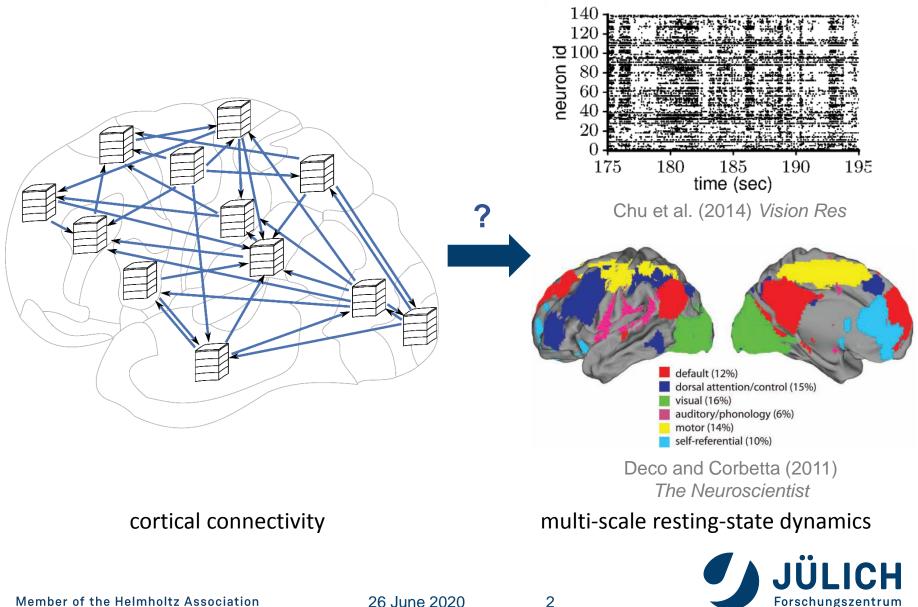


#### LARGE-SCALE SPIKING NEURAL NETWORK MODELS OF RESTING-STATE DYNAMICS IN PRIMATE CORTEX

BigBrain Workshop, 26 June 2020 | Sacha van Albada Institute of Neuroscience and Medicine (INM-6), Jülich Research Centre & Institute of Zoology, University of Cologne, Germany



