

CNeuro2024

VINCENT HAKIM

Reading Material: Lectures 1 & 2

Reading material.

Lecture 1. The cerebellum: a computational perspective.

- Marr, D. (1969) A theory of cerebellar cortex. *J Physiol.*, 202, 437-470
- Clopath, C., et al (2014). A cerebellar learning model of vestibulo-ocular reflex adaptation in wild-type and mutant mice. *Journal of Neuroscience*, 34(21), 7203-7215.
- Litwin-Kumar, A., et al. "Optimal degrees of synaptic connectivity." *Neuron* 93.5 (2017): 1153-1164
- Lanore, F., et al. "Cerebellar granule cell axons support high-dimensional representations." *Nature neuroscience* 24.8 (2021): 1142-1150.
- Brunel, N, et al. "Optimal information storage and the distribution of synaptic weights: perceptron versus Purkinje cell." *Neuron* 43.5 (2004): 745-757.
- Barri, A., et al. "Synaptic basis of a sub-second representation of time in a neural circuit model." *Nature Communications* 13.1 (2022): 7902.

Lecture 2. The credit assignment problem and the cerebellum.

- Williams, R. J. (1992). Simple statistical gradient-following algorithms for connectionist reinforcement learning. *Machine learning*, 8, 229-256.
- Fiete, I. R., Fee, M. S., & Seung, H. S. (2007). Model of birdsong learning based on gradient estimation by dynamic perturbation of neural conductances. *Journal of neurophysiology*, 98(4), 2038-2057.
- Bouvier, G. et al. "Cerebellar learning using perturbations." *Elife* 7 (2018): e31599.
- Hiratani, N. et al. "On the stability and scalability of node perturbation learning." *Advances in Neural Information Processing Systems* 35 (2022): 31929-31941.
- Ohmae, S., & Medina, J. F. (2015). Climbing fibers encode a temporal-difference prediction error during cerebellar learning in mice. *Nature neuroscience*, 18(12), 1798-1803