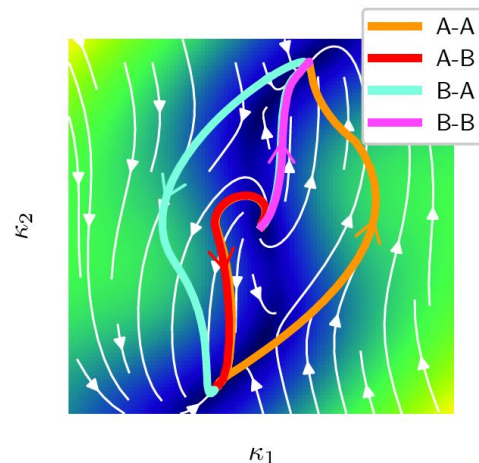
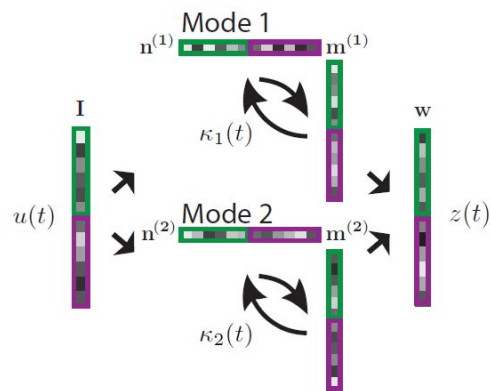
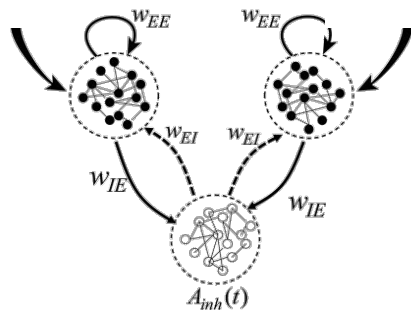
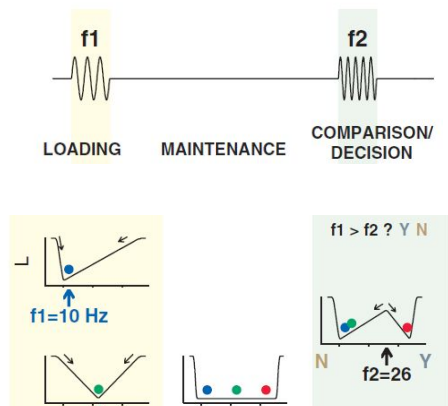


# The complementary roles of dimensionality and population structure in neural computations

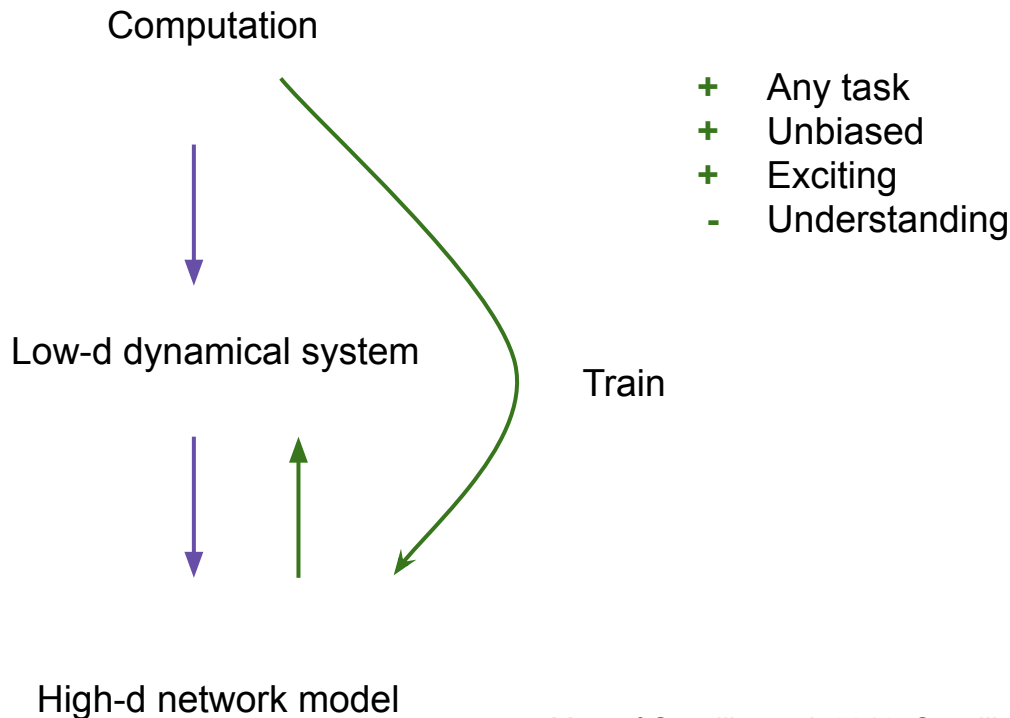
Flexible learning reading group  
Dubreuil et al. 2020 *bioRxiv* 2020.07.03.185942