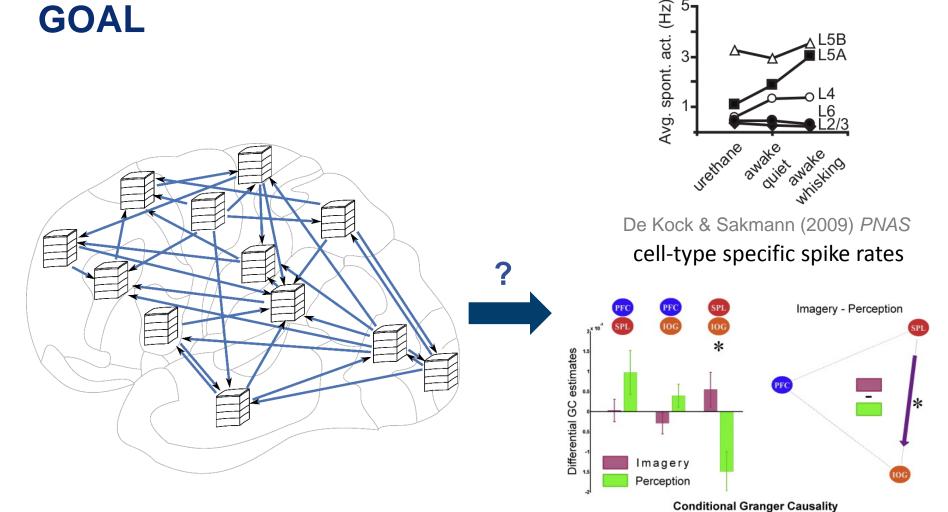
GOAL



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26 June 2020

GOAL



cortical connectivity

Dentico et al. (2014) NeuroImage

all units

5

#### inter-area propagation

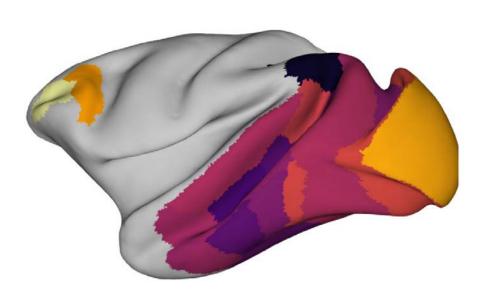


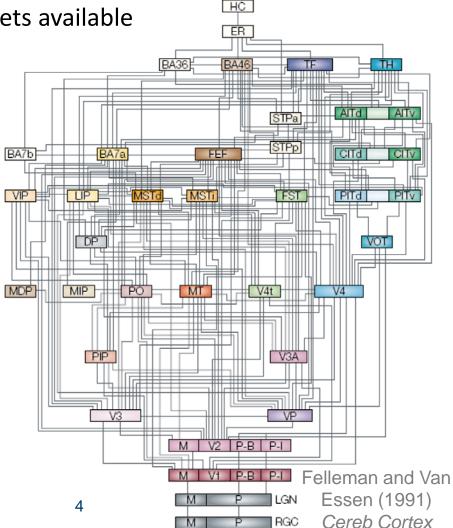
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### MULTI-AREA MODEL OF MACAQUE VISION-RELATED CORTEX

26 June 2020

- rich anatomical and physiological data sets available
- stepping stone to human
- regularities of organization





# MULTI-AREA MODEL OF MACAQUE VISUAL

- 800 million neurons in one hemisphere
- 32 areas in Felleman & Van Essen parcellation
- representing each area by a 1 mm<sup>2</sup> microcircuit
- 4.10<sup>6</sup> neurons and 2.4.10<sup>10</sup> synapses
- simulated using NEST on Jülich supercomputers



Schmidt



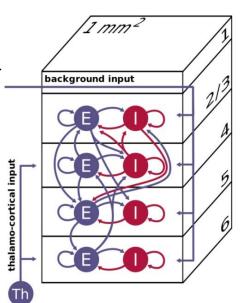
mm **T** Potjans adapted from Reid et al. (2009) 26 June 2020

Schmidt M, Bakker R, Shen K, Bezgin G, Diesmann M, van Albada SJ (2018) *PLOS CB* 

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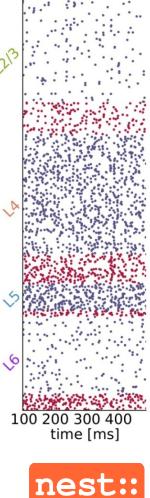
### MINIMAL LAYERED CORTICAL NETWORK MODEL

- 1 mm<sup>2</sup>
- 0.3 billion synapses; 80,000 leaky integrateand-fire neurons
- 2 populations of neurons per layer:
  - E: Excitatory
  - I: Inhibitory
- E and I identical neuronal dynamics
- laterally homogeneous connectivity
- layer- and type-specific connection probability based on collation of experimental data
- Poisson drive



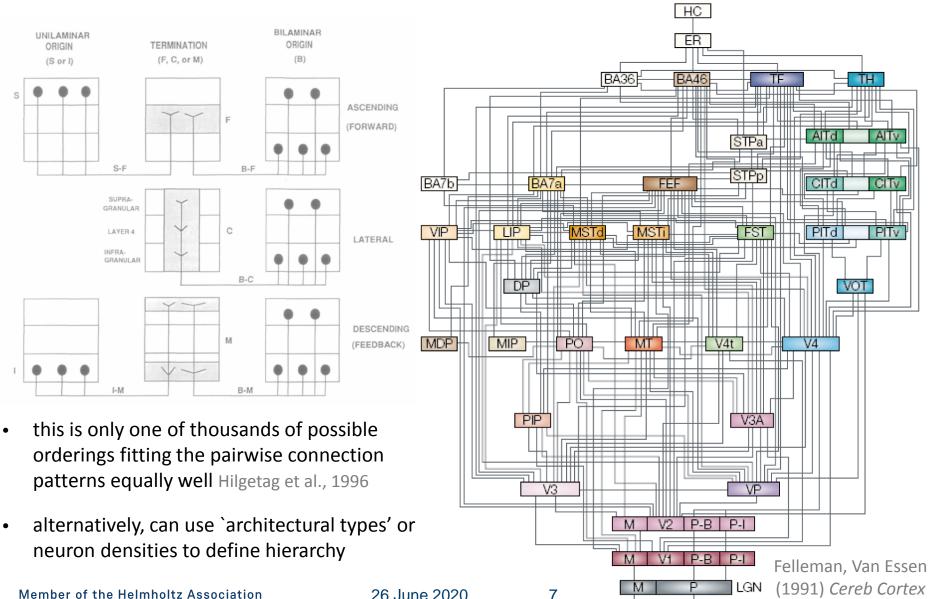
Potjans & Diesmann (2014) Cereb Cortex

now available in various simulators (NEST, PyNN, NetPyNE/NEURON, Brian) and in Open Source Brain Gleeson...van Albada et al. (2019) *Neuron* 





