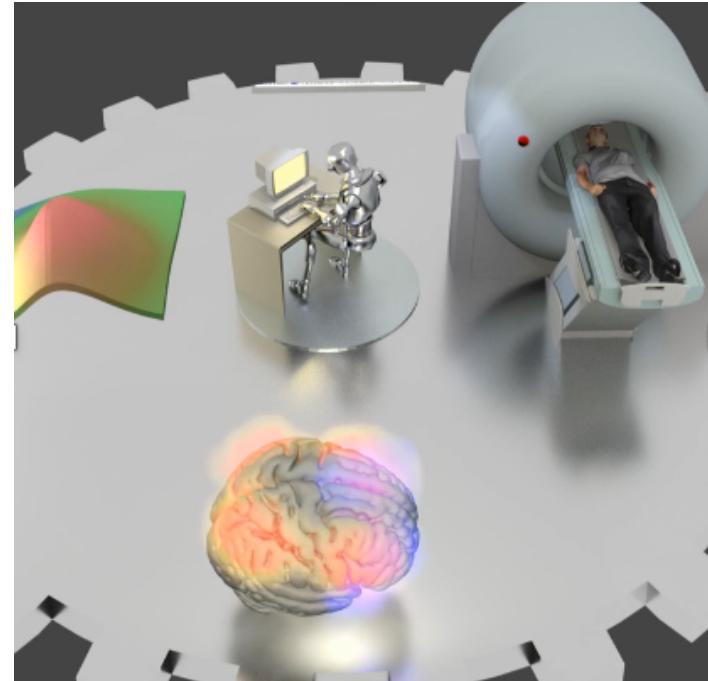


Neuroadaptive Bayesian Optimization

Implications for the Cognitive Sciences

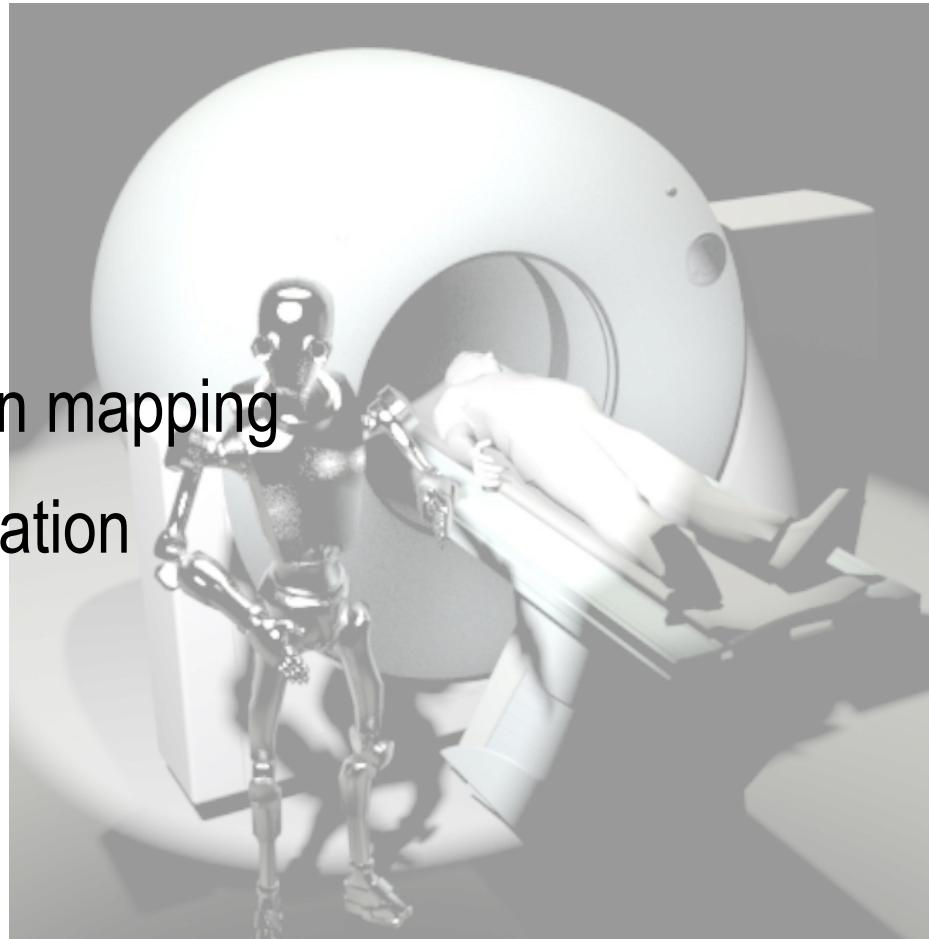
Romy Lorenz

Postdoctoral Research Fellow
Cognitive, Clinical and Computational Neuroimaging Lab
Imperial College London



