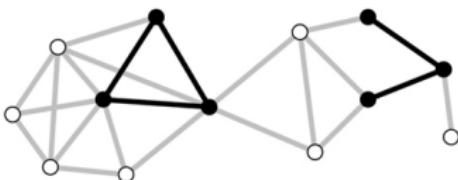


# Modelling the nonrandom and dynamic structure of local cortical circuits

Felix Z. Hoffmann

Slides: [bit.ly/xbcn18](http://bit.ly/xbcn18)



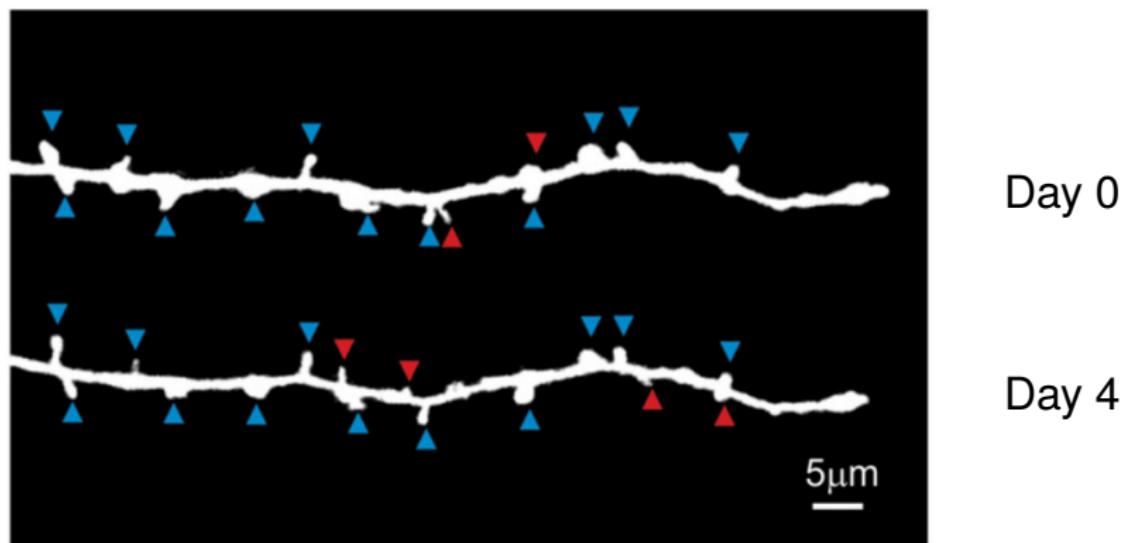
Bernstein Conference 2018 Satellite Workshop:  
*Adaptivity and Inhomogeneity in Neuronal Networks*



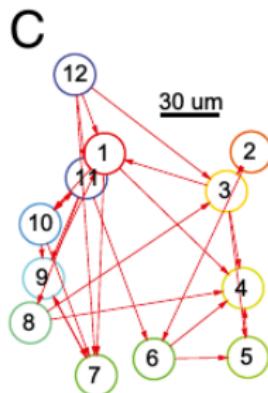
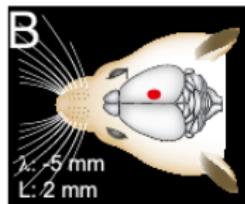
FIAS Frankfurt Institute  
for Advanced Studies



# The dynamic connectome

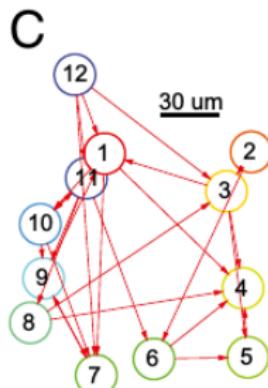
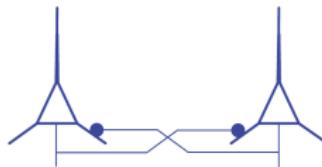
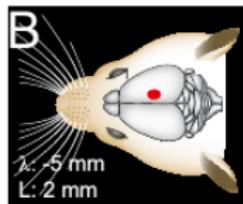


# Robust nonrandom connectivity patterns



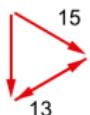
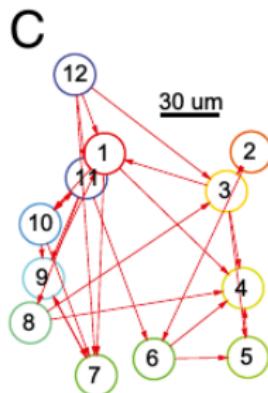
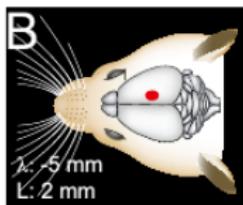
Perin et al. 2011; Song et al. 2005; Markram et al. 1997; Miner and Triesch 2016; Gal et al. 2017; Vegué et al. 2017

# Robust nonrandom connectivity patterns



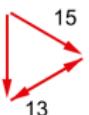
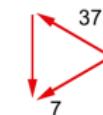
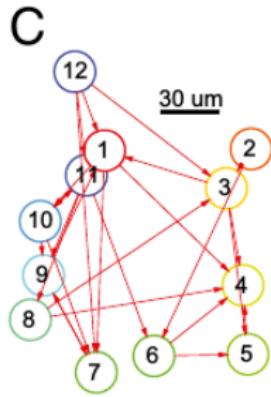
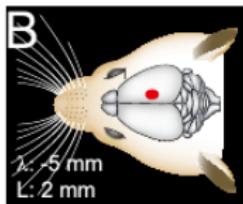
Perin et al. 2011; Song et al. 2005; Markram et al. 1997; Miner and Triesch 2016; Gal et al. 2017; Vegué et al. 2017

# Robust nonrandom connectivity patterns



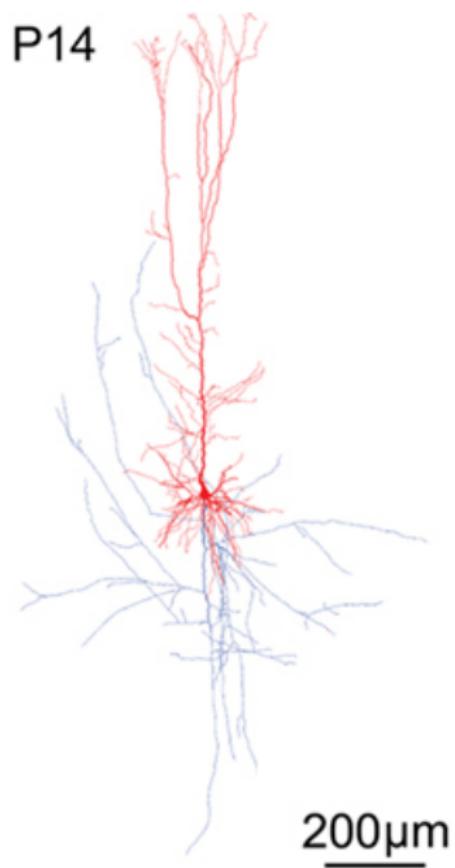
Perin et al. 2011; Song et al. 2005; Markram et al. 1997; Miner and Triesch 2016; Gal et al. 2017; Vegué et al. 2017

# Robust nonrandom connectivity patterns

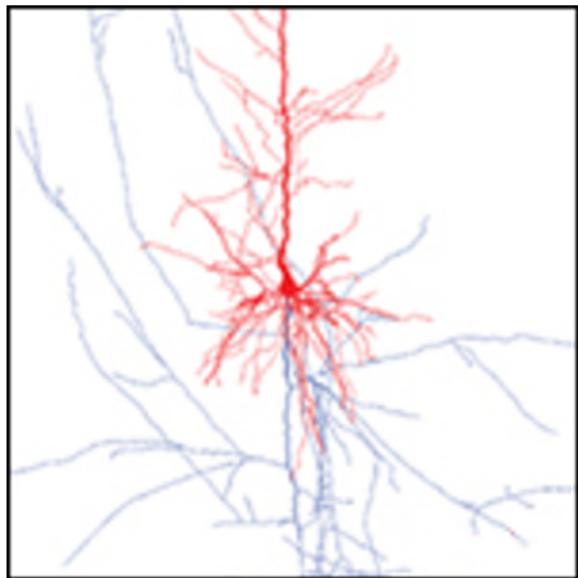


Perin et al. 2011; Song et al. 2005; Markram et al. 1997; Miner and Triesch 2016; Gal et al. 2017; Vegué et al. 2017