# Modeling neural computations: classical vs novel approach



Machens et al. 2005



Mante&Sussillo et al. 2013; Sussillo et al. 2015;  
Sohn&Nairan et al. 2019; Yang et al. 2019

# Low rank RNNs

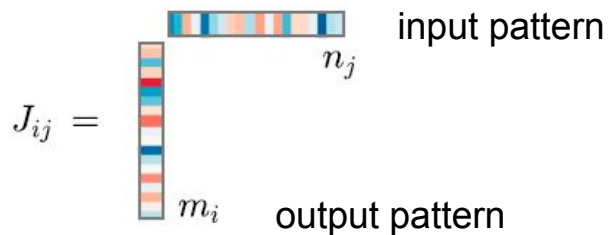
Recurrent

External

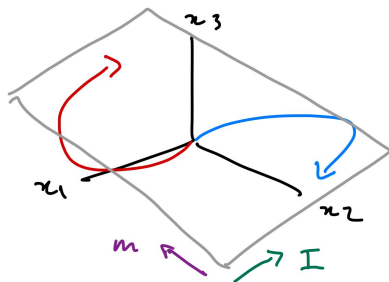
$$\tau \dot{x}(t) = -x(t) + \int \phi(x(t)) + I^{ff}(t)$$

$$z(t) = \omega^T \phi(x)$$

Readout

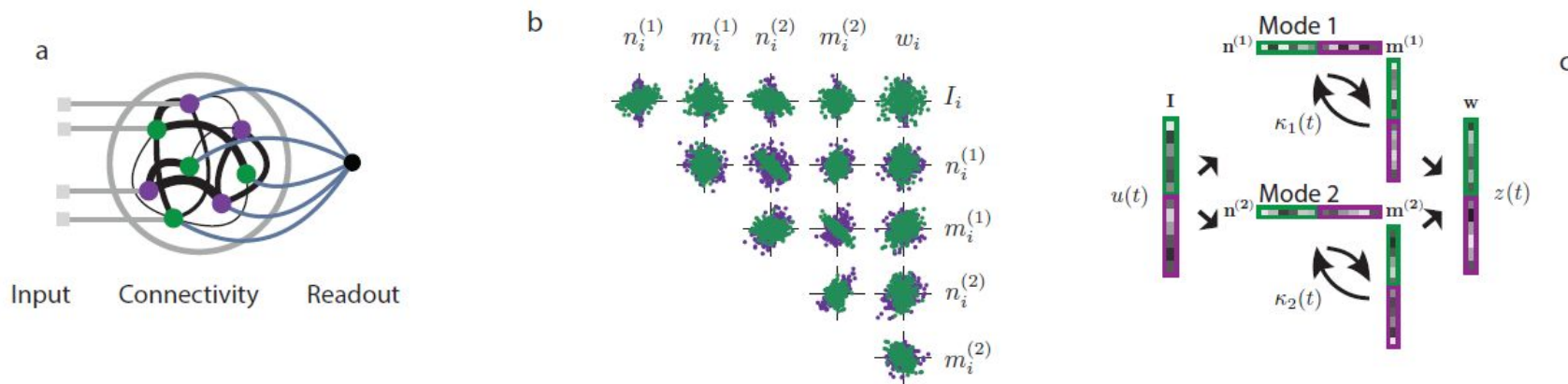


$$I^{ff}(t) = u(t) \cdot I$$



$$x(t) = k(t) m + v(t) I$$

This paper (Dubreuil et al. 2020): effect of “**populations**” in connectivity

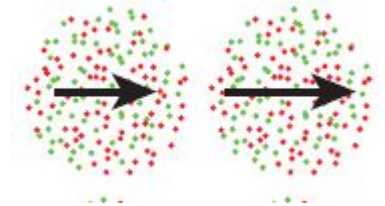
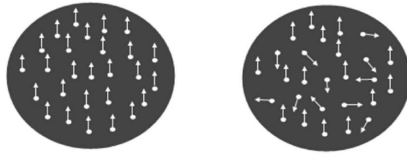


