

Advanced cognitive modeling

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Course material: https://github.com/LegrandNico/CognitiveModeling







Introduction to cognitive and computational modeling

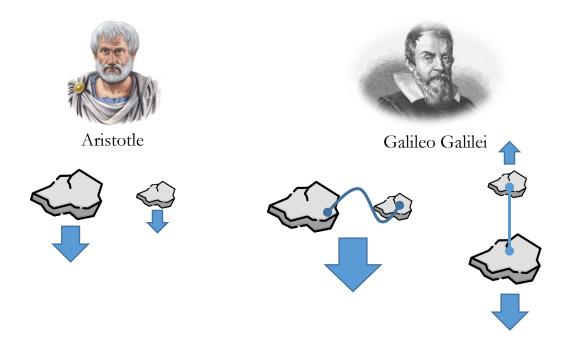
Why cognitive/computational models are interesting?

What is a cognitive/computational model?

The example of cognitive computational neuroscience and computational psychiatry.

Introduction

Computational modeling is not something new.



Actually, this is not computational, and this is not about cognition, BUT...

- Data never speak for themselves
- A model instantiates algorithmic hypotheses
- The model explains the data
- The model is unobservable and exists only in the mind of people who use it
- There are many models for a single dataset
- We can compare models with each other
- Models can be checked for internal consistencies
- Model can make predictions and compare with experimental data
- Models can postulate the existence of hidden/latent phenomenon's

"Verbally expressed statements are sometimes flawed by internal inconsistencies, logical contradictions, theoretical weaknesses and gaps. A running computational model, on the other hand, can be considered as a sufficiency proof of the internal coherence and completeness of the ideas it is based upon"

Fum, Del Missier, and Stocco (2007), p. 136

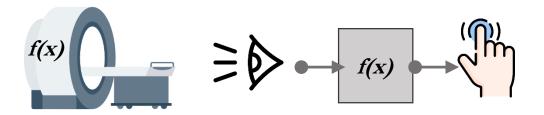








Introduction

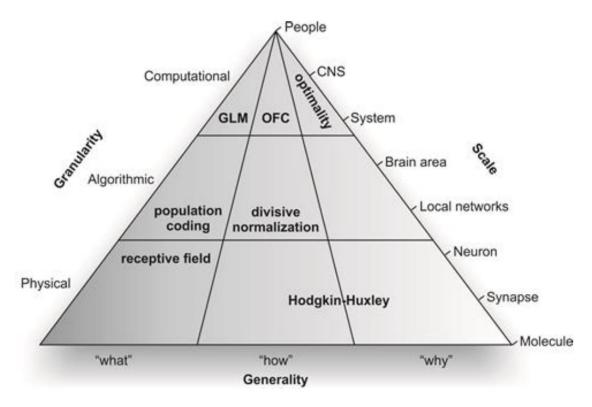


Behaviors and neurocognitive processes are complex phenomenon's that cannot be fully understood just by looking at it.





Models should be as simple as possible but as detailed as needed.



Schrater et al. - Modeling in neuroscience as a decision process







Introduction

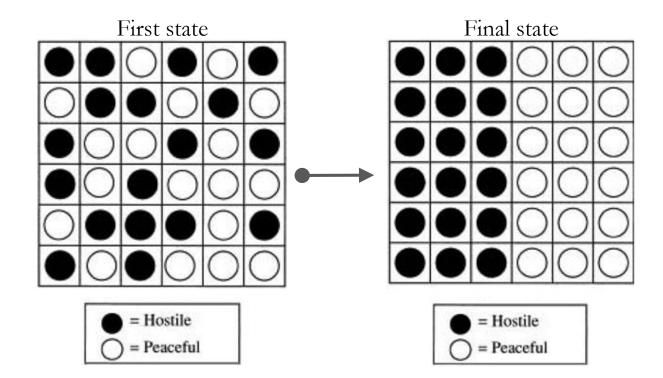
If $n_{\text{Neighbors}} = \text{Hostile} > n_{\text{Neighbors}} / 2$

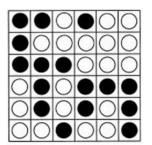
 $State_{t+1} = Hostile$

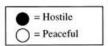
State_{t+1}=Peaceful

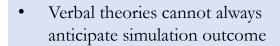
Else

The emergence of social structure from simple individual behaviors.

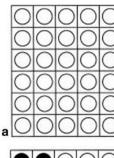


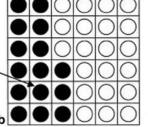


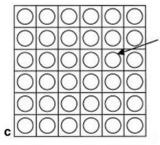


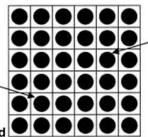


- The simulation yield insight on emergent behavior beyond the rules initially programmed
- It would have been difficult to do the backward inference
- Complex behaviors can be reduced to very simple assumptions.









Kenrick DT, Li NP, Butner J (2003) Dynamical evolutionary psychology: individual decision rules and emergent social norms. *Psychol Rev 110:3–28*.



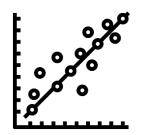






Models of cognition

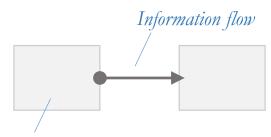
Data analysis models



Help establish relationships between measured variables

They are not models of brain information processing.