# Anisotropy in stereotypical axon morphology

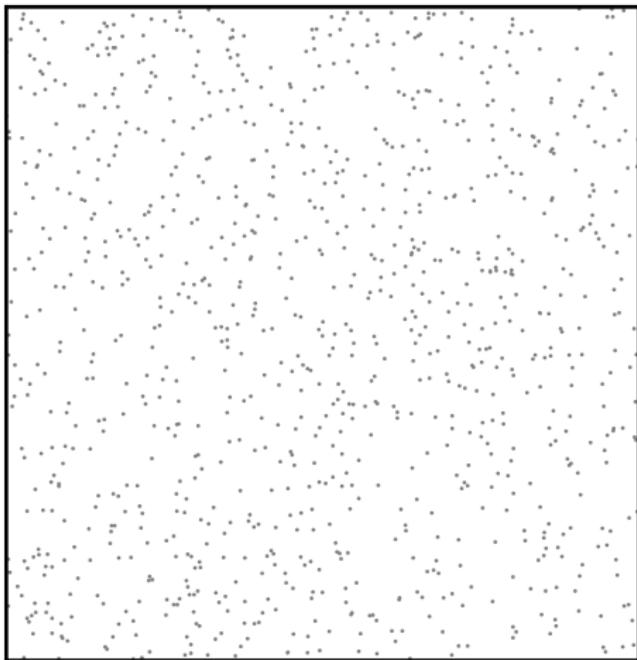


# Anisotropy in stereotypical axon morphology



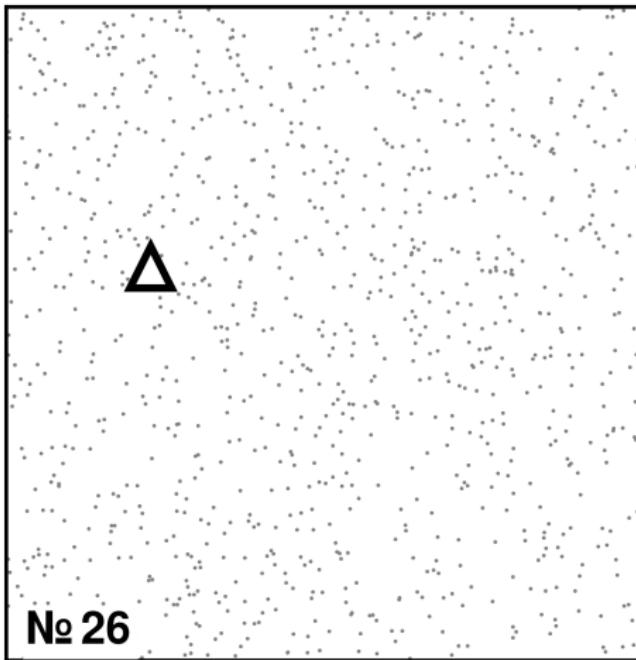
$200\mu\text{m}$

# Anisotropic network model



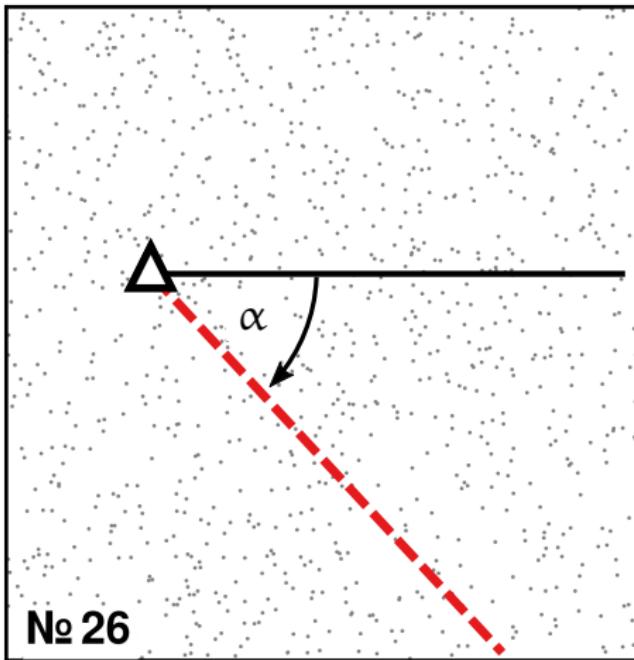
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# Anisotropic network model



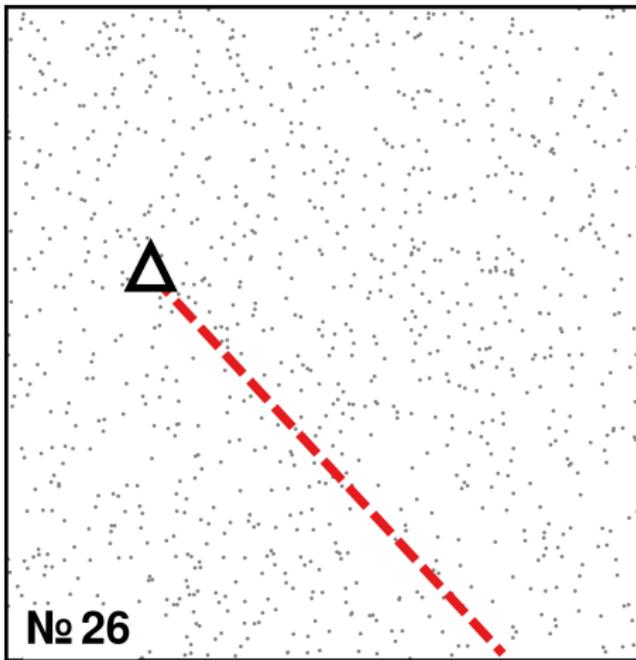
$N = 1000$

# Anisotropic network model



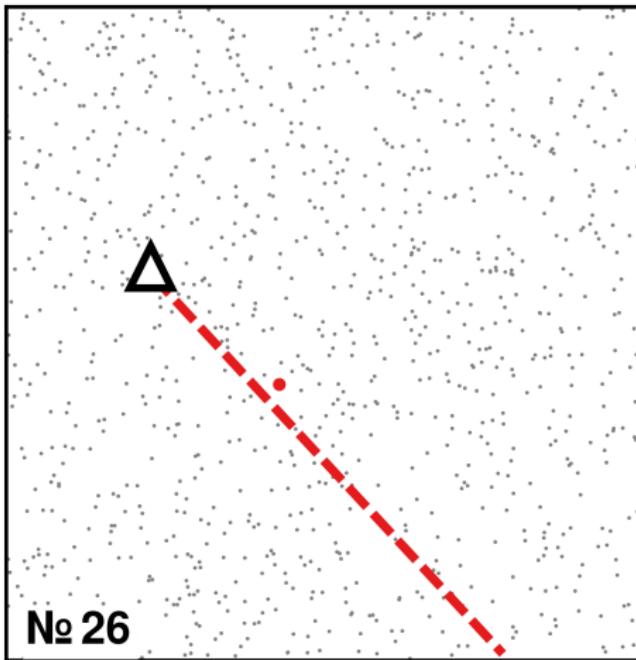
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# Anisotropic network model



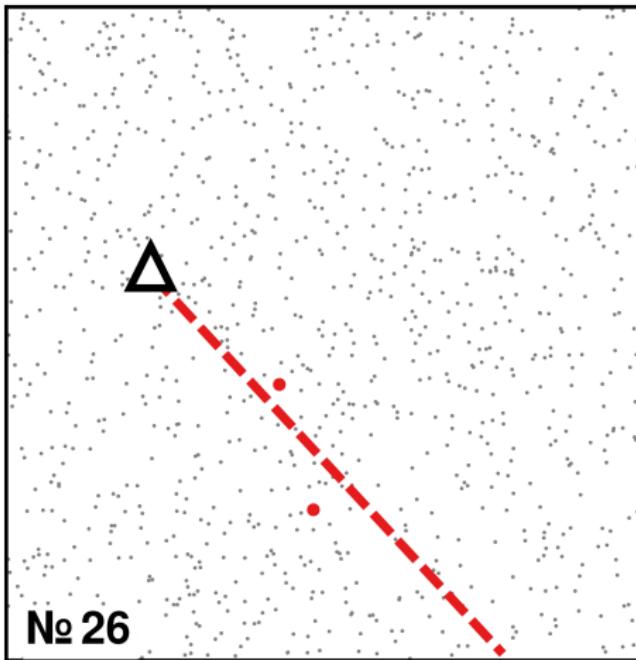
$N = 1000$

# Anisotropic network model



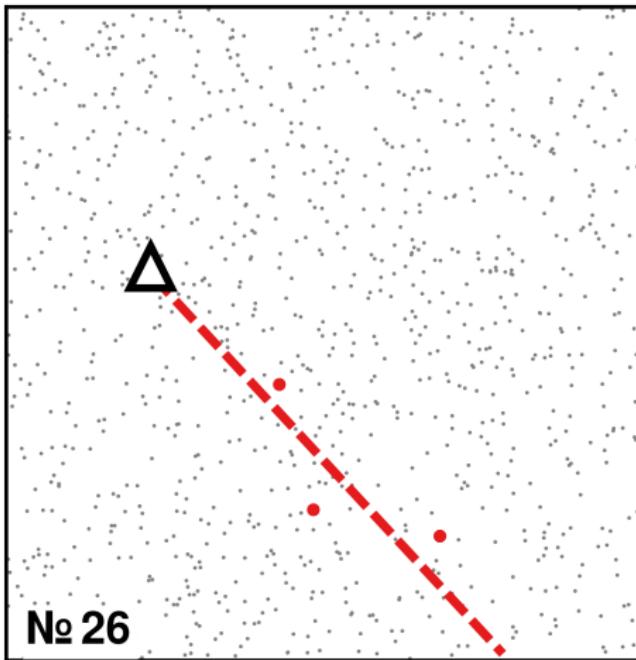
$N = 1000$

# Anisotropic network model



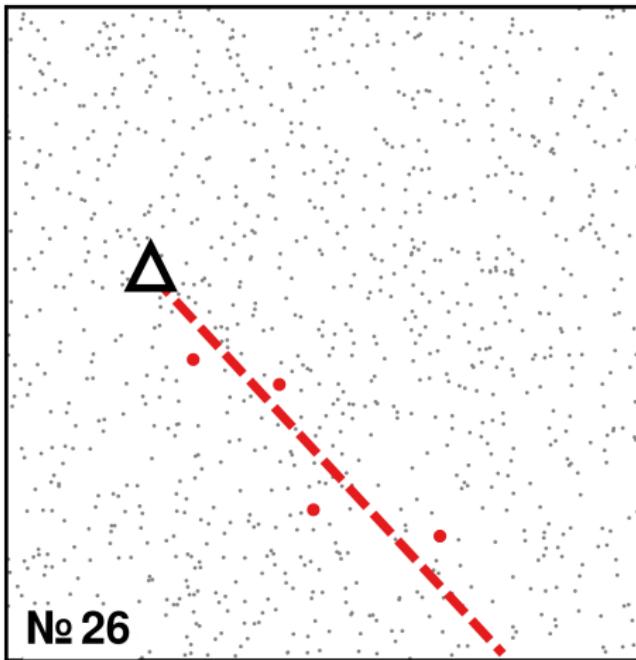
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# Anisotropic network model



