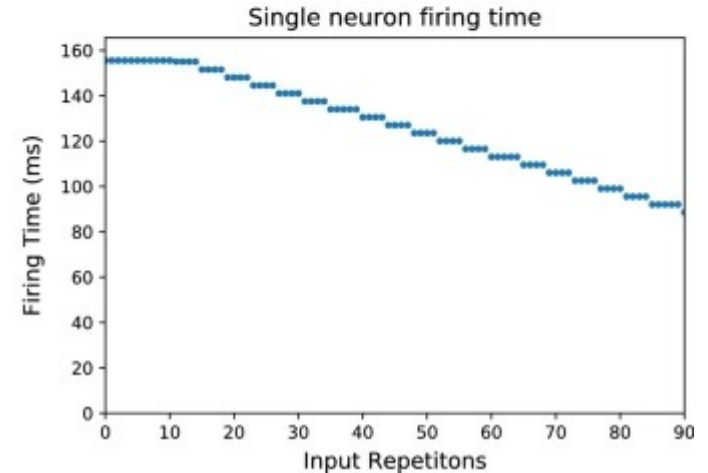
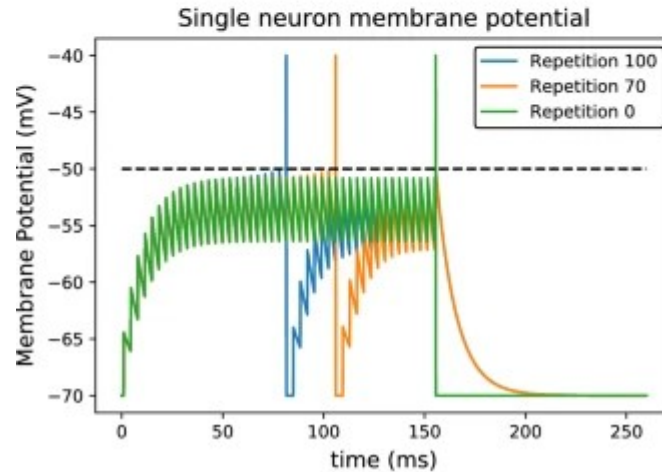
- Input Spike trains:
  - Few spikes per presynaptic synapse
  - Input time random, but fixed across repetitions
  - E/I, noise, multiple input spike trains

# Spike evolution: Latency

- STDP reinforces inputs arriving *before* a postsynaptic spike, creating an *earlier* spike
- Observed in hippocampus (Mehta et al. *Neuron*)

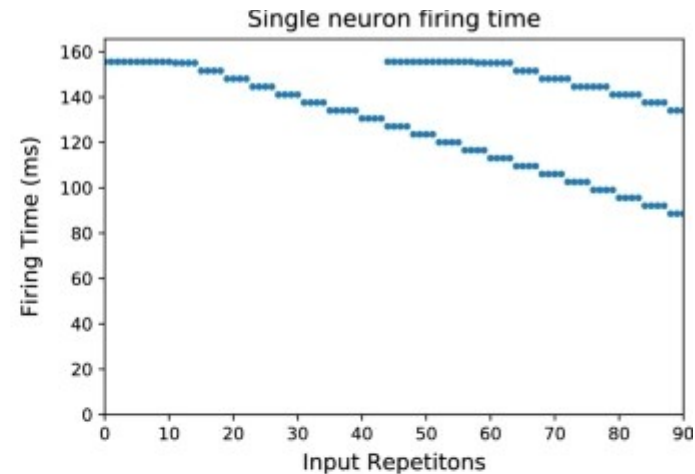
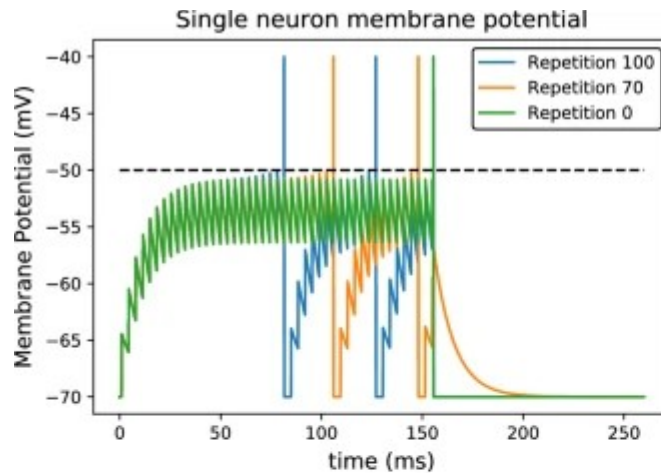


