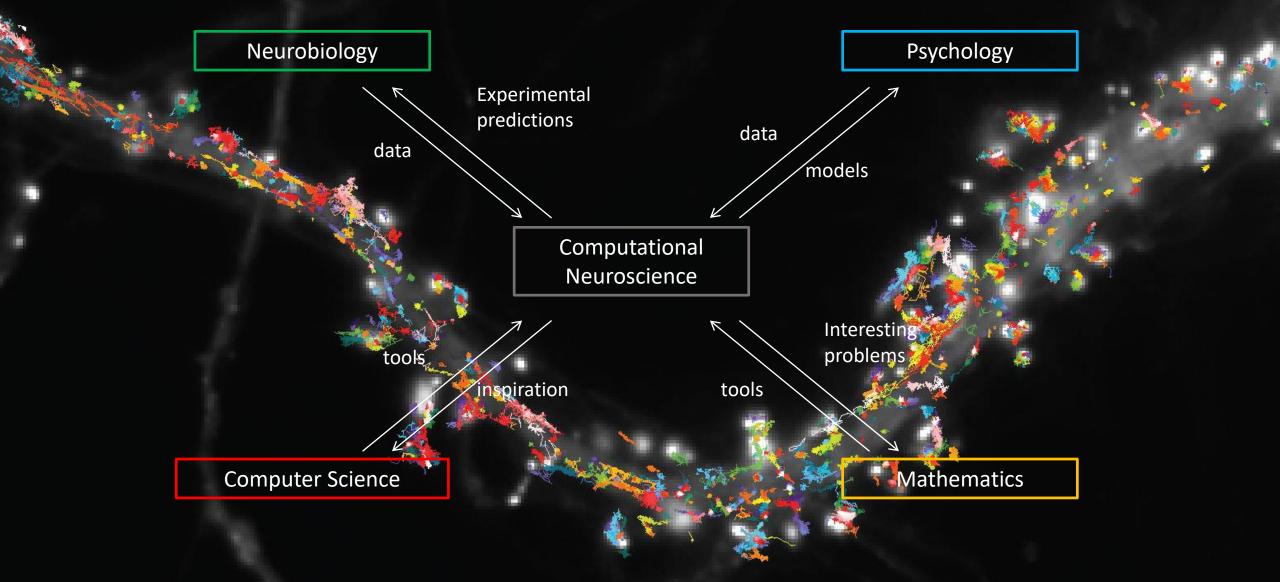
Computational Neuroscience

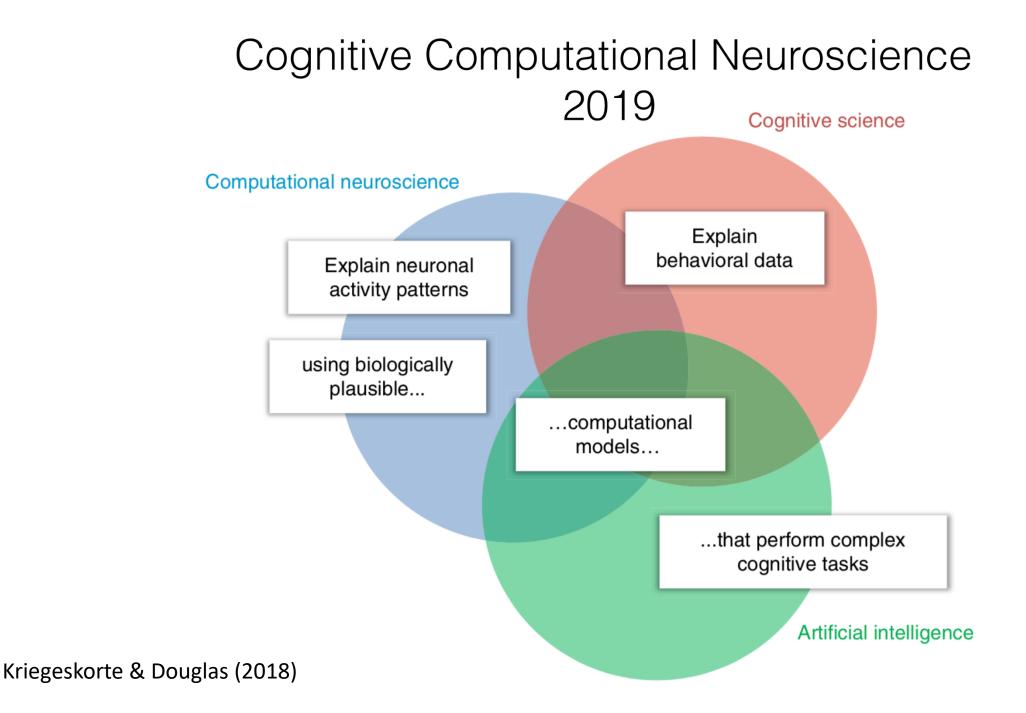
Gunnar Blohm

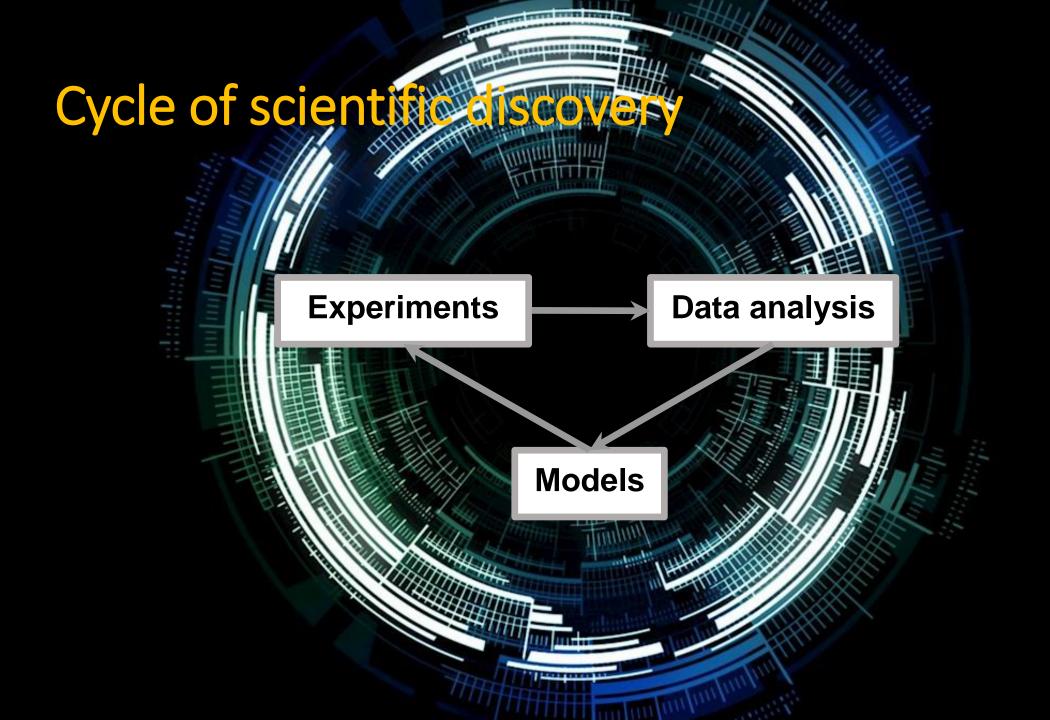
What is Computational Neuroscience?

- A multi-disciplinary <u>approach</u> to studying the brain
- The use of mathematics, engineering and computer simulations in neuroscience
 - **Statistics**
 - Neural networks / Machine learning
 - Dynamical systems theory
 - Control theory
- Includes computational analysis methods!

A multi-disciplinary endeavour







What are models?