# #Populations: flexibility

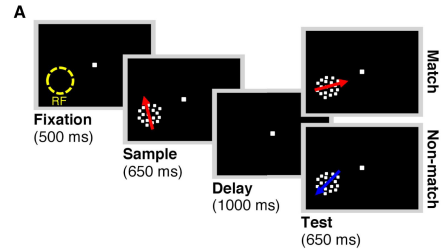
P = 1

P = 2

R = 1



R = 2



Rank: dimensionality

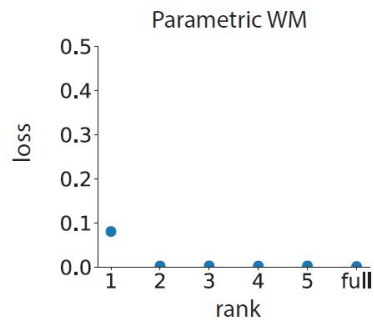
Min. Rank RNN



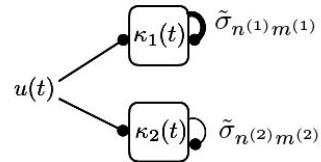
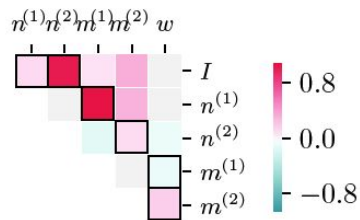
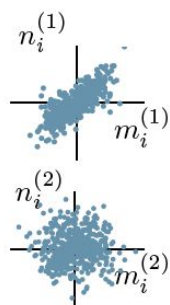
Covariance &  
population structure



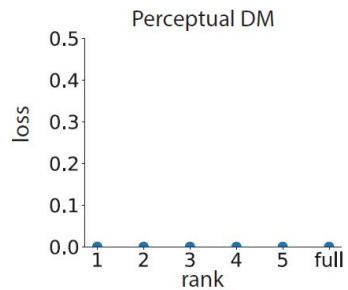
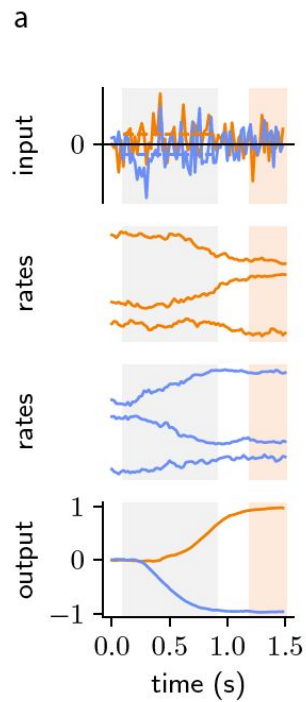
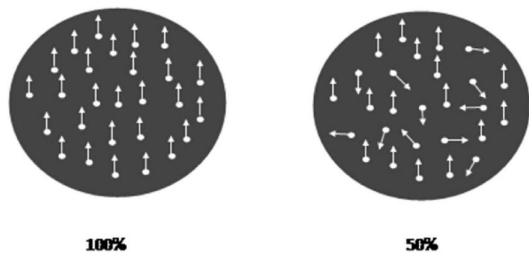
Reduced circuit



b



# Perceptual decision making with a rank 1 RNN

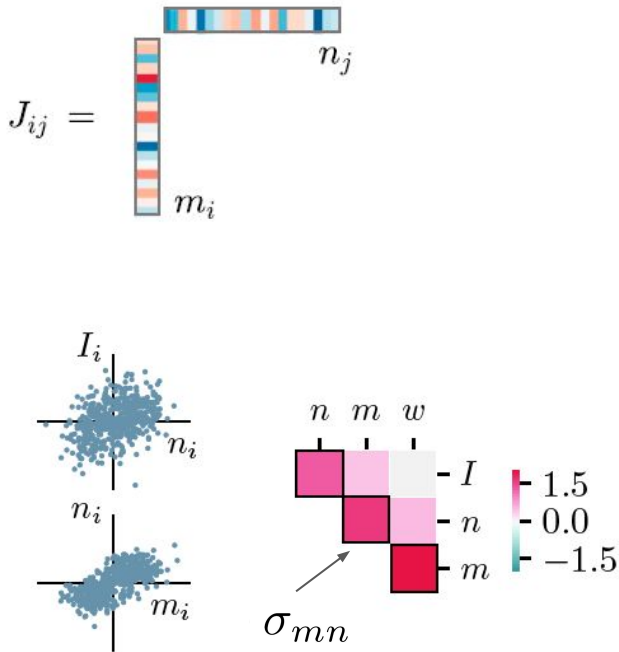


$$J_{ij} =$$

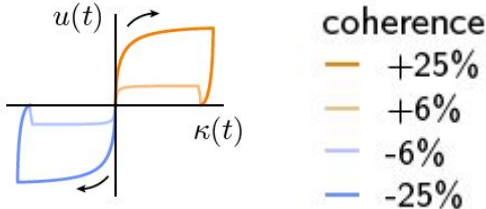
$n_j$

$m_i$

# Perceptual decision making: reduced circuit



e



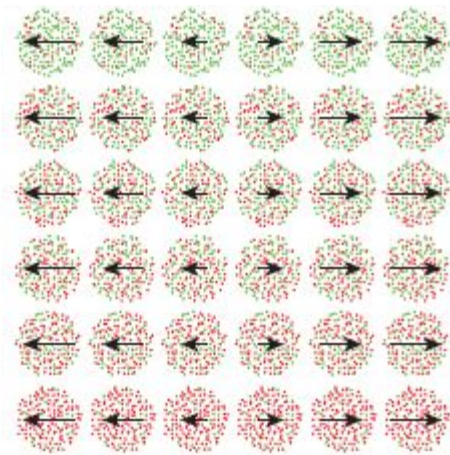
$$\frac{d\kappa}{dt} = -\kappa + \tilde{\sigma}_{mn}\kappa + \tilde{\sigma}_{nI}v(t)$$

$$\sigma_{nm}\langle\Phi'\rangle(\Delta) \quad \Delta = \sqrt{\sigma_m^2\kappa^2 + \sigma_I^2v^2}.$$