Box-and-arrow models



Cognitive component functions

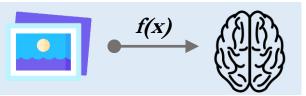
Models of the world

- Model-based reinforcement learning
- Model-based cognition

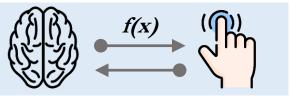
Brain-computational model (BCM)

Mimics the brain information processing

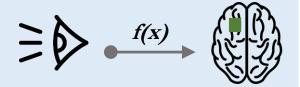
Deep neural nets



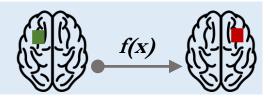
Reinforcement learning models



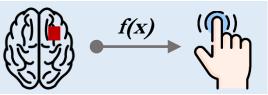
Sensory encoding models



Internal transformation models



Behavioral decoding models



Kriegeskorte and Douglas (2018). Nature neuroscience





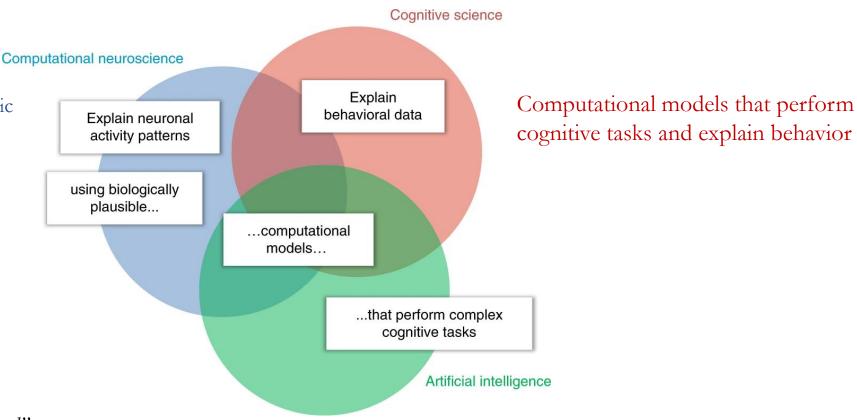




Cognitive computational neuroscience

Create biologically plausible computational models that perform cognitive tasks and explain brain information processing and behavior.

Neurobiologically plausible mechanistic models that explain brain activity



"What I cannot create, I do not understand" Richard Feynman

Kriegeskorte and Douglas (2018). Nature neuroscience



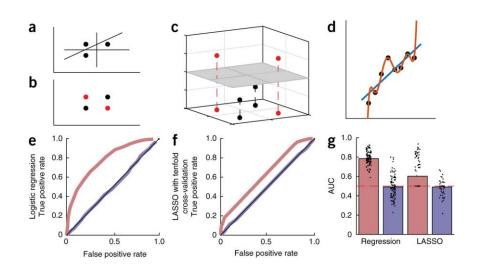




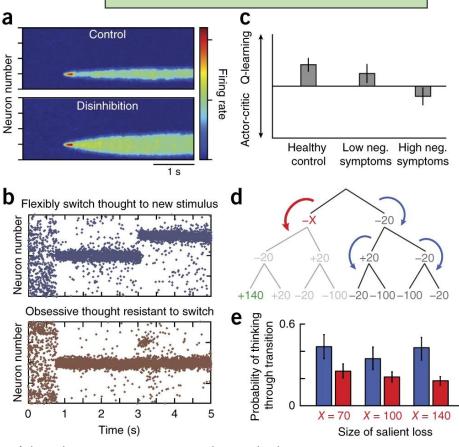


Computational psychiatry

Data-driven



Theory-driven



Bayesian decision theory broadly provides three routes to psychopathology:

- solving the wrong problem correctly (e.g. consistently prioritizing alcohol intake over health)
- solving the correct problem wrongly (e.g. using alcohol to 'treat' emotional problems),
- solving the correct problem correctly, but in an unfortunate environment or after unfortunate prior experiences (e.g. having persecutive worries after persecutory experiences).

Huys et al. (2016) Nat. Neurosci.







Course goals

1. (Hierarchical) Bayesian modeling.



2. (Deep) Reinforcement learning.











Model goals

Descriptive

WHAT

Describe the data

Mechanistic

HOW

Describe the data in term of mechanisms

Explanatory



Describe the data in term of mechanisms following optimal solutions to (constrained) problems in the real world







Resources

Huys, Q. J. M., Maia, T. V., & Frank, M. J. (2016). Computational psychiatry as a bridge from neuroscience to clinical applications. Nature Neuroscience, 19(3), 404–413. https://doi.org/10.1038/nn.4238

Kriegeskorte, N., & Douglas, P. K. (2018). Cognitive computational neuroscience. Nature Neuroscience, 21(9), 1148–1160. https://doi.org/10.1038/s41593-018-0210-5

Lewandowsky, S. & Farrell, S. (2011). *Computational modeling in cognition:* principles and practice. Thousand Oaks: Sage Publications.

Forstmann, B. & Wagenmakers. (2015). *An introduction to model-based cognitive neuroscience*. New York, NY: Springer. Ch. 1: An Introduction to Cognitive Modeling.

BayesCog Summer 2020 (Lei Zhang)

https://www.youtube.com/watch?v=8RpLF7ufZs4&list=PLfRTb2z8 k2x9gNBypgMIj3oNLF8lqM44-

Neuromatch Academy (W1D1)

- https://www.youtube.com/watch?v=KxldhMR5PxA
- https://www.youtube.com/watch?v=KZQXfQL1SH4