Insights not directly accessible by experiments / data

Models allow for **understanding** and **control** (Rosenblueth & Wiener, 1945)

Interventions, e.g. experimental, clinical

Requires model validation \rightarrow experiments! In other words, a model is a **Hypothesis**!

Studying complex systems requires computational approaches!

Why is modeling important?

- Gain complete understanding of some experimental phenomenon
- Estimate latent states
- Identify hypotheses, assumptions, unknowns
- Make quantitative predictions
- Build a theoretical brain as a model of the real brain (e.g. stroke lesions)
- Inspire new technologies
- Models of neurological diseases to help treatment, rehabilitation, quality of life
- Guidance in designing useful experiments (i.e. animal research)

What can we do?

Models help answering three potential types of questions about the brain

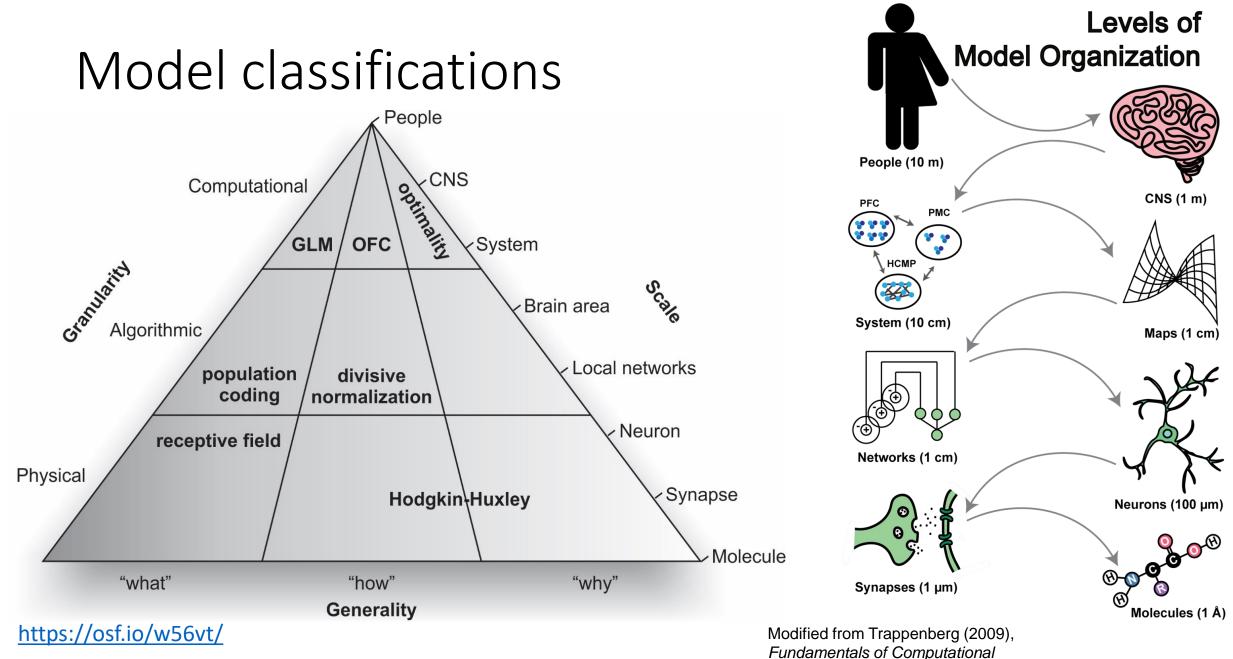
- Descriptive = What?
 - Compact summary of large amounts of data
- Mechanistic = How?
 - Show how neural circuits perform complex function
- Interpretive = Why?
 - Computations in the brain are usually performed in an optimal or nearly optimal way
 - Understanding optimal algorithms and their implementation to explain why the brain is designed the way it is

Dayan & Abbott, 2001

Marr's model hierarchy

- Computational level 1
 - Objective?
 - How close to optimal?
 - This is what most computational neuroscience papers do!
- Algorithmic level 2
 - Data structures?
 - Approximations?
 - Runtime?
 - Some studies get into this (computer science)
- Implementation level -3
 - Hardware? Neurons? Synapses? Molecules?
 - Not addressed enough!