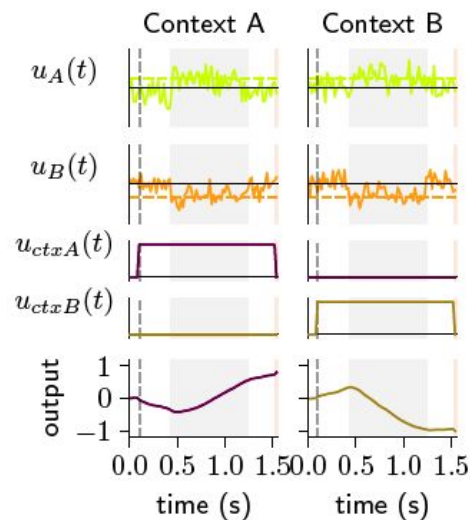
$$\langle\Phi'\rangle(\Delta) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{+\infty} dz \, e^{-z^2/2} \phi'(\Delta z)$$



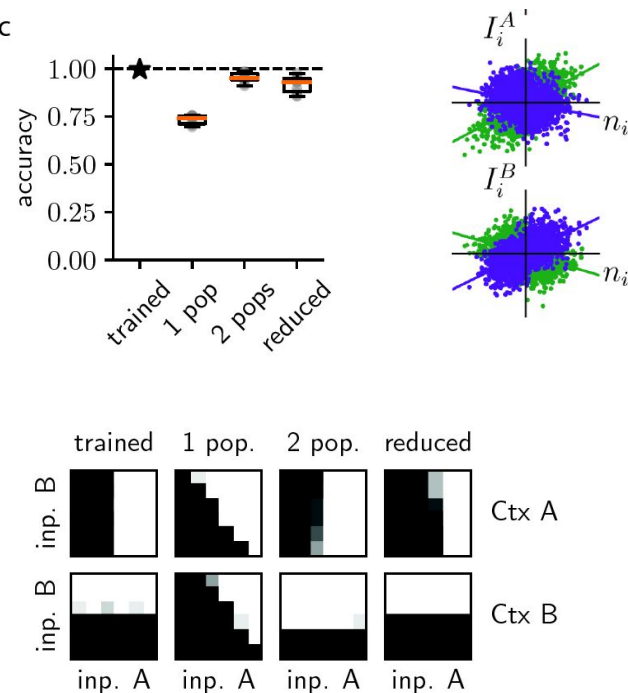
# Contextual decision making: need 2 populations



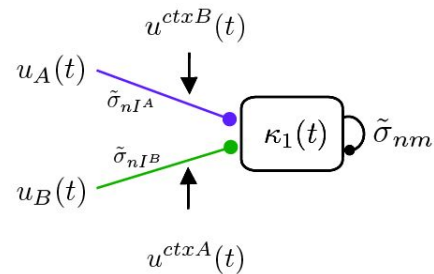
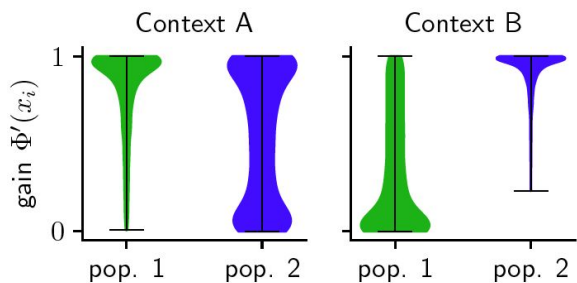
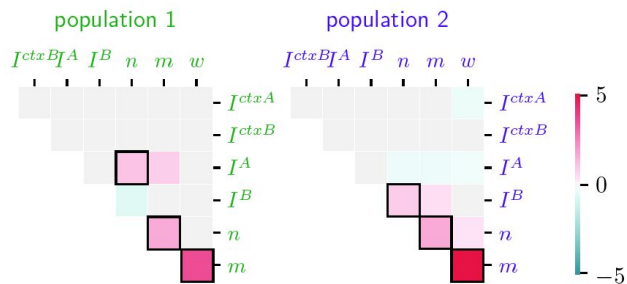
Mante&Sussillo et al. 2013



c



# Contextual decision making: reduced circuit



$$\frac{d\kappa}{dt} = -\kappa + \tilde{\sigma}_{mn}\kappa + \sigma_{nIA}^{(1)} \langle \Phi' \rangle_1 u_A(t) + \sigma_{nIB}^{(2)} \langle \Phi' \rangle_2 u_B(t)$$

$$\frac{d\kappa}{dt} = -\kappa + \tilde{\sigma}_{mn}\kappa + \sigma_{nIB}^{(2)} \langle \Phi' \rangle_2 u_B(t)$$

$$\frac{d\kappa}{dt} = -\kappa + \tilde{\sigma}_{mn}\kappa + \sigma_{nIA}^{(1)} \langle \Phi' \rangle_1 u_A(t)$$

