

# Model based inference of synaptic plasticity rules

Yash Mehta, Danial Tyulmankov, Adithya Rajagopalan, Glenn Turner James Fitzgerald\*, Jan Funke\*







## Standard forms of synaptic plasticity



$$\Delta w = g_{\theta}(x, y, w, r) = ?$$





$$g_{\theta}^{\text{Taylor}} = \sum_{\alpha,\beta,\gamma=0}^{2} \theta_{\alpha,\beta,\gamma} x_{i}^{\alpha} y_{j}^{\beta} w_{ij}^{\gamma}.$$

#### Note,

Hebbian.....
$$xy$$
 $\theta_{110} = 1$ , $\theta_{rest} = 0$ Anti-Hebbian ..... $-xy$  $\theta_{110} = -1$ , $\theta_{rest} = 0$ Oja's ..... $xy - y^2w$  $\theta_{110} = 1$ , $\theta_{021} = -1$ , $\theta_{rest} = 0$ 





#### We can't directly measure the changes in synaptic weights during learning!

#### Obtainable measurements:

A. Neural activity









## Model (plasticity from neural activity)

$$\Delta w_{ij}^{}=g_{ heta}^{}(x_{j}^{},y_{i}^{},w_{ij}^{},r_{j}^{})$$



#### Assumptions

- 1. Single layer of feed forward neural network with plasticity
- 2. All weights evolve according to same plasticity rule,  $g_{\theta}$
- 3. Plasticity  $(g_{\theta})$  depends only on state (x, y, w, r) of the current time step

Measured neural activity





## Method overview (neural activity)

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### Method validation: simulated data

RAL INFORMATION

Generate a trajectory with a known plasticity rule, then try to infer it.





## Model (plasticity from behavior)

$$\Delta w_{ij}^{} = g_{ heta}^{}(x_{j}^{},y_{i}^{},w_{ij}^{},r_{j}^{})$$



#### Assumptions

p(accept)

- Behavior: accept, reject decisions. Given by network output
- 2. Plasticity happens in the first layer, last layer is fixed (averages neural activity)





### Method overview (behavior)







## Fly experimental setup

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#### Yash Mehta, Cognitive Science PhD student, Johns Hopkins



Paper Website



GitHub Repo