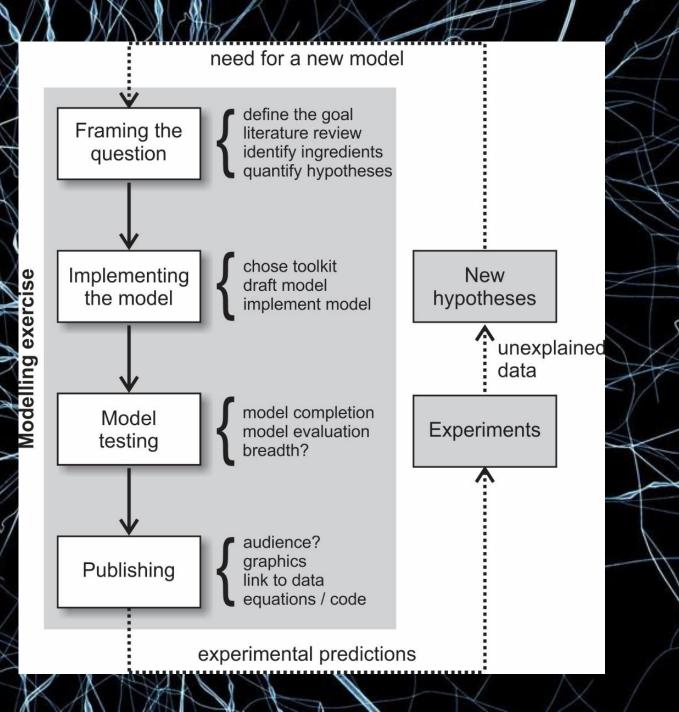
https://onlinelibrary.wiley.com/doi/full/10.1111/tops.12137



Fundamentals of Computatio

How to model...?

https://osf.io/8gz65/



What's a good model?

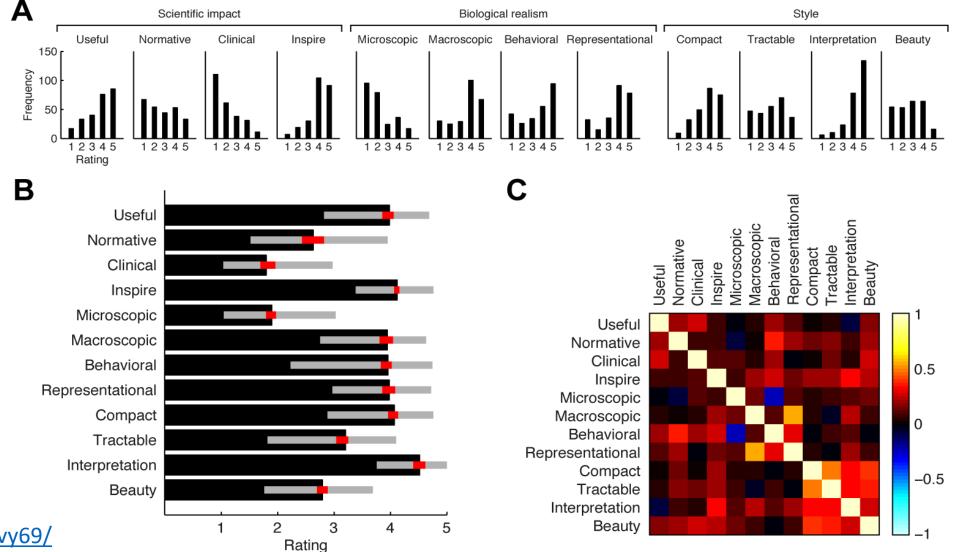
• Explain data

- p(data | model 1)>>p(data | model 2)
- Model interfaces with data / data connects to model
- Generalization
 - Out of experiment (new experiment)
 - Out of sample (same experiment, new data)
 - Occam's razor
 - Reproducibility?
 - Data / experiment / model?
 - Robustness
- New insight
 - Predictions on new domains
 - Interpretability
 - Transfer knowledge by model equivalence between domains