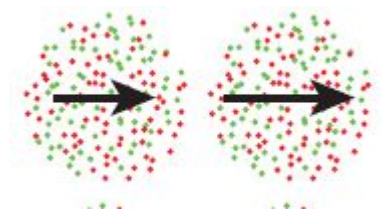
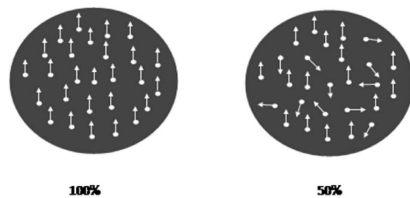
# #Populations: flexibility

Rank: dimensionality

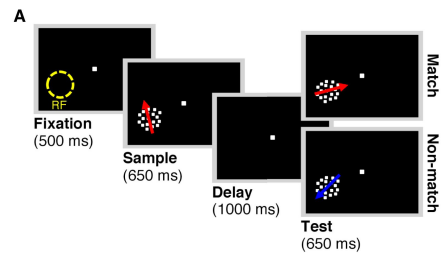
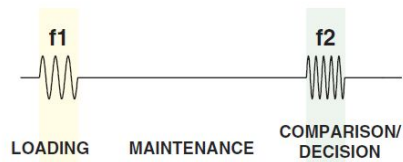
P = 1

P = 2

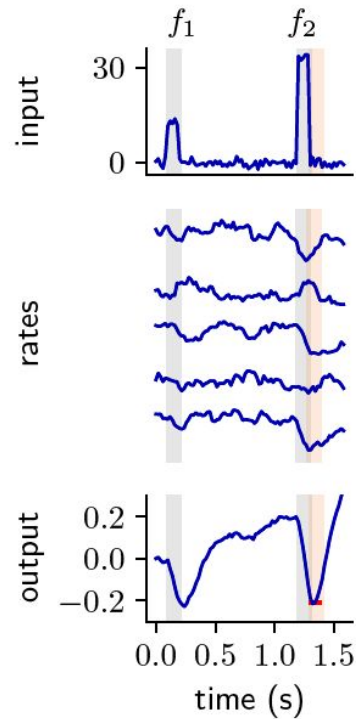
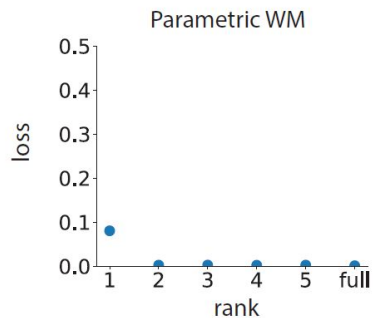
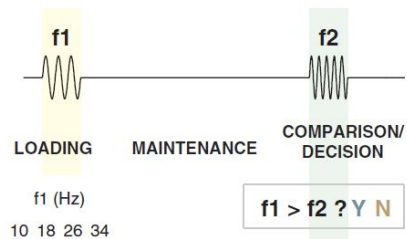
R = 1



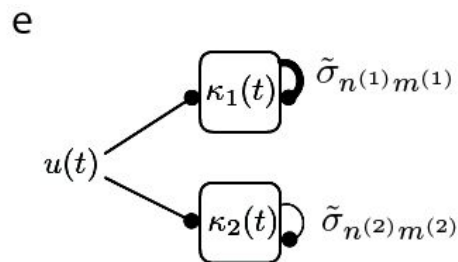
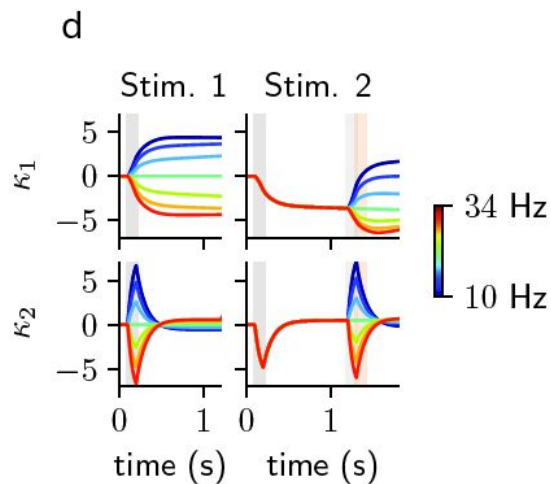
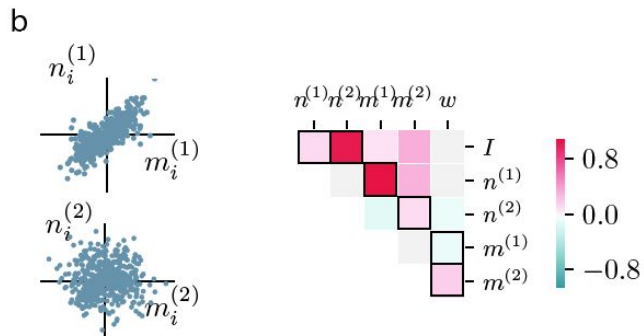
R = 2