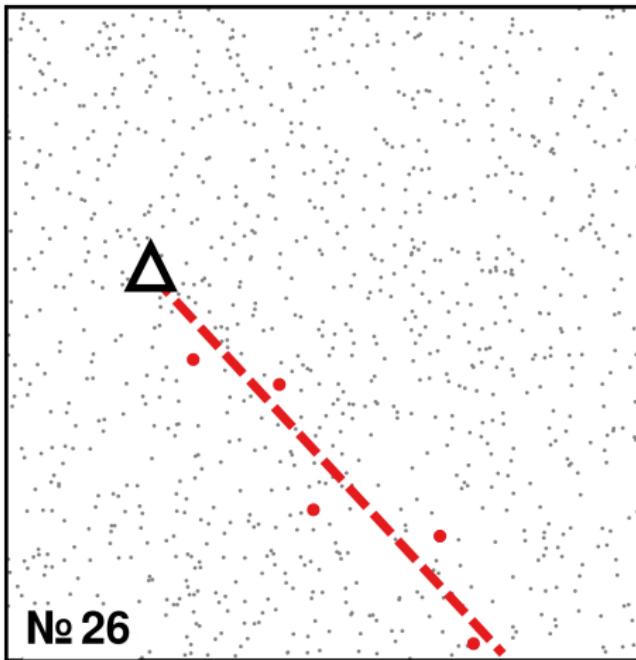
$N = 1000$

# Anisotropic network model



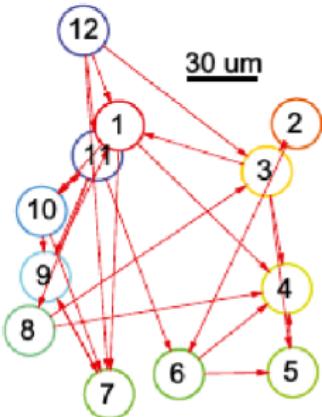
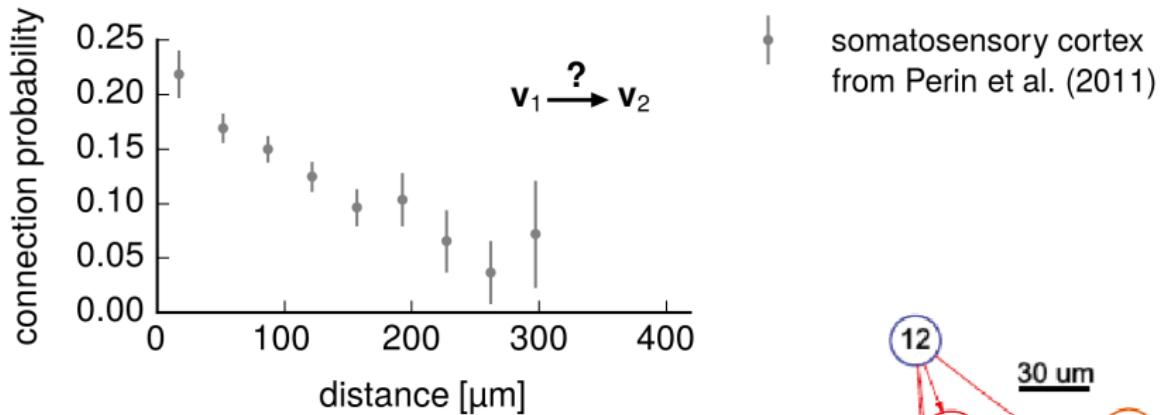
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# Anisotropic network model

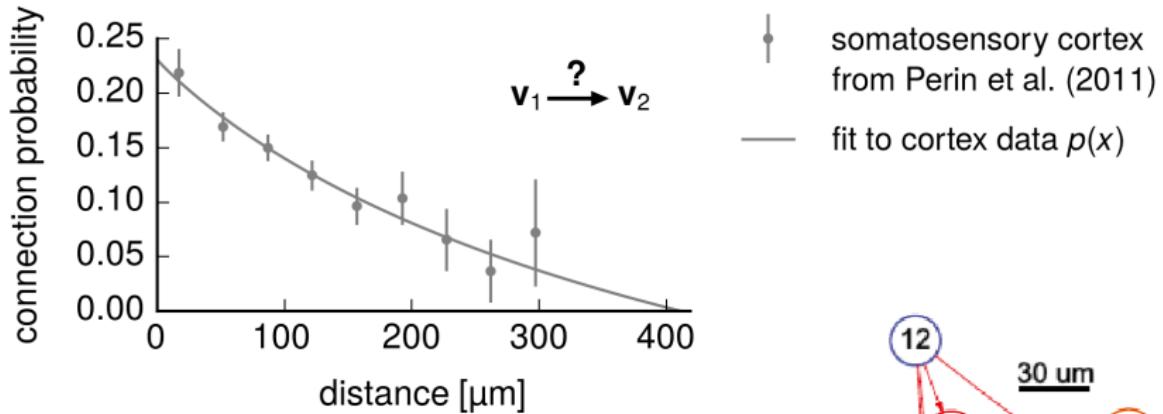


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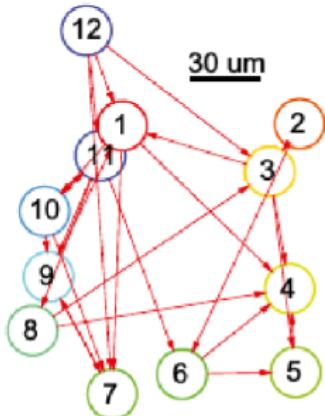
# Anisotropic network model – Distance-dependency



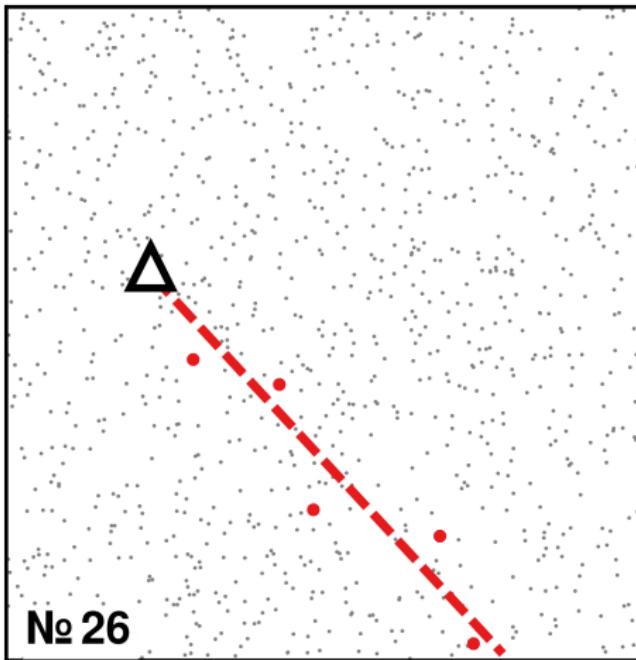
# Anisotropic network model – Distance-dependency



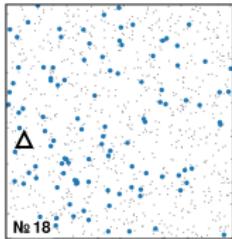
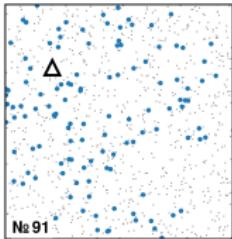
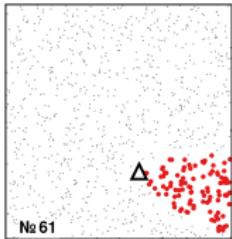
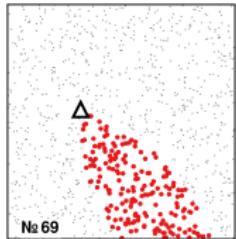
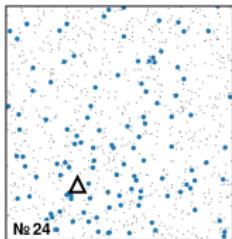
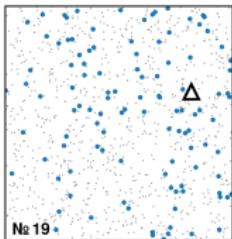
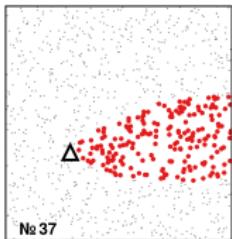
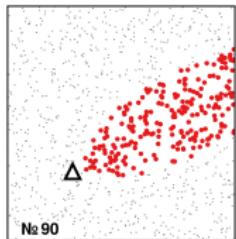
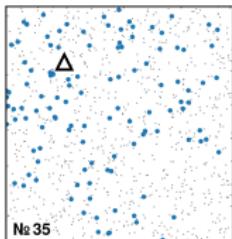
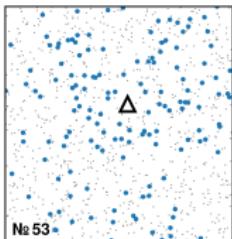
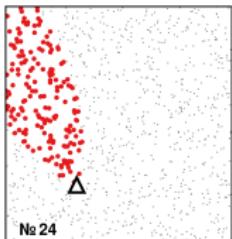
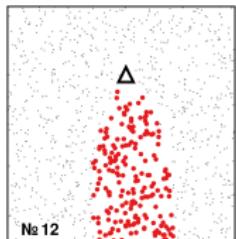
● somatosensory cortex  
from Perin et al. (2011)  
— fit to cortex data  $p(x)$



# Anisotropic network model



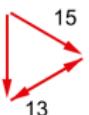
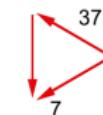
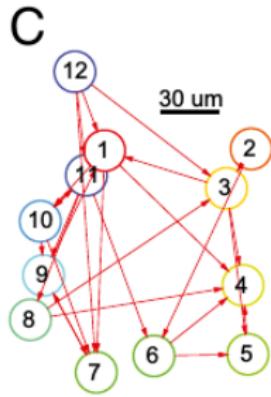
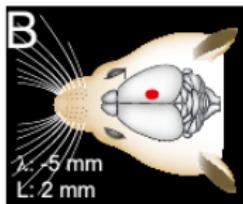
# Anisotropic network model



targets

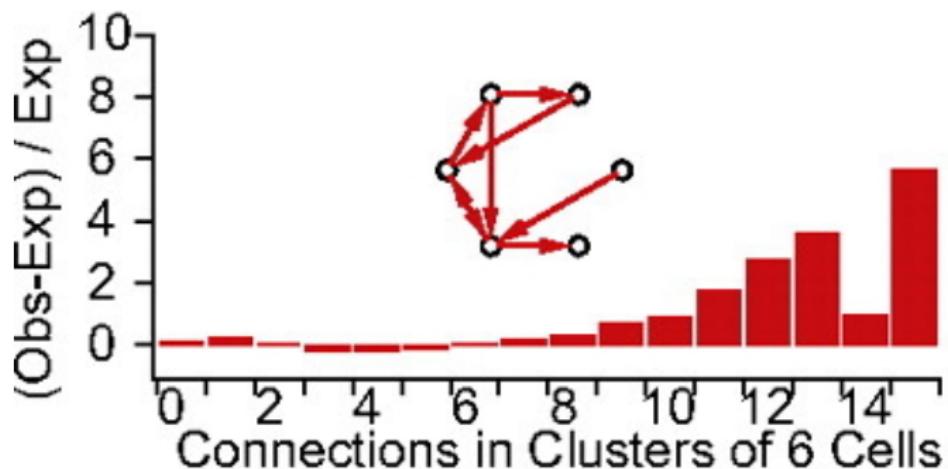
inputs

# Robust nonrandom connectivity patterns

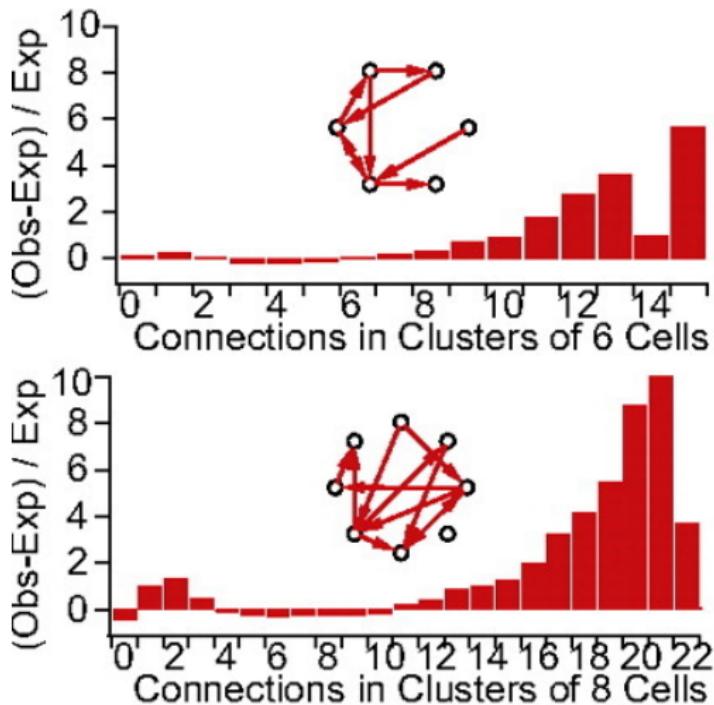


Perin et al. 2011; Song et al. 2005; Markram et al. 1997; Miner and Triesch 2016; Gal et al. 2017; Vegué et al. 2017