### HIERARCHY OF VISUAL CORTICAL AREAS



26 June 2020

P

RGC

### **DIFFERENTIAL LOCAL ARCHITECTURE**

# Agranular Dysgranular Eulaminate I Dombrowski et al. (2001) Cereb Cortex

synapse density remains roughly constant

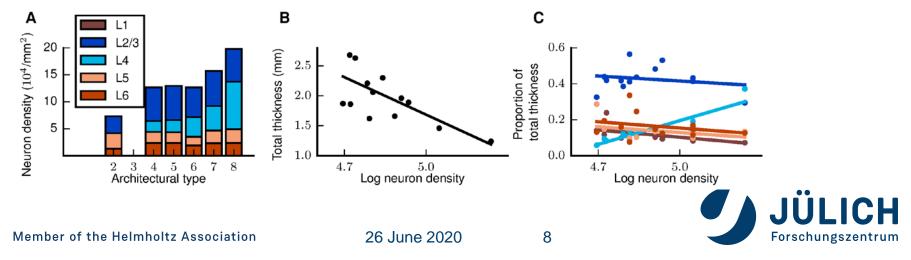
ightarrow higher areas receive more synapses per neuron

review on relationship between cortical architecture and connectivity: Hilgetag, Beul, van Albada, Goulas (2019) *Network Neurosci* 

- total cortical thicknesses and overall neuron densities for 14 areas
- estimated for remaining areas based on architectural types

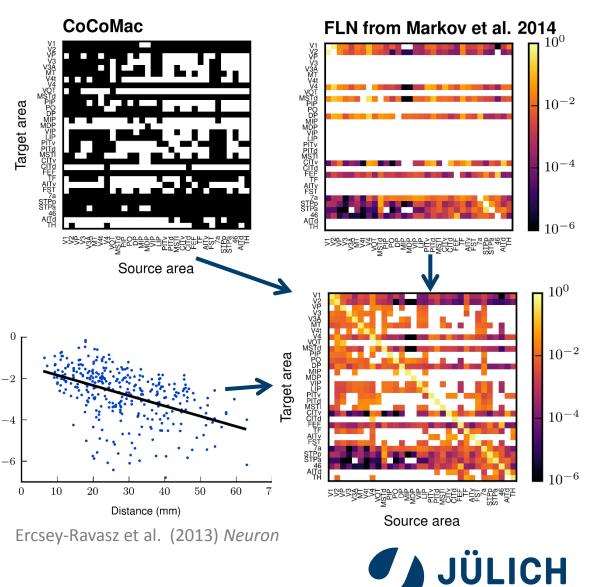
Hilgetag et al. (2016) NeuroImage

• reduction in L4 thickness toward higher areas based on micrographs from the literature



### **CONNECTIVITY MAP FROM TRACING DATA**

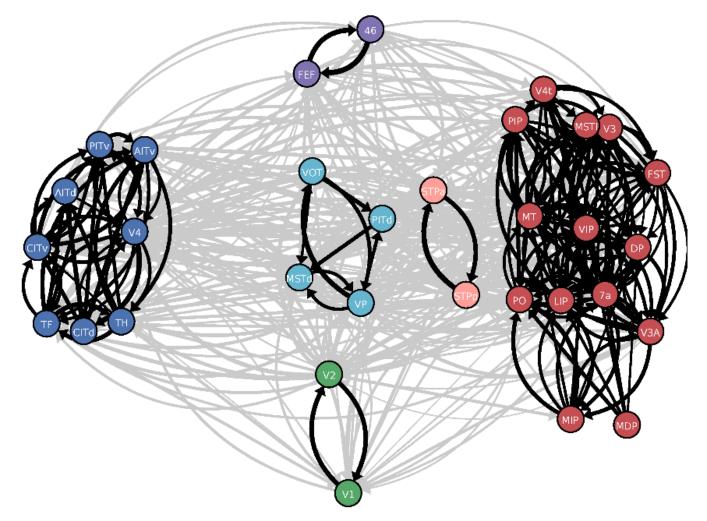
- partly binary, partly quantitative data
- connection probability decays with distance also for inter-area connections
- use this decay to estimate missing data based on distance between areas
- roughly 2/3 of area pairs are connected
- more important: connection density, spanning ~6 orders of magnitude