# Parametric working memory (Rank = 2, #Populations = 1)



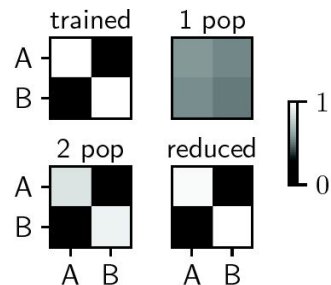
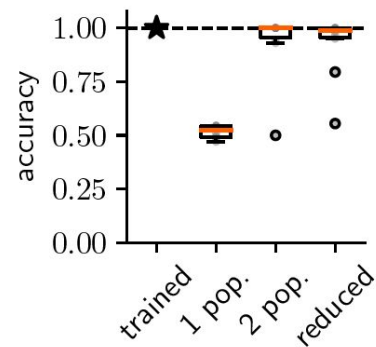
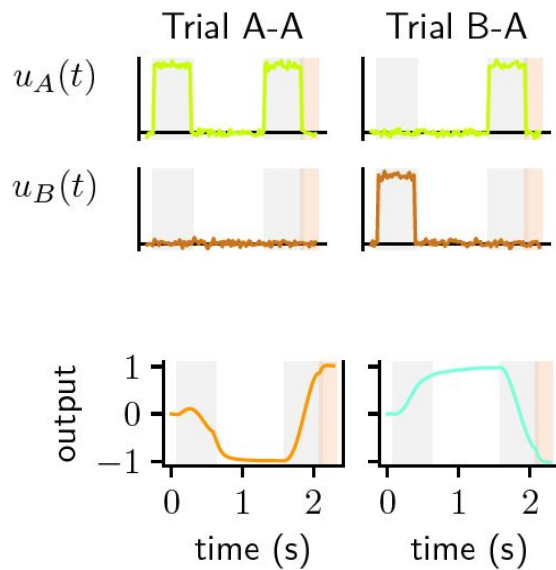
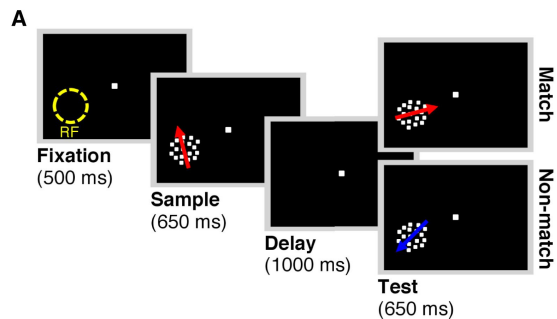
# Parametric working memory (Rank = 2, #Populations = 1)



$$\frac{d\kappa_1}{dt} = -\kappa_1 + \tilde{\sigma}_{n^{(1)}m^{(1)}}\kappa_1 + \tilde{\sigma}_{n^{(1)}}Iv(t)$$

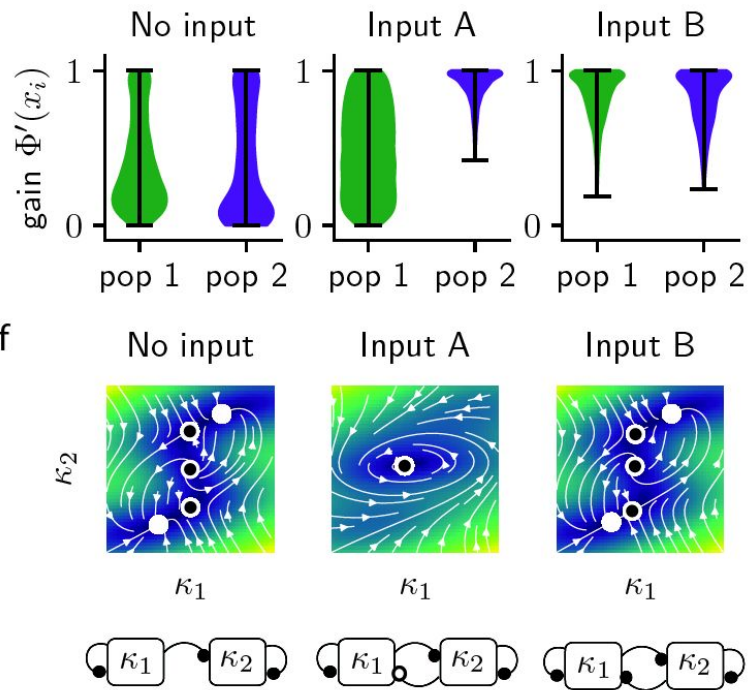
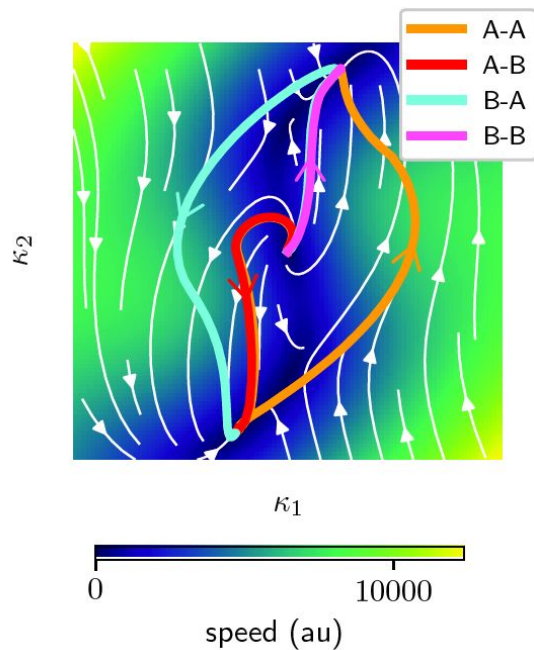
$$\frac{d\kappa_2}{dt} = -\kappa_2 + \tilde{\sigma}_{n^{(2)}m^{(2)}}\kappa_2 + \tilde{\sigma}_{n^{(2)}}Iv(t)$$