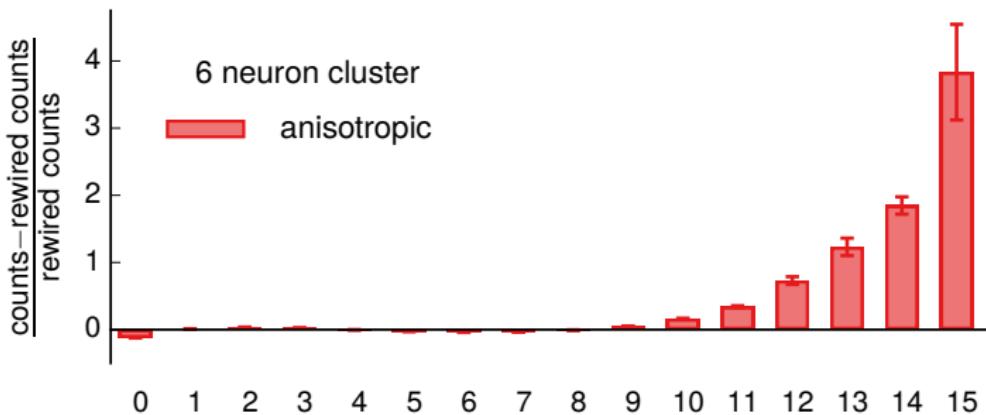
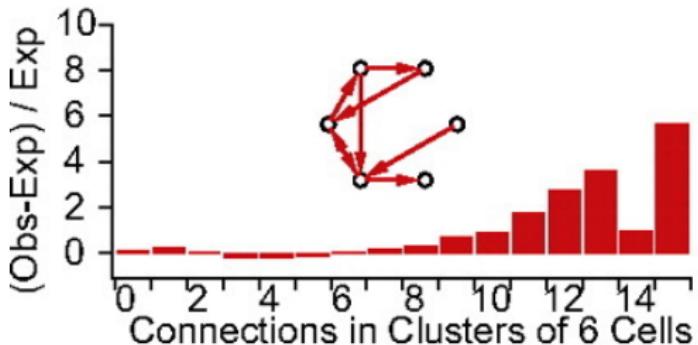
## Connection counts in neuron clusters



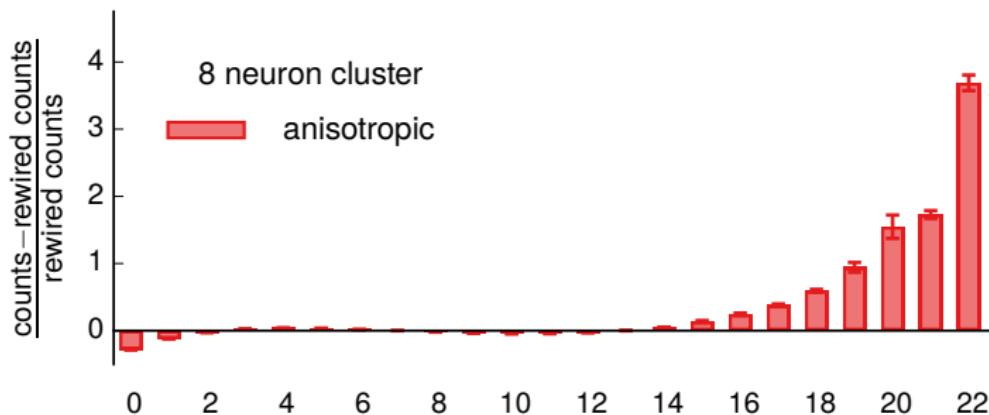
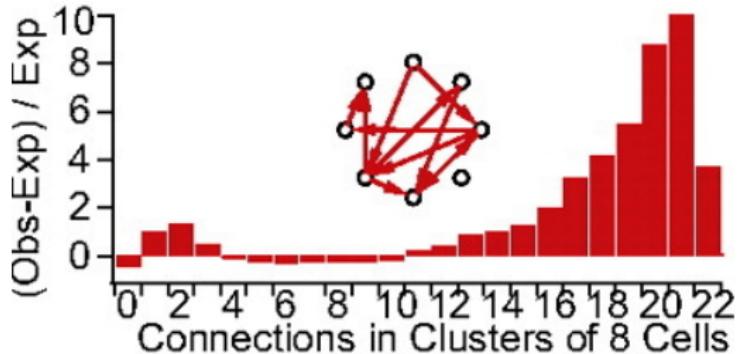
# Connection counts in neuron clusters



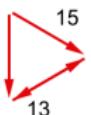
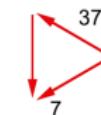
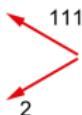
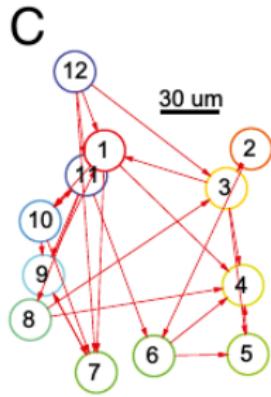
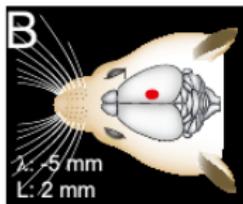
# Connection counts in neuron clusters



# Connection counts in neuron clusters



# Robust nonrandom connectivity patterns



Perin et al. 2011; Song et al. 2005; Markram et al. 1997; Miner and Triesch 2016; Gal et al. 2017; Vegué et al. 2017

# Overrepresentation of reciprocal connections

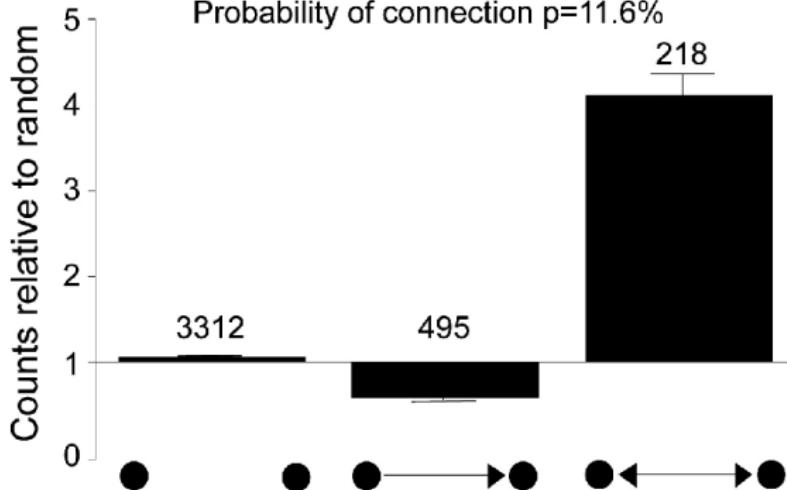
**A**

Null hypothesis assumes independent connection probabilities

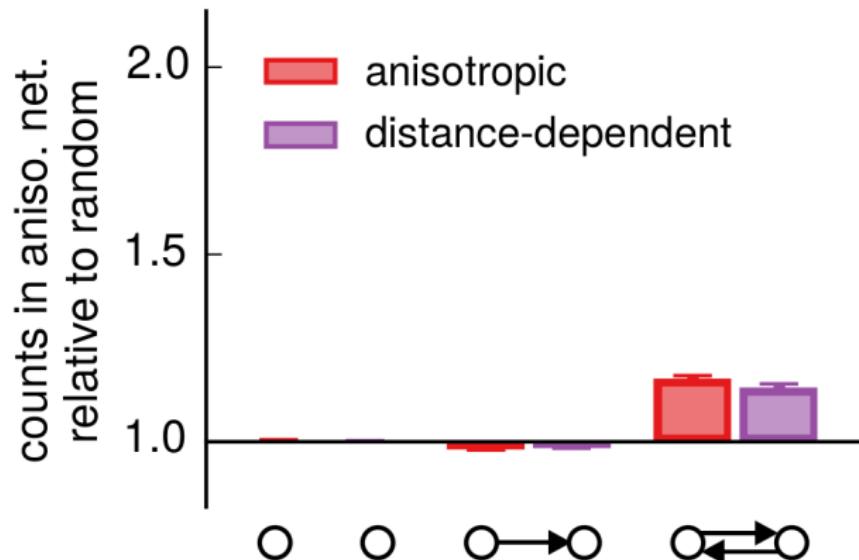


**B**

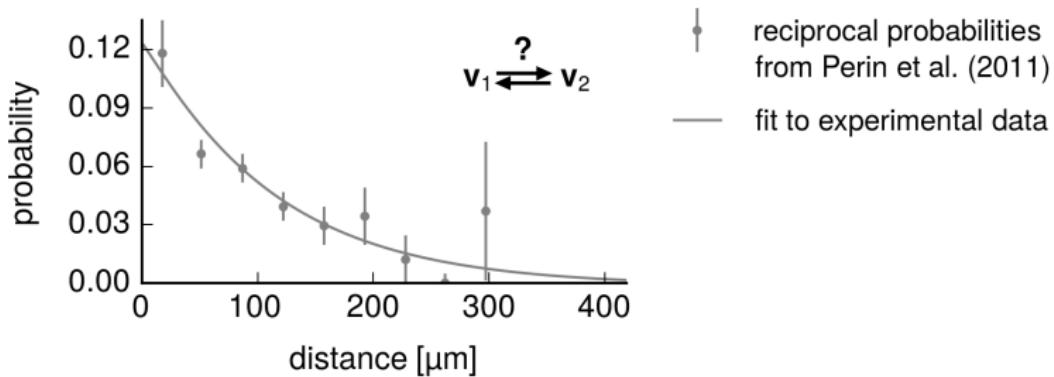
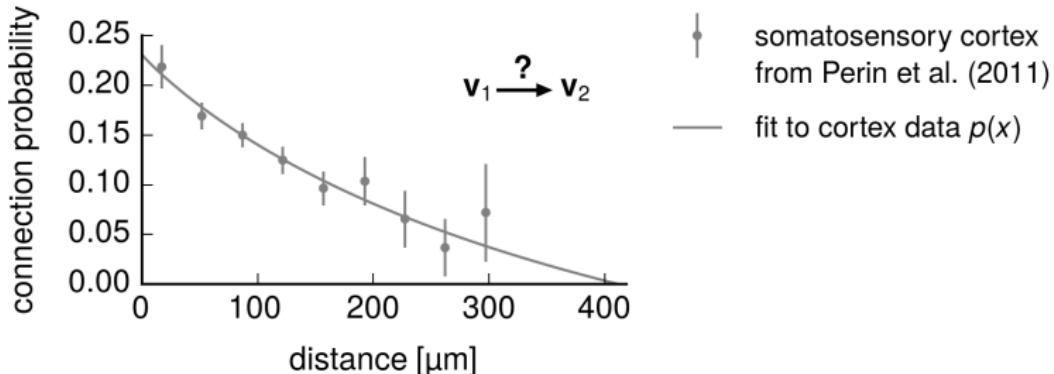
Probability of connection  $p=11.6\%$