Forschungszentrum

 $\log(FLN)$ 

#### **PLAUSIBILITY CHECK**

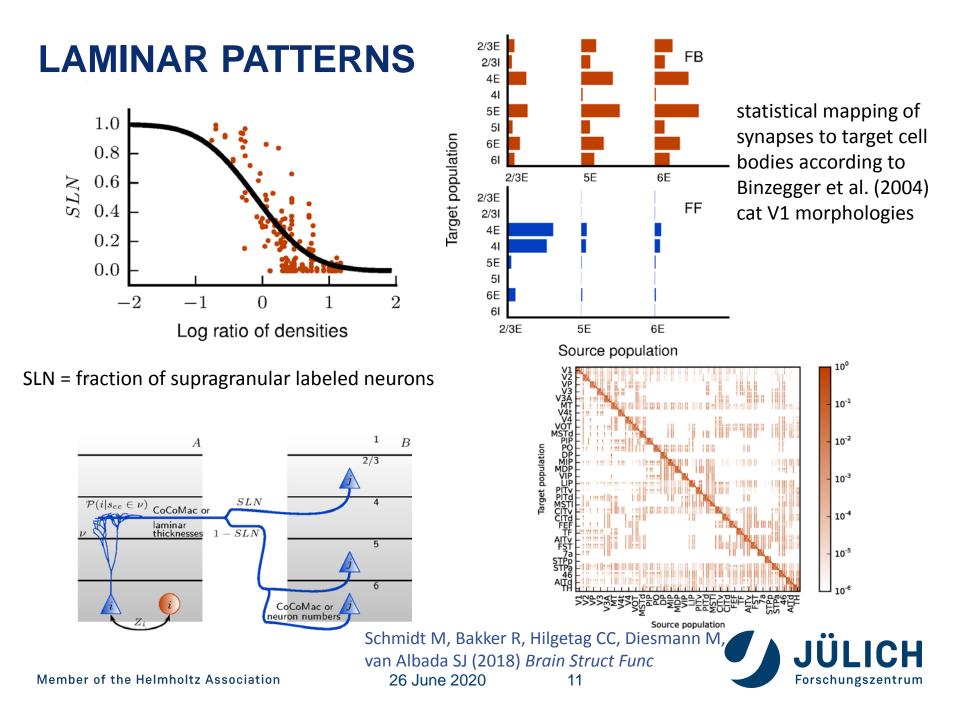


structural connectivity exhibits functionally relevant community structure

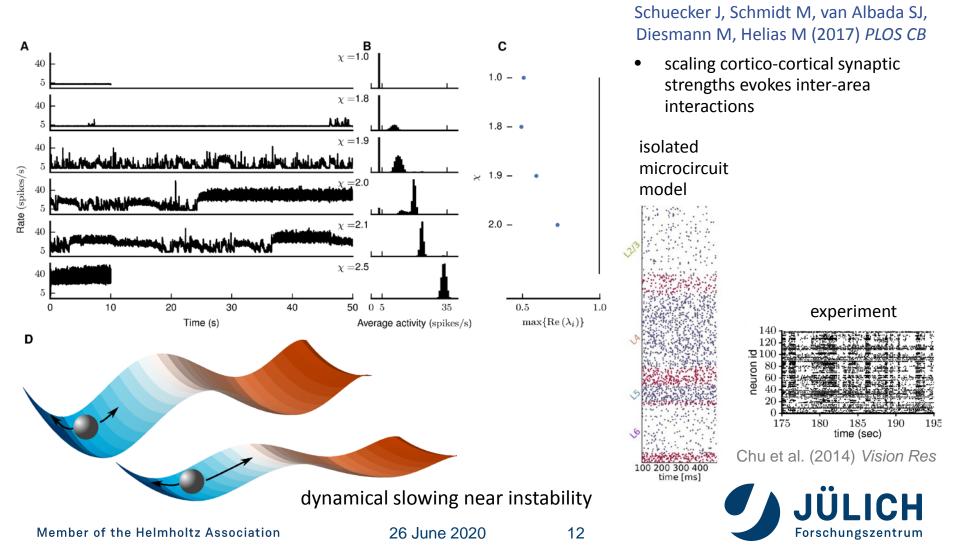


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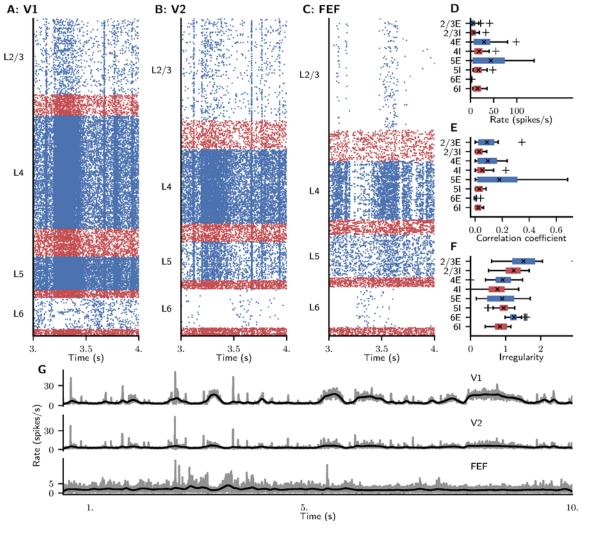