Spike-Timing Dependent Plasticity, Sjöström & Gestner, *Scholarpedia*



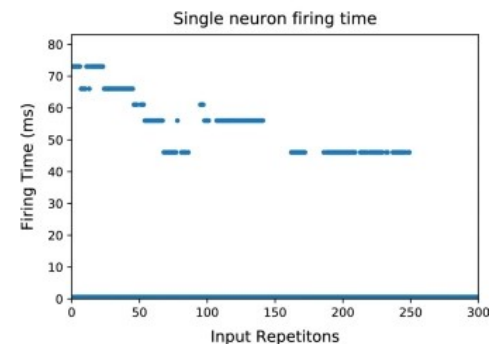
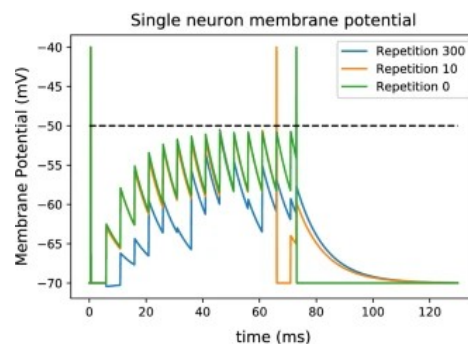
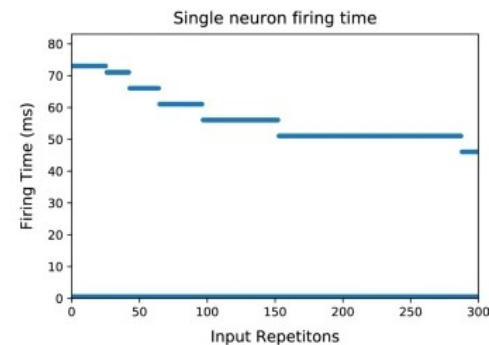
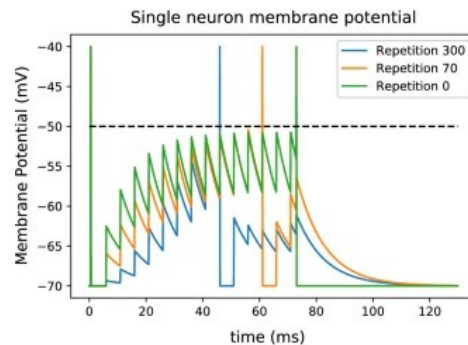
# Spike evolution: proliferation

- If a spike moves forward in time without altering the inputs (antisymmetric STDP), the original spike will reappear  
→ presynaptic depression > presynaptic potentiation



# Spike evolution: disappearance

- When two postsynaptic spikes are close in time the second one tends to disappear
- But only if there is some noise
- Similar to triplet rules

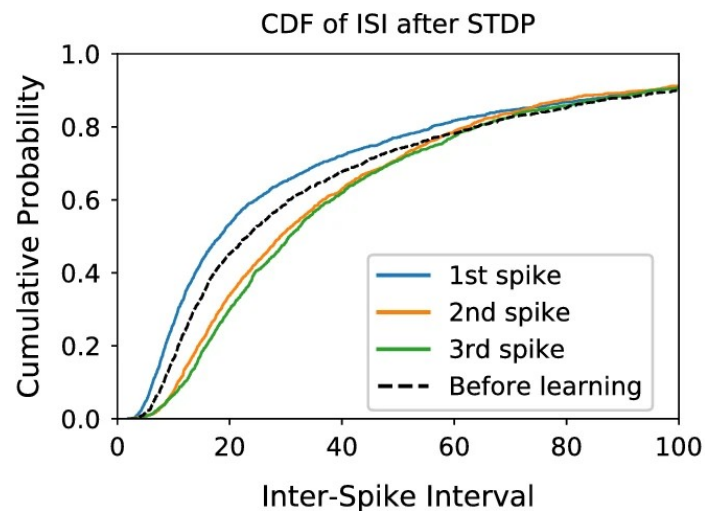
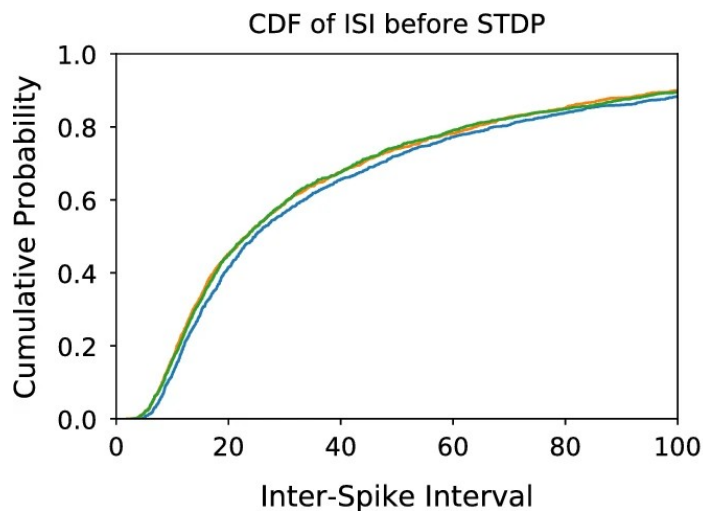


# Spike Trains: technicalities

- All the phenomena assume that postsynaptic spikes evolve **independently** → Constraints on in/out rate
  - Depression/potential happen only from a single pre/post pair
  - Synapses with multiple spikes don't overlap in time
- **Inhibitory/Excitatory balance** can be maintained
  - If the inhibitory plasticity fulfills specific conditions
- Multiple variants of plasticity yield same results
  - Triplet rules, multiple inhibitory variants, short-term plasticity...