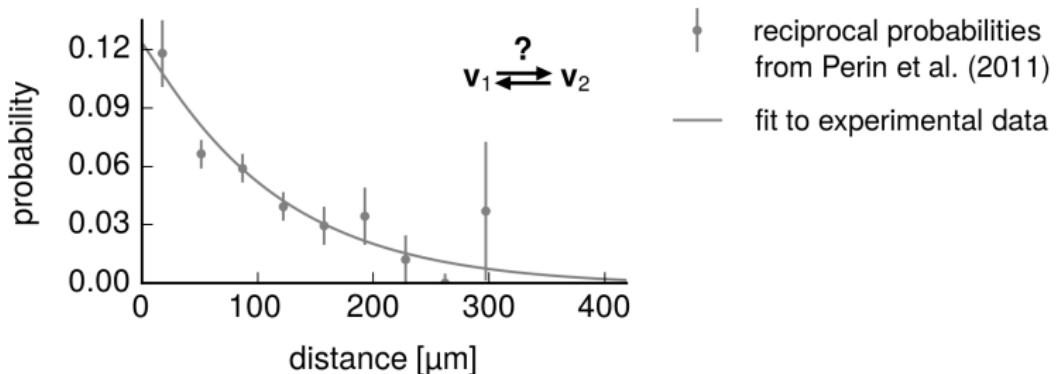
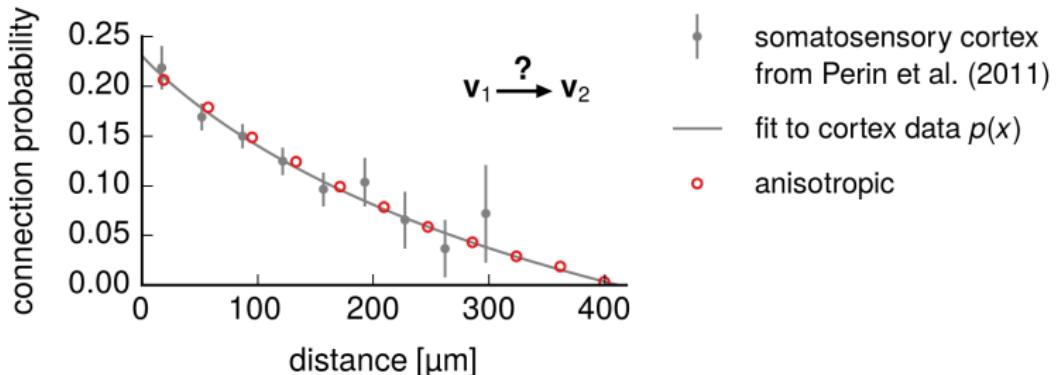
# Overrepresentation of reciprocal connections



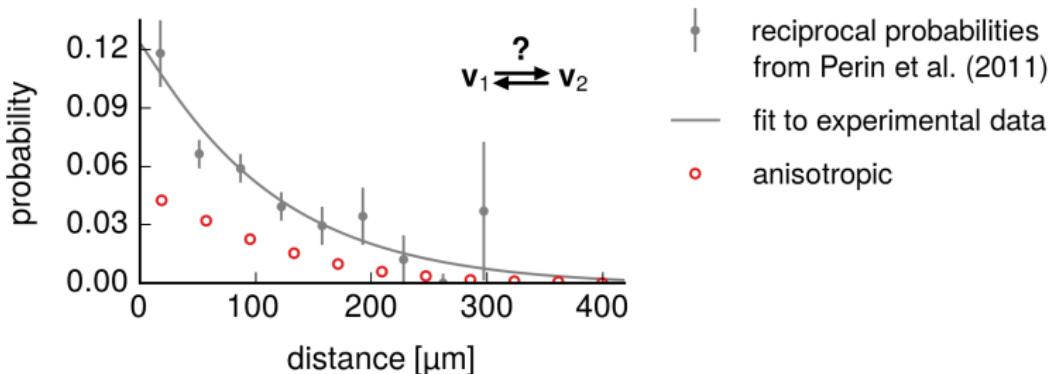
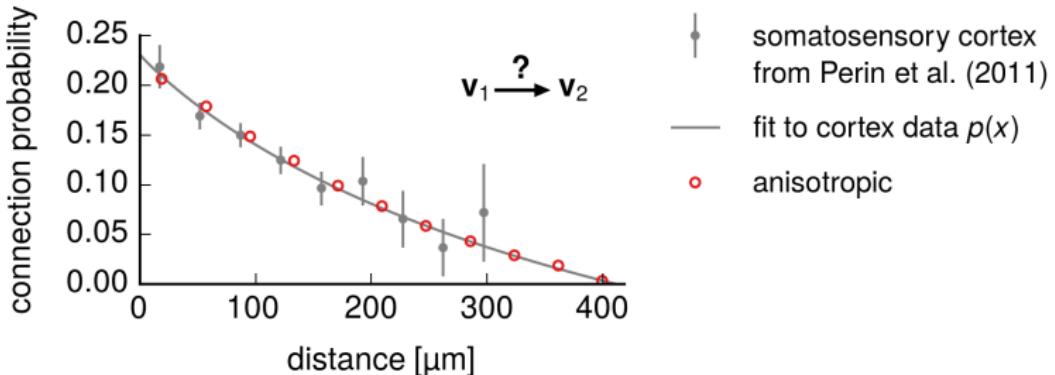
# Reciprocity – Distance-dependent?



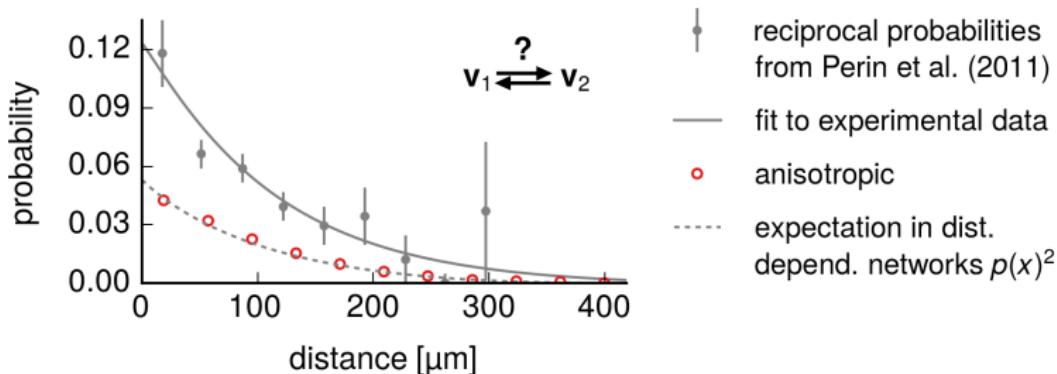
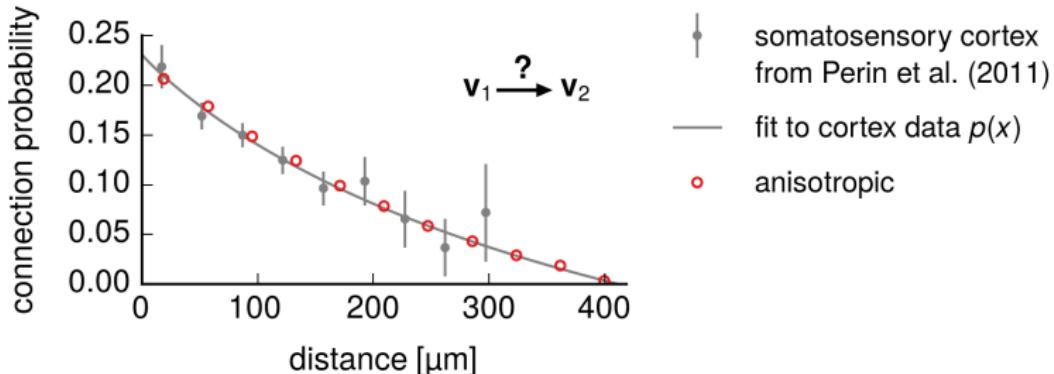
# Reciprocity – Distance-dependent?



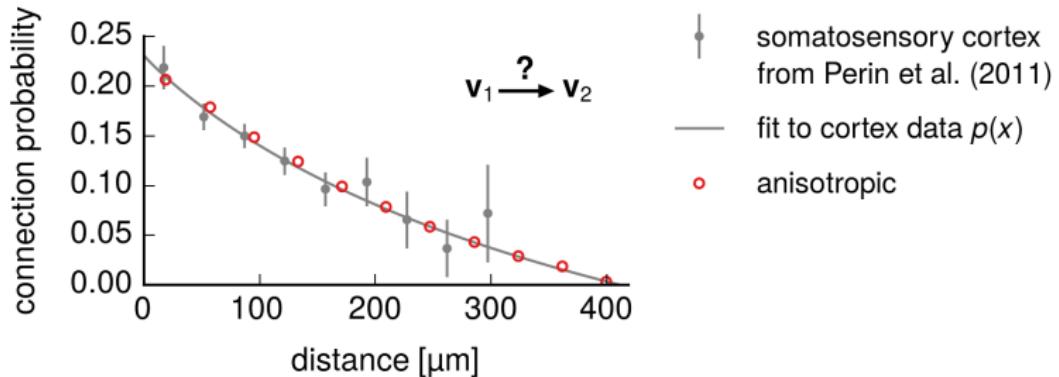
# Reciprocity – Distance-dependent?