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#### INTER-AREA INTERACTIONS CAUSE SLOW FLUCTUATIONS • stable ground state obtained after mean-field-based stabilization



#### **GROUND STATE WITH MULTIPLE TIME SCALES**

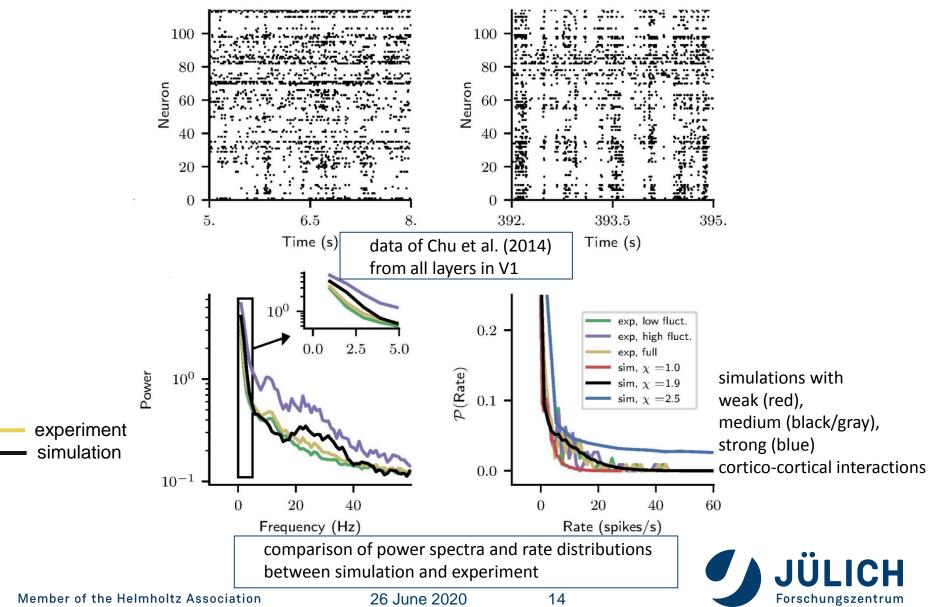


JÜLICH Forschungszentrum

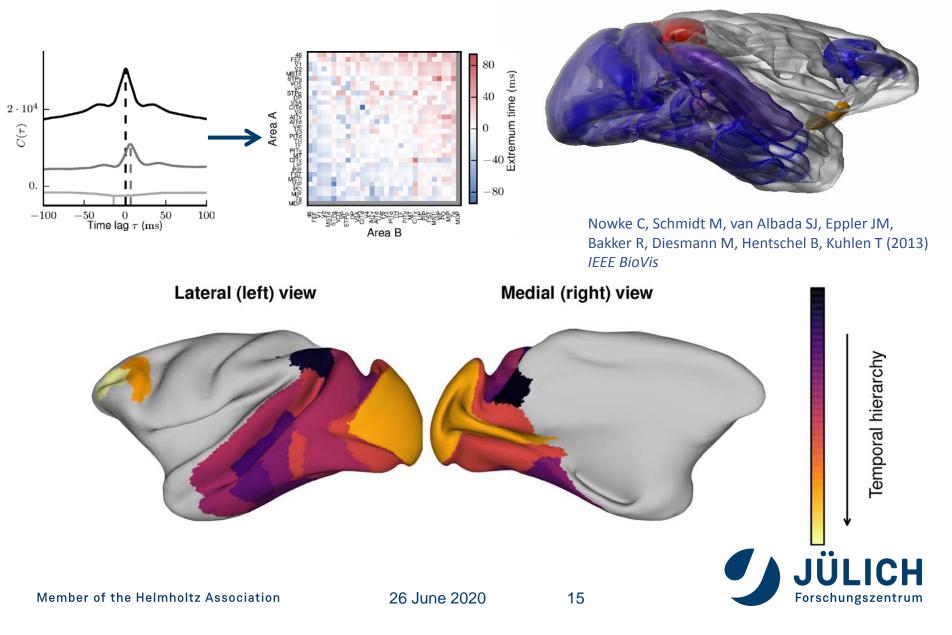
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#### **V1 SPIKING STATISTICS**