# Delayed match to sample



Freeman et al. 2001

# Gain modulation flexibly shapes network dynamics



# Gain modulation flexibly shapes network dynamics

Effective circuit:

$$\begin{aligned}\tau \frac{d\kappa_1}{dt} &= -\kappa_1 + \tilde{\sigma}_{n^{(1)}m^{(1)}} \kappa_1 + \tilde{\sigma}_{n^{(1)}m^{(2)}} \kappa_2 + \tilde{\sigma}_{n^{(1)}IA} u_A(t) + \tilde{\sigma}_{n^{(1)}IB} u_B(t) \\ \tau \frac{d\kappa_2}{dt} &= -\kappa_2 + \tilde{\sigma}_{n^{(2)}m^{(1)}} \kappa_1 + \tilde{\sigma}_{n^{(2)}m^{(2)}} \kappa_2 + \tilde{\sigma}_{n^{(2)}IA} u_A(t) + \tilde{\sigma}_{n^{(2)}IB} u_B(t).\end{aligned}$$

Effective couplings:

$$\begin{aligned}\tilde{\sigma}_{ab} &= \sum_{p=1}^P \alpha_p \langle \Phi' \rangle_p \sigma_{ab}^{(p)} . \\ \tilde{\sigma}_{n^{(2)}m^{(2)}} &= \frac{1}{2} \sigma_{n^{(2)}m^{(2)}}^{(1)} \langle \Phi' \rangle_1 + \frac{1}{2} \sigma_{n^{(2)}m^{(2)}}^{(2)} \langle \Phi' \rangle_2\end{aligned}$$



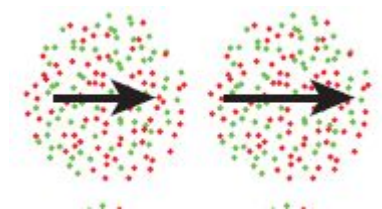
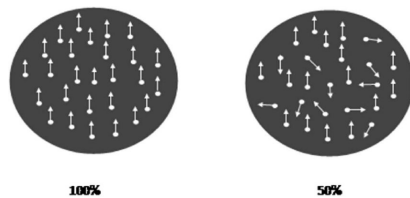
# #Populations: flexibility

Rank: dimensionality

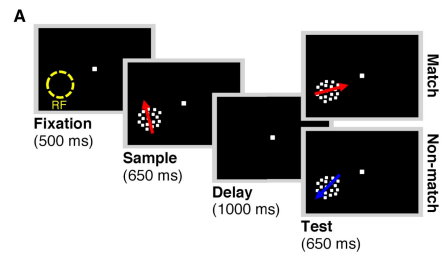
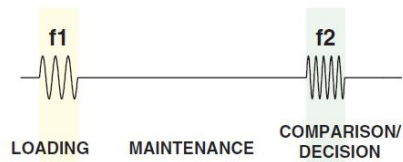
P = 1