# Reciprocity – Distance-dependent?

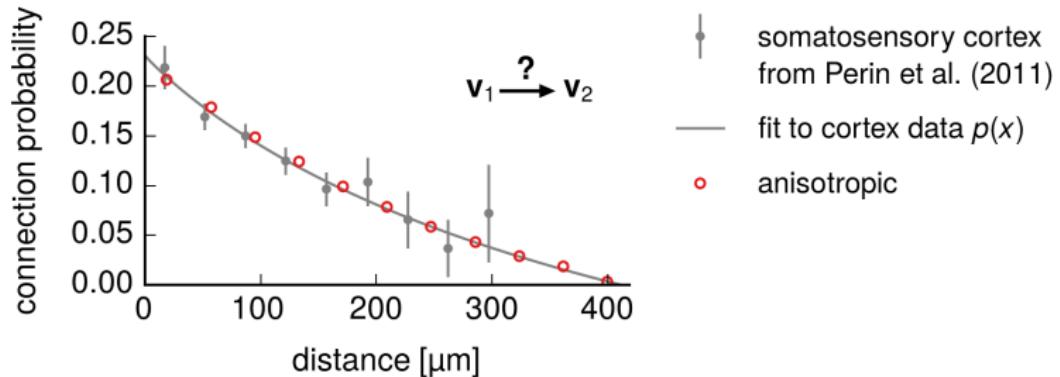


# Reciprocity – Distance-dependent?



Other sources for the overrepresentation of bidirectional connections?

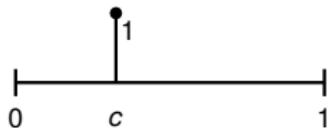
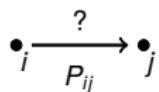
# Reciprocity – Distance-dependent?



Other sources for the overrepresentation of bidirectional connections?

⇒ Hoffmann, FZ and Triesch, J (2017). Nonrandom Network Connectivity Comes in Pairs. *Network Neuroscience*

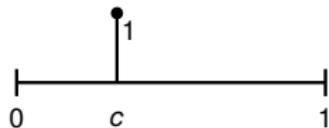
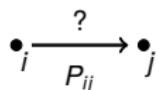
## Standard random network model



Probability of connection a constant  $P_{ij}$ ,

$$P_{ij} = c$$

## Standard random network model



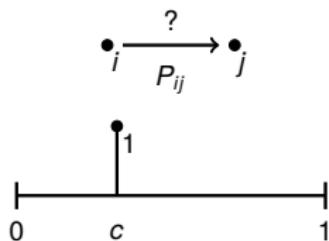
Probability of connection a constant  $P_{ij}$ ,

$$P_{ij} = c$$

**Overall connection probability**

$$\mu = P_{ij} = c$$

## Standard random network model



Probability of connection a constant  $P_{ij}$ ,

$$P_{ij} = c$$

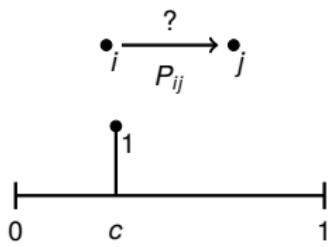
**Overall connection probability**

$$\mu = P_{ij} = c$$

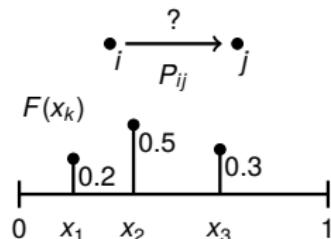
**Bidirectional connection**

$$P_{\text{bidir}} = P_{ij}P_{ji} = c^2$$

## Standard random network model



## Varying connection probabilities



Probability of connection a constant  $P_{ij}$ ,

$$P_{ij} = c$$

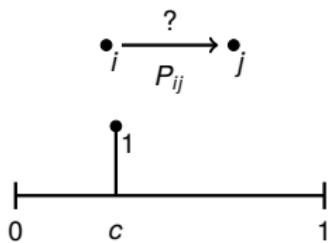
**Overall connection probability**

$$\mu = P_{ij} = c$$

**Bidirectional connection**

$$P_{\text{bidir}} = P_{ij}P_{ji} = c^2$$

## Standard random network model



Probability of connection a constant  $P_{ij}$ ,

$$P_{ij} = c$$

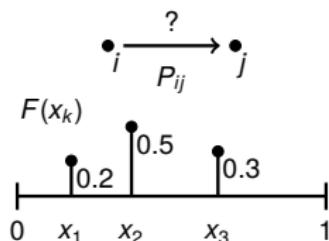
## Overall connection probability

$$\mu = P_{ij} = c$$

## Bidirectional connection

$$P_{\text{bidir}} = P_{ij}P_{ji} = c^2$$

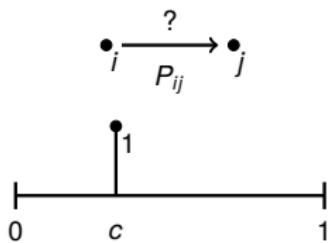
## Varying connection probabilities



Probability of connection a random variable  $P_{ij}$ ,

$$\mathbf{Prob}(P_{ij} = x_k) = F(x_k)$$

## Standard random network model



Probability of connection a constant  $P_{ij}$ ,

$$P_{ij} = c$$

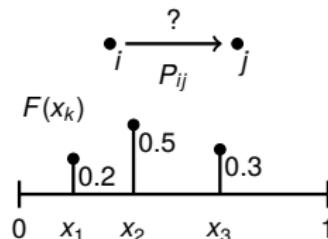
**Overall connection probability**

$$\mu = P_{ij} = c$$

**Bidirectional connection**

$$P_{\text{bidir}} = P_{ij}P_{ji} = c^2$$

## Varying connection probabilities



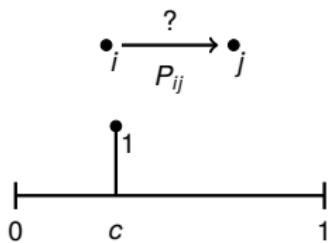
Probability of connection a random variable  $P_{ij}$ ,

$$\mathbf{Prob}(P_{ij} = x_k) = F(x_k)$$

**Overall connection probability**

$$\mu = \sum_{k=1}^m F(x_k)x_k$$

## Standard random network model



Probability of connection a constant  $P_{ij}$ ,

$$P_{ij} = c$$

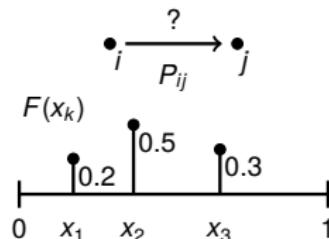
**Overall connection probability**

$$\mu = P_{ij} = c$$

**Bidirectional connection**

$$P_{\text{bidir}} = P_{ij}P_{ji} = c^2$$

## Varying connection probabilities



Probability of connection a random variable  $P_{ij}$ ,

$$\text{Prob}(P_{ij} = x_k) = F(x_k)$$

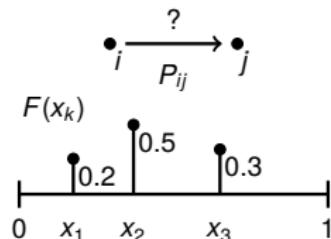
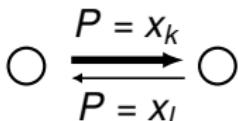
**Overall connection probability**

$$\mu = \sum_{k=1}^m F(x_k)x_k$$

**Bidirectional connection**

$$P_{\text{bidir}} = ?$$

## Varying connection probabilities



Probability of connection a constant  $P_{ij}$ ,

$$P_{ij} = c$$

**Overall connection probability**

$$\mu = P_{ij} = c$$

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$$P_{\text{bidir}} = P_{ij}P_{ji} = c^2$$

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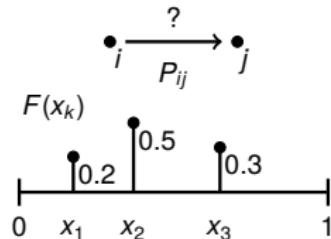
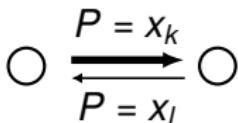
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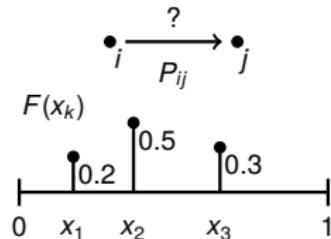
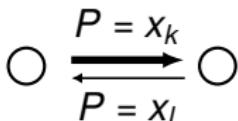
**Overall connection probability**

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**Bidirectional connection**

$$P_{\text{bidir}} = \sum_{k=1}^m \sum_{l=1}^m F(x_k)x_k F(x_l|x_k)x_l$$

## Varying connection probabilities



Probability of connection a constant  $P_{ij}$ ,

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Probability of connection a random variable  $P_{ij}$ ,

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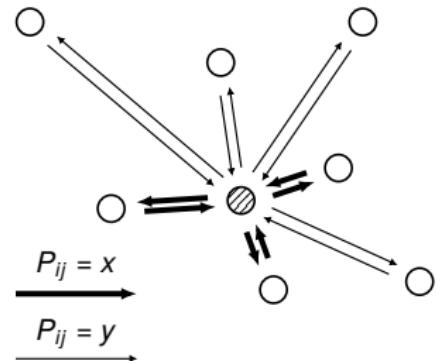
**Overall connection probability**

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**Bidirectional connection**

$$P_{\text{bidir}} = \sum_{k=1}^m \sum_{l=1}^m F(x_k)x_k F(x_l|x_k)x_l$$

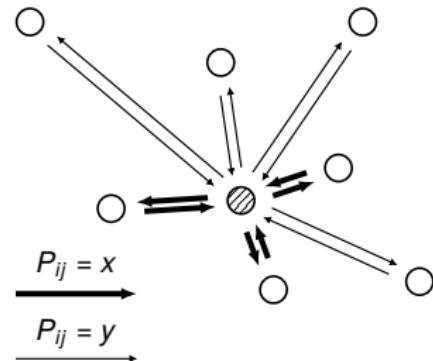
$$F(x_l|x_k) = \begin{cases} 1 & \text{if } l = k \\ 0 & \text{otherwise.} \end{cases}$$



$$P_{\text{bidir}} = \sum_{k=1}^m \sum_{l=1}^m F(x_k) x_k F(x_l | x_k) x_l$$

$$= \sum_{k=1}^m F(x_k) x_k^2.$$

$$F(x_l | x_k) = \begin{cases} 1 & \text{if } l = k \\ 0 & \text{otherwise.} \end{cases}$$