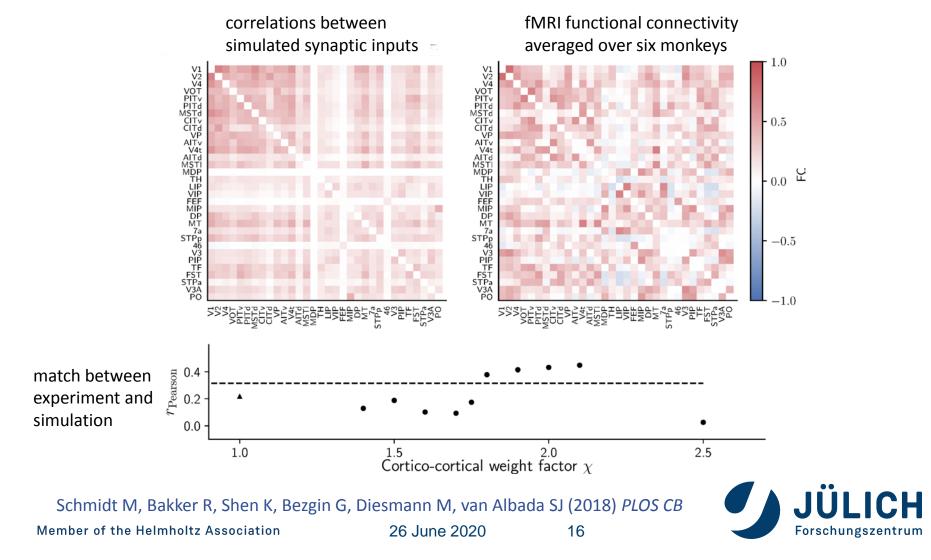


#### **TEMPORAL HIERARCHY**



#### **FUNCTIONAL CONNECTIVITY**

inter-area interactions in metastable state resemble experimental resting-state fMRI



#### **GITHUB REPOSITORY**

#### https://inm-6.github.io/multi-area-model/

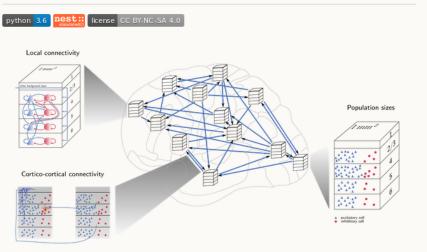
#### multi-areamodel

A large-scale spiking model of the visionrelated areas of macaque cortex.

View On GitHub

This project is maintained by <u>INM-6</u>

### Multi-scale spiking network model of macaque visual cortex



This code implements the spiking network model of macaque visual cortex developed at the Institute of Neuroscience and Medicine (INM-6), Research Center Jülich. The model has been documented in the following publications:

- Schmidt M, Bakker R, Hilgetag CC, Diesmann M & van Albada SJ Multi-scale account of the network structure of macaque visual cortex Brain Structure and Function (2018), 223: 1409 https://doi.org/10.1007/s00429-017-1554-4
- Schuecker J, Schmidt M, van Albada SJ, Diesmann M & Helias M (2017) Fundamental Activity Constraints Lead to Specific Interpretations of the Connectome. PLOS Computational Biology, 13(2): e1005179. https://doi.org/10.1371/journal.pcbi.1005179
- Schmidt M, Bakker R, Shen K, Bezgin B, Diesmann M & van Albada SJ (accepted) A multi-scale layer-resolved spiking network model of resting-state dynamics in macaque cortex. PLOS Computational Biology, 14(9): e1006359.