P = 2

R = 1

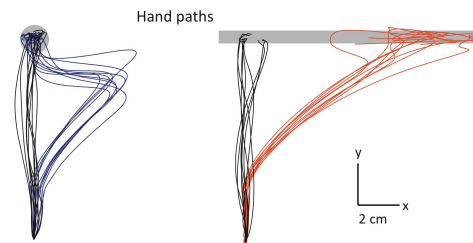
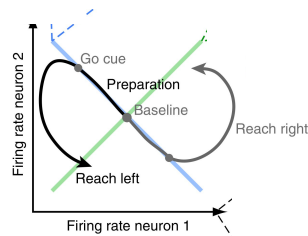
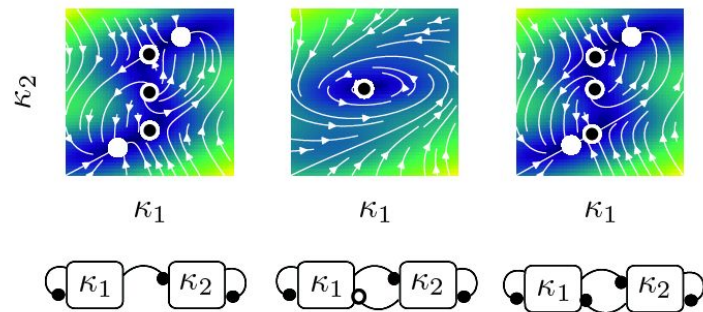


R = 2



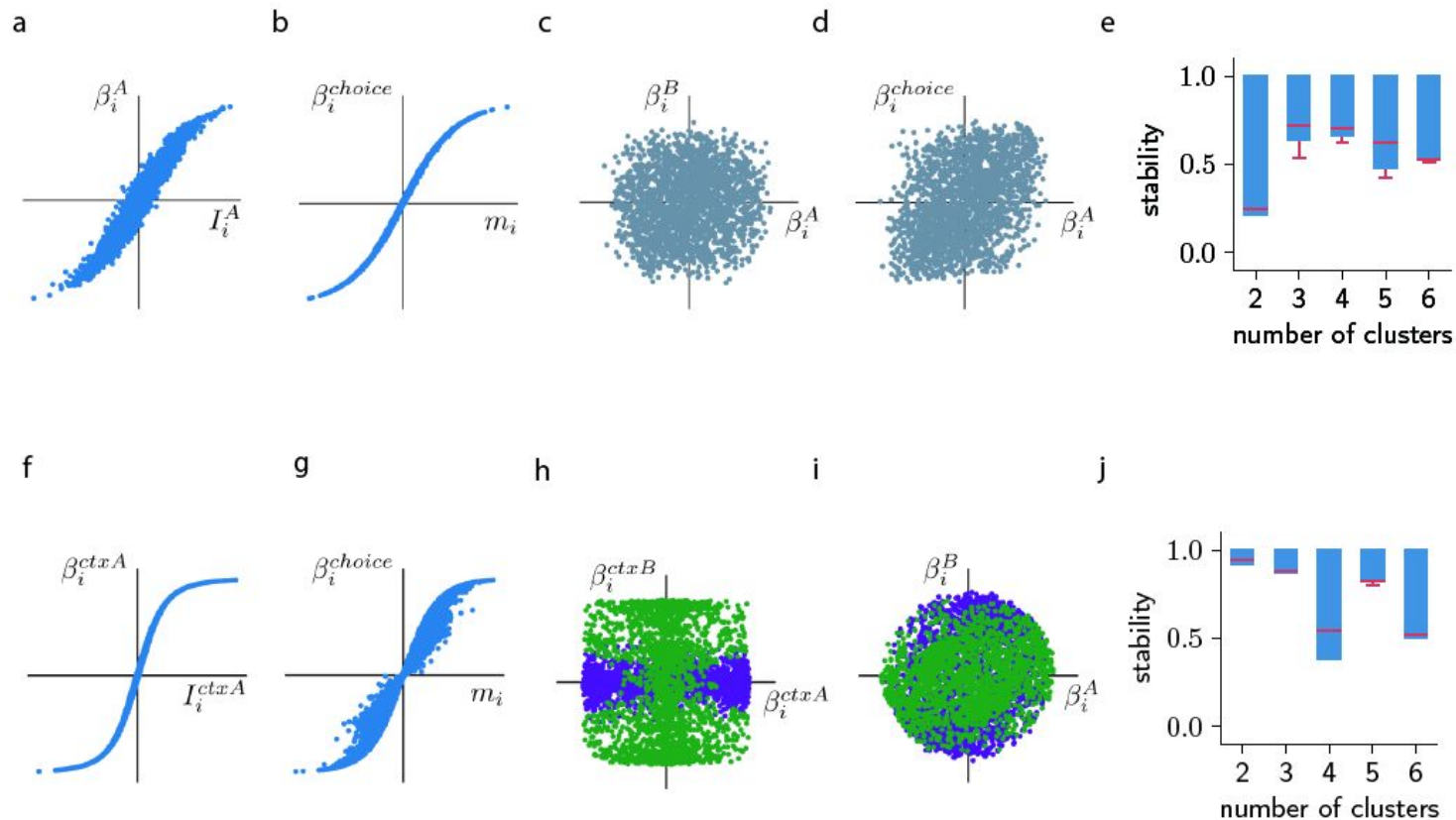
# Discussion

- Awesome: dynamics & circuit



Additional slides

# Connectivity determines selectivity



# Inactivating specific populations