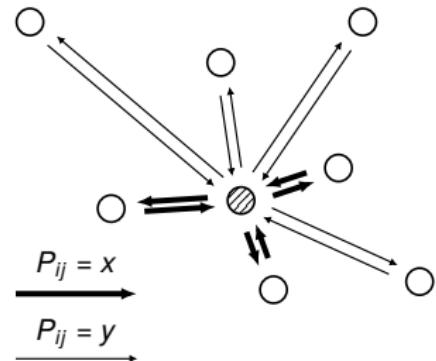


$$\begin{aligned}
 P_{\text{bidir}} &= \sum_{k=1}^m \sum_{l=1}^m F(x_k) x_k F(x_l | x_k) x_l \\
 &= \sum_{k=1}^m F(x_k) x_k^2.
 \end{aligned}$$

$$F(x_l | x_k) = \begin{cases} 1 & \text{if } l = k \\ 0 & \text{otherwise.} \end{cases}$$

*Relative overrepresentation*  $\rho$  is the fraction

$$\rho = \frac{P_{\text{bidir}}}{\mu^2} = \frac{\sum_{k=1}^m F(x_k) x_k^2}{\left(\sum_{k=1}^m F(x_k) x_k\right)^2}.$$



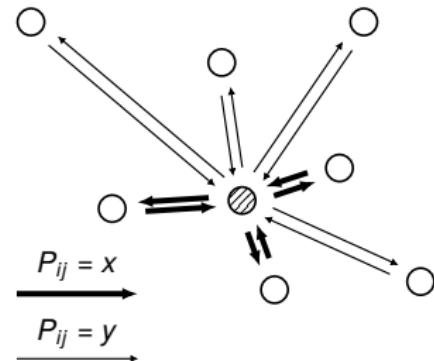
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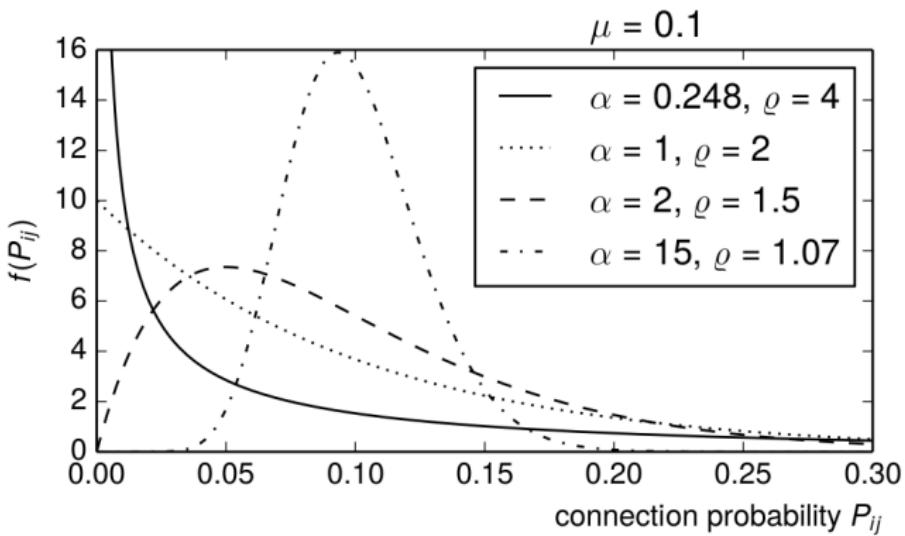
*Relative overrepresentation*  $\rho$  is the fraction

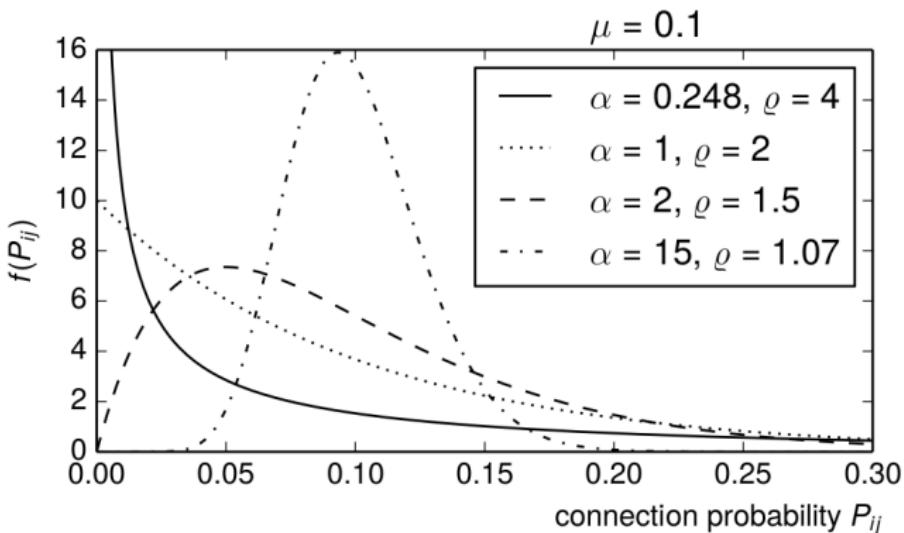
$$\rho = \frac{P_{\text{bidir}}}{\mu^2} = \frac{\sum_{k=1}^m F(x_k) x_k^2}{\left(\sum_{k=1}^m F(x_k) x_k\right)^2}.$$



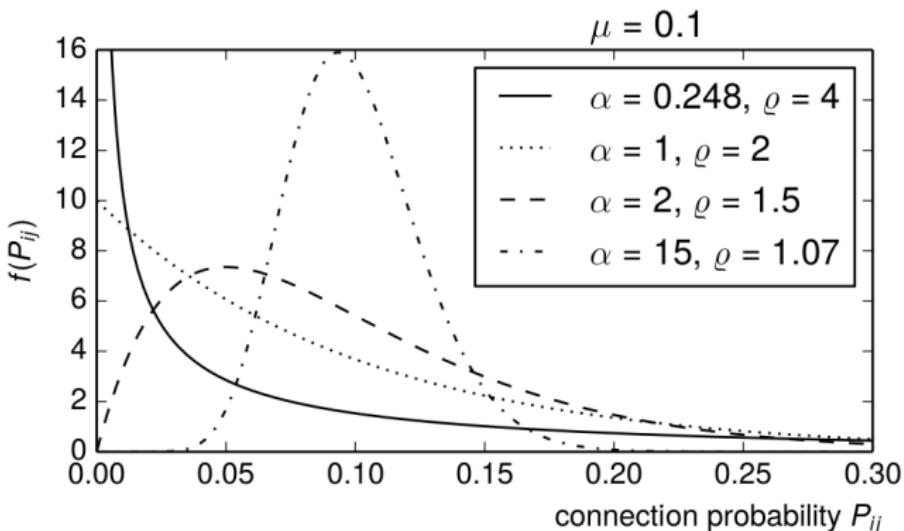
By Jensen's inequality,

$$\left( \sum_{k=1}^m F(x_k) x_k \right)^2 \leq \sum_{k=1}^m F(x_k) x_k^2 \quad \text{and thus} \quad \rho \geq 1.$$



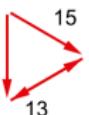
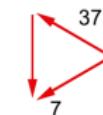
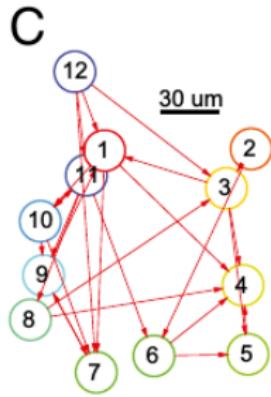
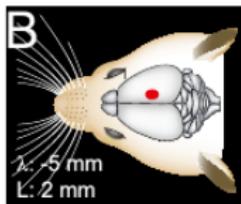
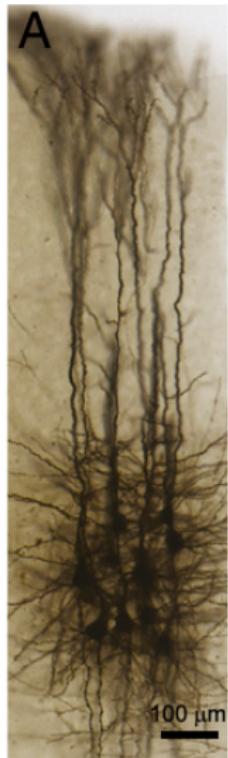


- multiple neuron properties together can cause strong overrepresentation



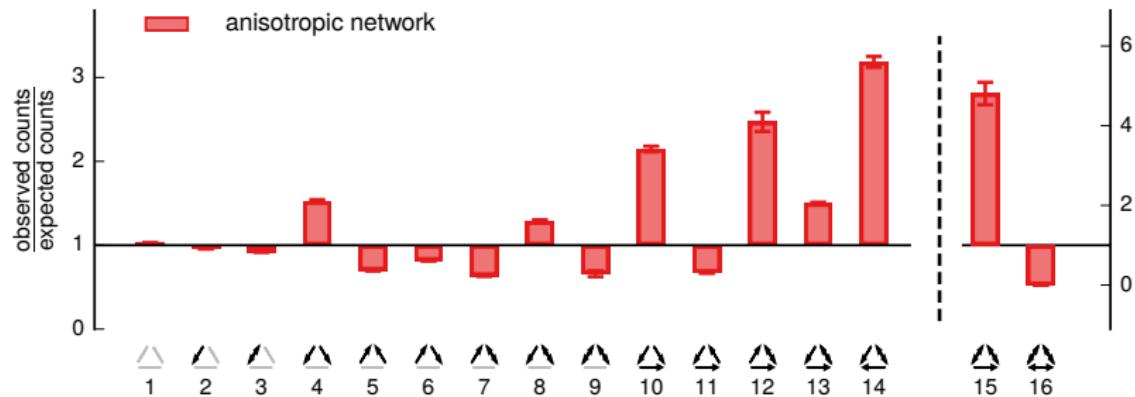
- multiple neuron properties together can cause strong overrepresentation
- example: higher connection probability in functionally related cells (Lee et al. 2016)

# Robust nonrandom connectivity patterns



Perin et al. 2011; Song et al. 2005; Markram et al. 1997; Miner and Triesch 2016; Gal et al. 2017; Vegué et al. 2017

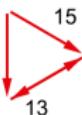
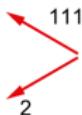
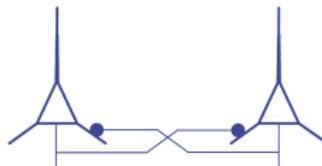
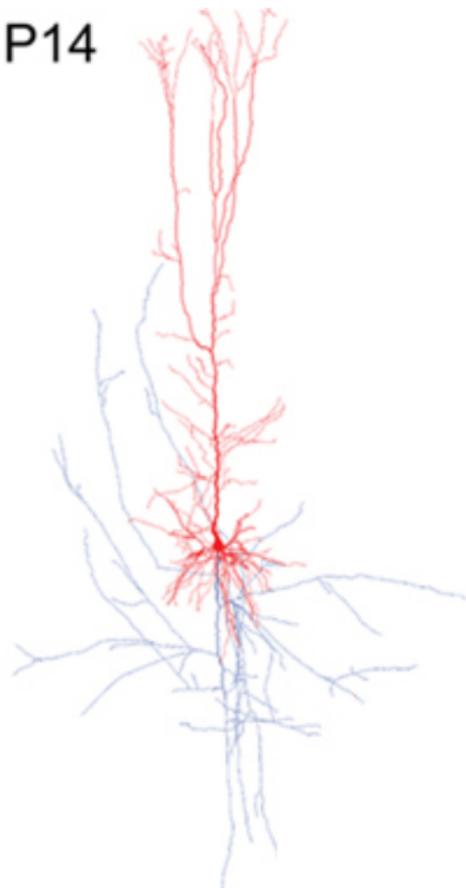
# Triplet motif statistics in anisotropic networks