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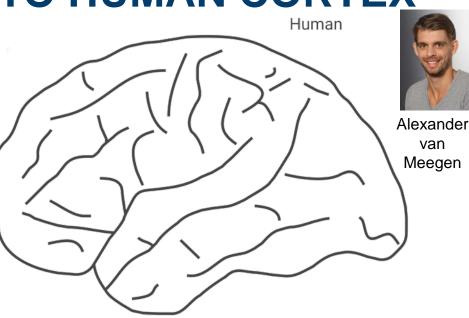
Hosted on GitHub Pages

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# FROM MACAQUE TO HUMAN CORTEX



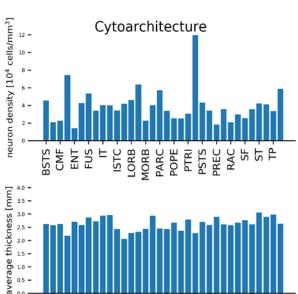




- 34 areas
- one hemisphere of human cortex
- Desikan-Killiany parcellation
- 3.6 million neurons
- 44 billion synapses
- cytoarchitectonic data
  - von Economo & Koskinas (1925)
  - layer thicknesses (Wagstyl et al. 2020 PLOS Biol) extracted from BigBrain
  - neuron densities measured by Timo Dickscheid et al. extracted from BigBrain where available





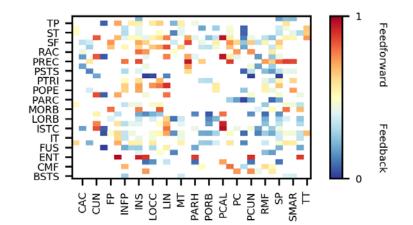


CAC CUN CUN CUN INF CUN CAL PCAL Jari Pronold

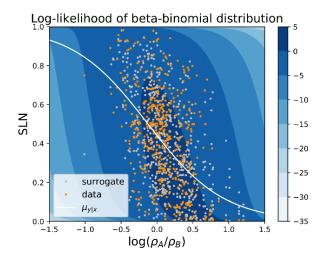
SMAR TT

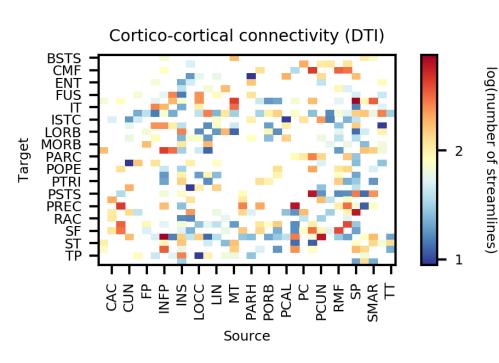
# HUMAN CONNECTIVITY DATA AND PREDICTIVE CONNECTOMICS

Directionality



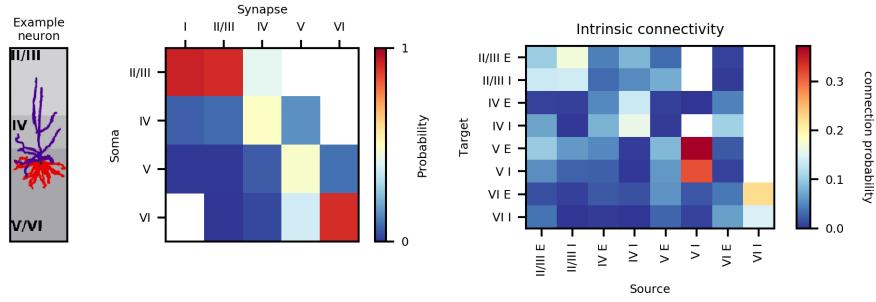
Directionality of cortico-cortical connections based on fit of macaque tracing data