Usefulness

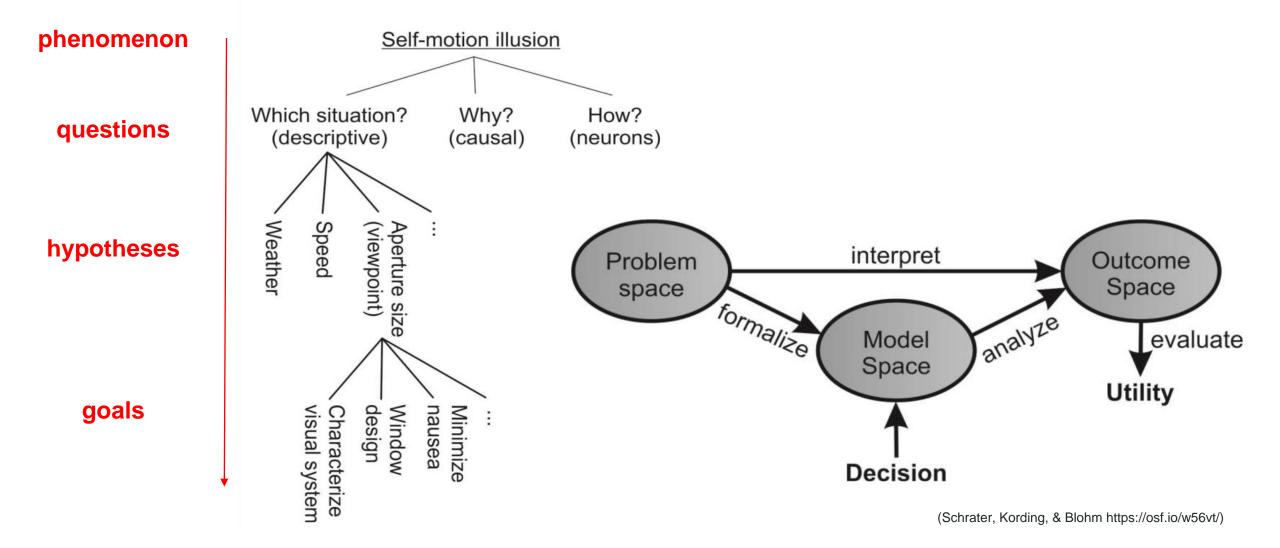
- Predictions
- Inspires new experiments
- Interventions / policy changes
- Tech applications
- Clinical relevance
- Falsifyable
- Make assumptions explicit
- Elegance
 - Non-arbitrary structure...
 - Simplicity
 - Computational complexity low
 - Normativity
 - Unification / subsumption
 - Micro-/meso-/macro-realism

People care about different things...



https://osf.io/3vy69/

Modeling as a decision process



3 examples...

•

Vie

Normative models of behaviour

$$\mathbf{x}^{(k+1)} = A\mathbf{x}^{(k)} + C\mathbf{u}^{(k)} + \mathbf{\epsilon}^{(k)}_{u}$$