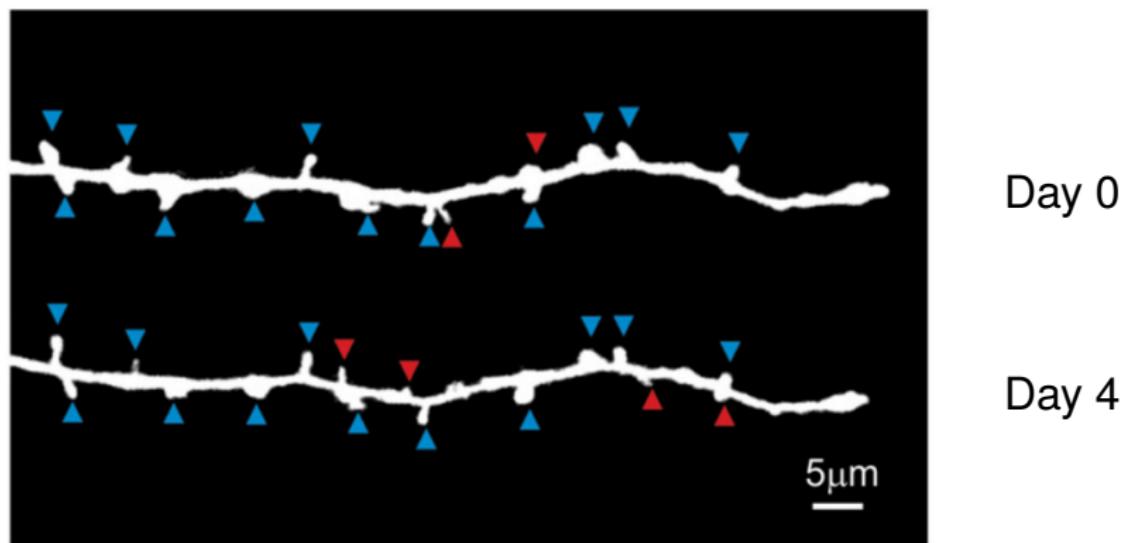
⇒ More about this: Poster W77, Wednesday!

# Nonrandom connectivity from anisotropy

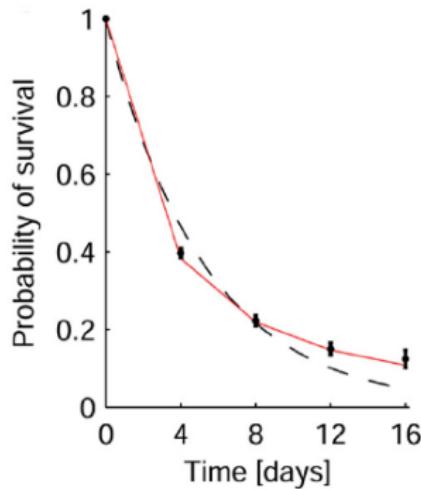
P14



# The dynamic connectome

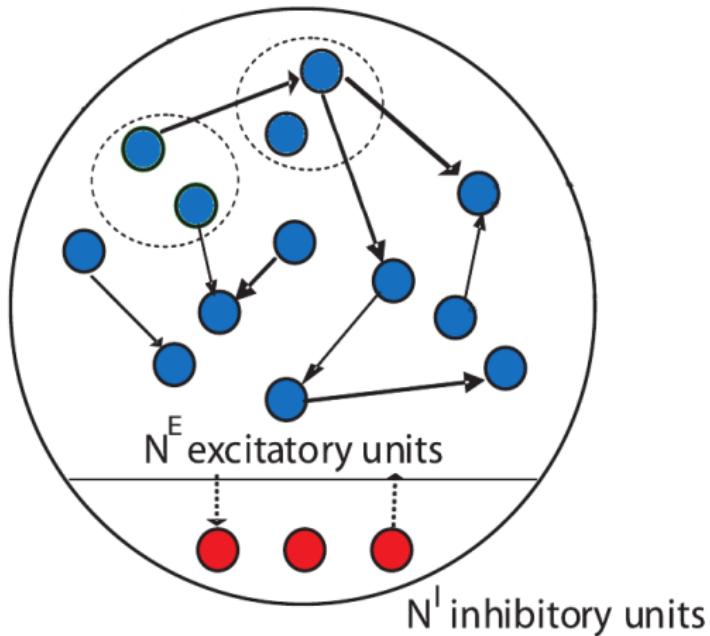


# The dynamic connectome

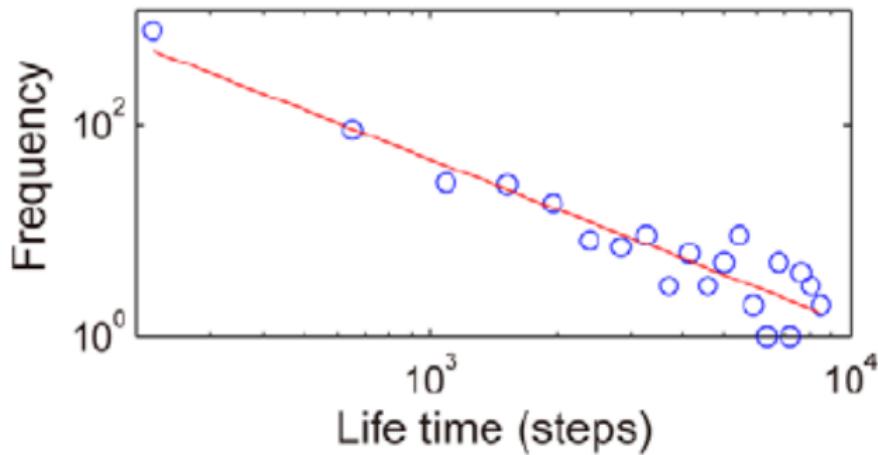


Survival probability of  $\rho(t) = (t + 1)^{-\gamma}$  with  $\gamma \approx 1.4$ ,  
equivalently lifetime distribution of  $f(t) = \gamma(t + 1)^{-(\gamma+1)}$

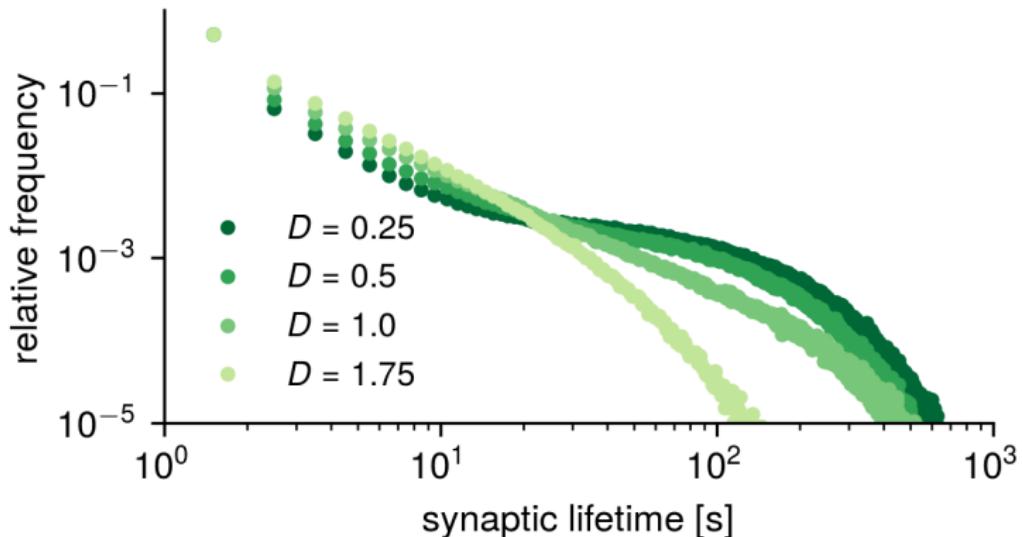
# Synapse dynamics in self-organizing recurrent networks



# Synapse dynamics in self-organizing recurrent networks



# Lifetimes in the self-organizing recurrent network



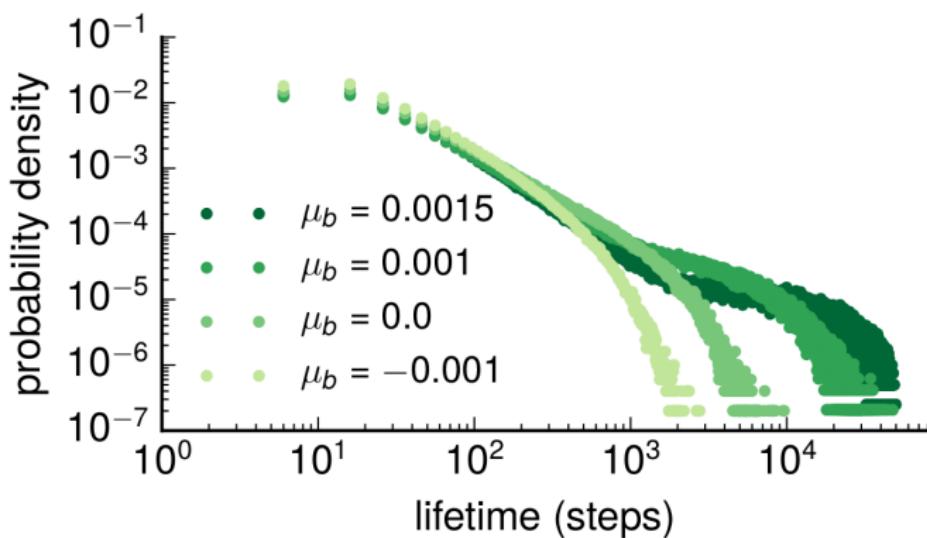
$D$  = LTD–LTP balance

# Lifetimes modelled by a stochastic process

Kesten process (Kesten 1973; Statman et al. 2014)

$$X_{n+1} = a_n X_n + b_n$$

as a model for synapse dynamics



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Stefano Cardanobile*

*Florence Kleberg  
Triesch lab*



**FIAS** Frankfurt Institute  
for Advanced Studies



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