Data: Van Essen et al. (2013) NeuroImage

# **ASPECTS OF LOCAL CONNECTIVITY**



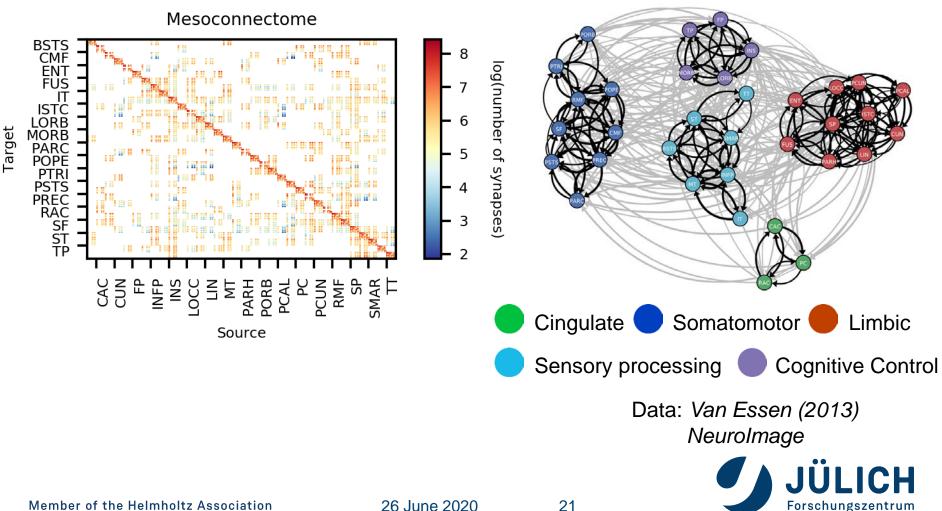
Synapse-to-soma probability

Data: Mohan et al. (2015) Cereb Cortex

Data: Potjans & Diesmann (2014) Cereb Cortex



## **STRUCTURAL CONNECTIVITY**



**MEAN-FIELD DYNAMICS** 

 mean-field theory describes population-averaged rates:

$$\begin{aligned} \frac{d\nu_i}{dt} &= -\nu_i + \phi(\mu_i, \sigma_i), \\ \mu_i &= \tau_{\rm m} \sum_j J_{ij} \nu_j + \tau_{\rm m} J_{\rm ext} \nu_{\rm ext} + \xi_i(t), \\ \sigma_i^2 &= \tau_{\rm m} \sum_j J_{ij}^2 \nu_j + \tau_{\rm m} J_{\rm ext}^2 \nu_{\rm ext}. \end{aligned}$$

- enables systematic exploration of parameter space
- strong effect of inhibitory connections onto layer 5 excitatory neurons

ate <

≤ A

0

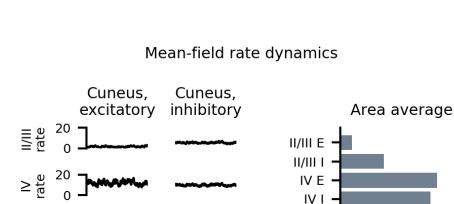
40

Time [ms]

0

Time [ms]

40





5

Rate [spks/s]

10

VΕ

V I VI E

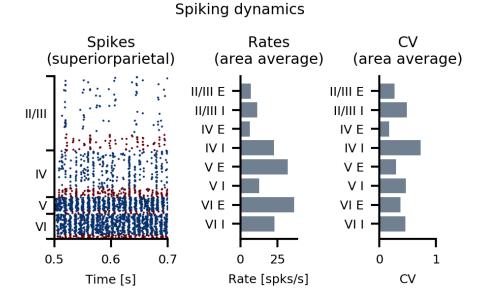
VH

0

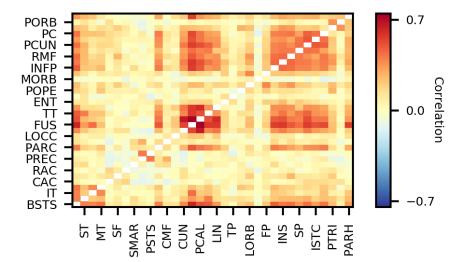


Alexander van Meegen

## SIMULATED DYNAMICS



#### Functional connectivity





#### ACKNOWLEDGMENTS Jülich

modeling,

modeling, theory





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Jari Pronold

anatomy

Hannah Vollenbröker



modeling

Aitor Morales-Gregorio

Rembrandt

Bakker

anatomical

data



Jannis

Schuecker



theory



Moritz Helias

Markus Diesmann



Human Brain Project



grant JINB33 for compute time on the Jülich supercomputers



fMRI data



Kelly Shen Baycrest, Toronto



Gleb

Bezgin

McGill,

**Montreal** 

Member of the Helmholtz Association

Claus Hilgetag UKE Hamburg



Christian Nowke

Bernd Hentschel

**RWTH Aachen University** 

Torsten



26 June 2020