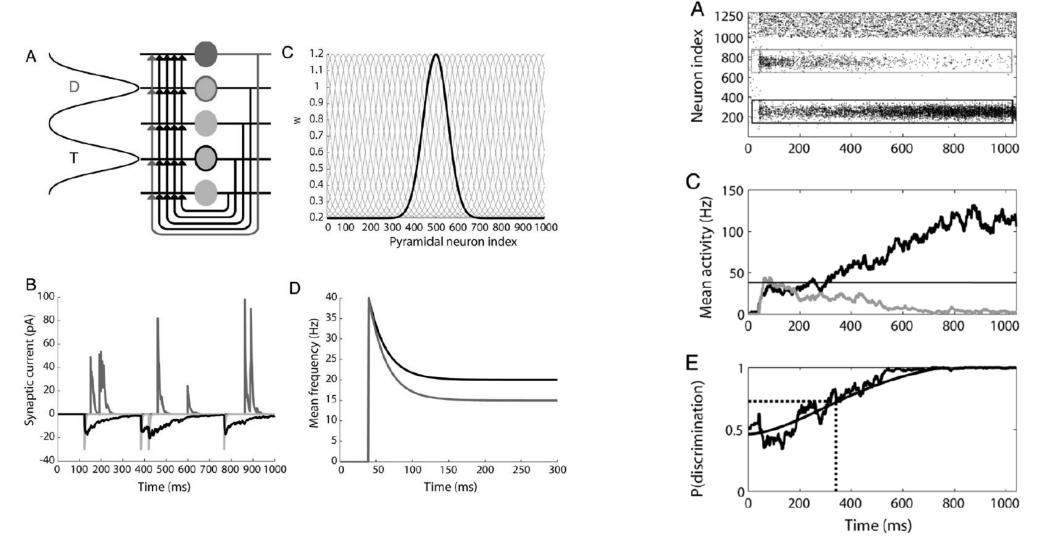
$$\mathbf{y}^{(k)} = B\mathbf{x}^{(k)} + \mathbf{\epsilon}^{(k)}_{y}$$

$$J = \sum_{k=0}^{p-1} \mathbf{u}^{(k)T} L^{(k)} \mathbf{u}^{(k)} + \mathbf{y}^{(k+1)T} T^{(k+1)} \mathbf{y}^{(k+1)}$$

$$\mathbf{u}^{(k)} = G^{(k)} \hat{\mathbf{x}}^{(k)}$$

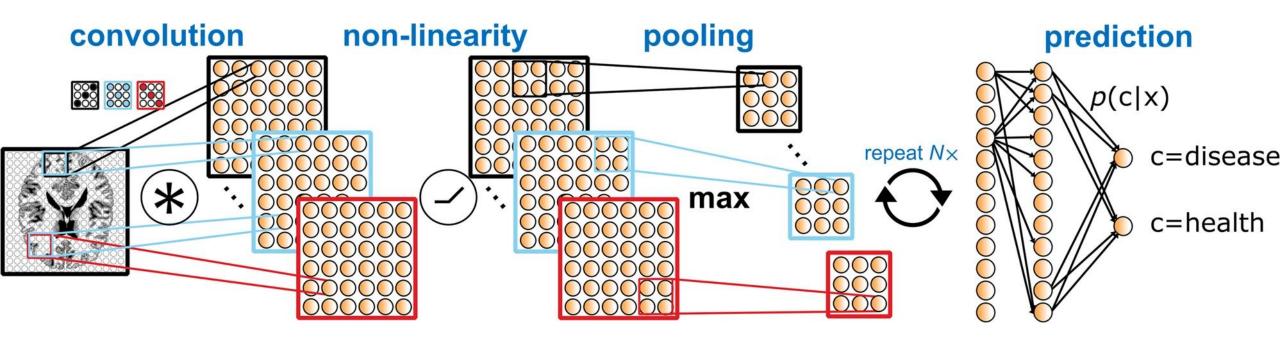
$$\hat{\mathbf{x}}^{(k+1)} = \hat{A} \hat{\mathbf{x}}^{(k)} + \hat{A} K^{(k)} \left(\mathbf{y}^{(k)} - \hat{\mathbf{y}}^{(k)} \right) + \hat{C} \mathbf{u}^{(k)}$$

Spiking network models



Standage & Pare 2011

Clinical models



Durstewitz, et al., 2019

Computational Neuroscience is key to scientific progress!

Want to know more?

Reuromatch academy

https://compneuro.neuromatch.io

Intro to modeling WIDI WID2