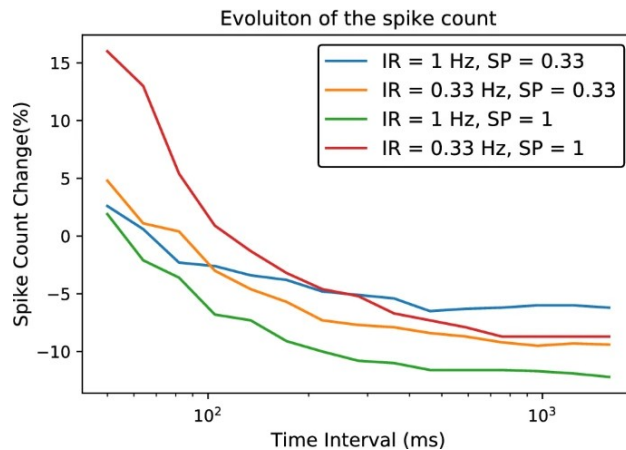
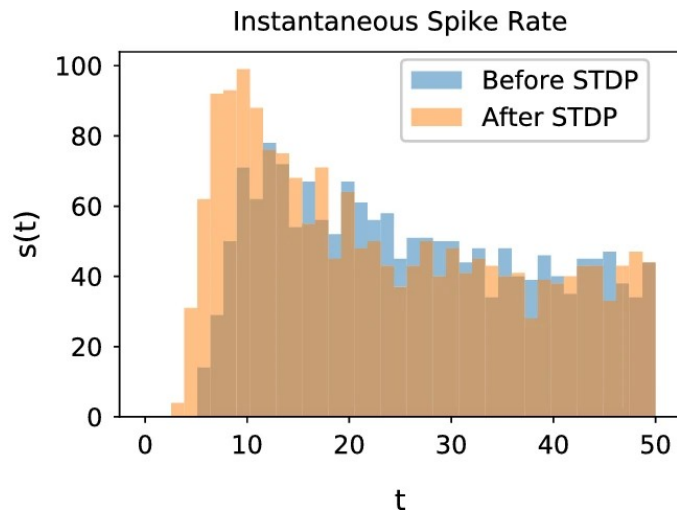
# Spike train evolution I

- Its a point process in time → Look at the CDF of the inter-spike interval
- First spike moves forward, later spikes spread apart



# Spike train evolution

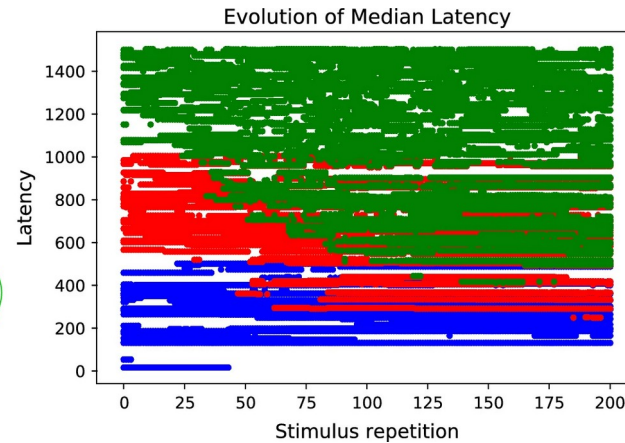
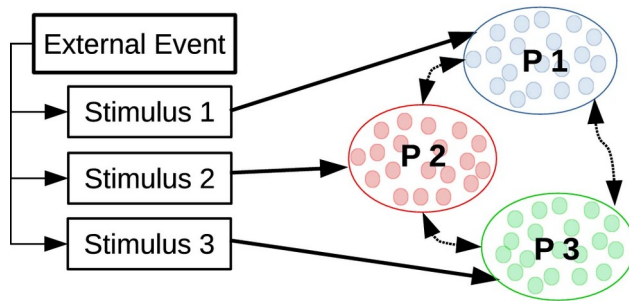
- Peak of spikes at the beginning of the spike train
- Total number of spikes decreases if the time interval is long and the input rate is high





# Predictions

- Spikes happen earlier and earlier until they happen *before* the stimulus that they originally encoded



# Conclusion & Next Steps

## Summary

- Spikes move forward in time through STDP
- A peak of activity appears at stimulus onset
- But the total number of spikes is reduced
- Predictions emerge

## Next Steps

- Study the efficiency of sequential multi-channel sparse codes
- Encoding predictions without predictive coding?
- Evolution of prediction?

# Acknowledgments

- Coauthors: Masud Ehsani, Jürgen Jost
- Institute: Max Planck Institute for Mathematics in the Sciences
- Funding: The Max Planck School of Cognition, Bundesministerium für Bildung und Forschung