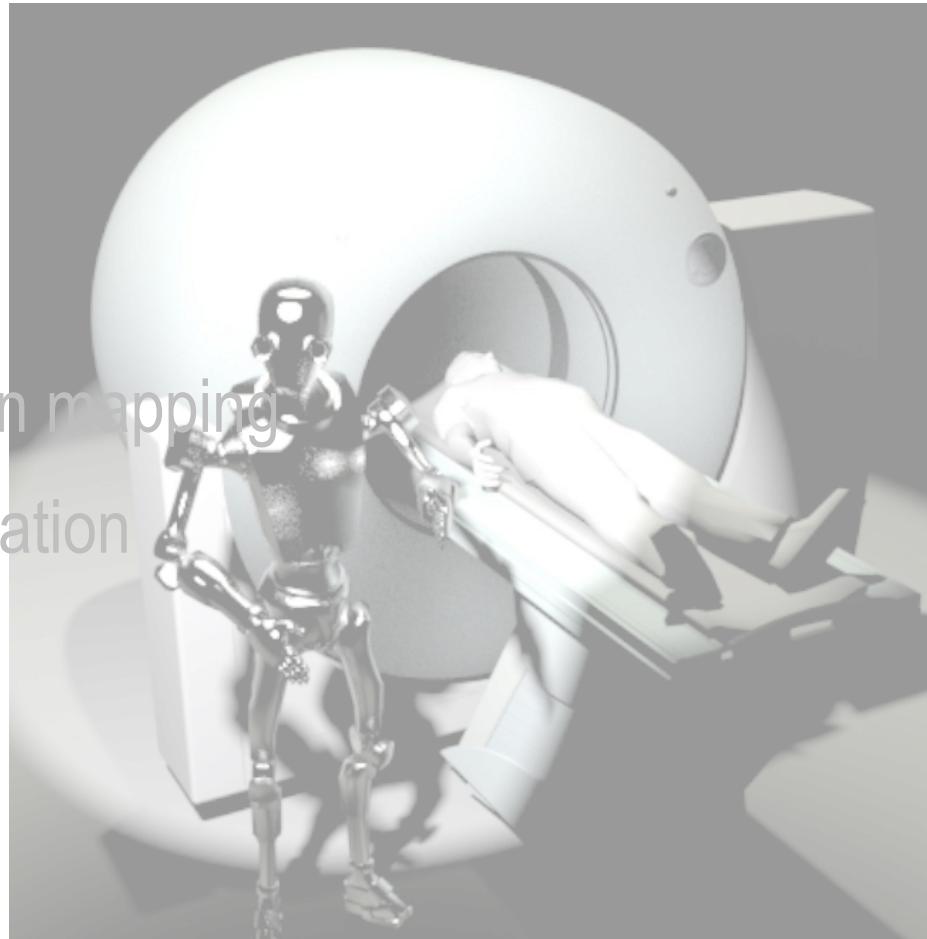
Overview

1. Motivation
2. The framework
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4. Application 1: Human brain mapping
5. Application 2: Brain stimulation
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7. Implications & Discussion

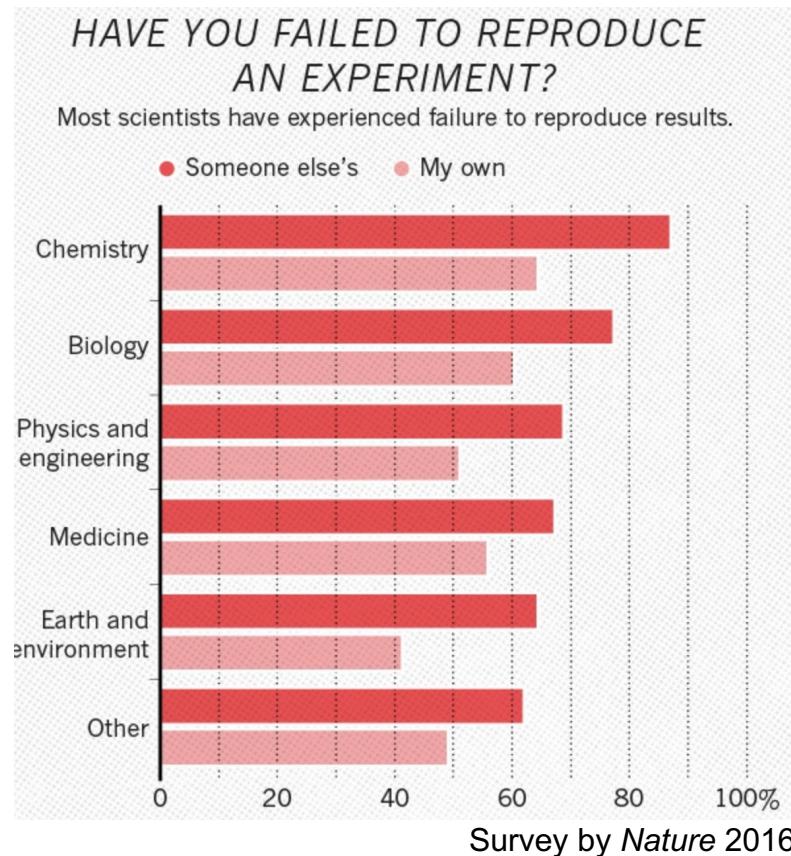
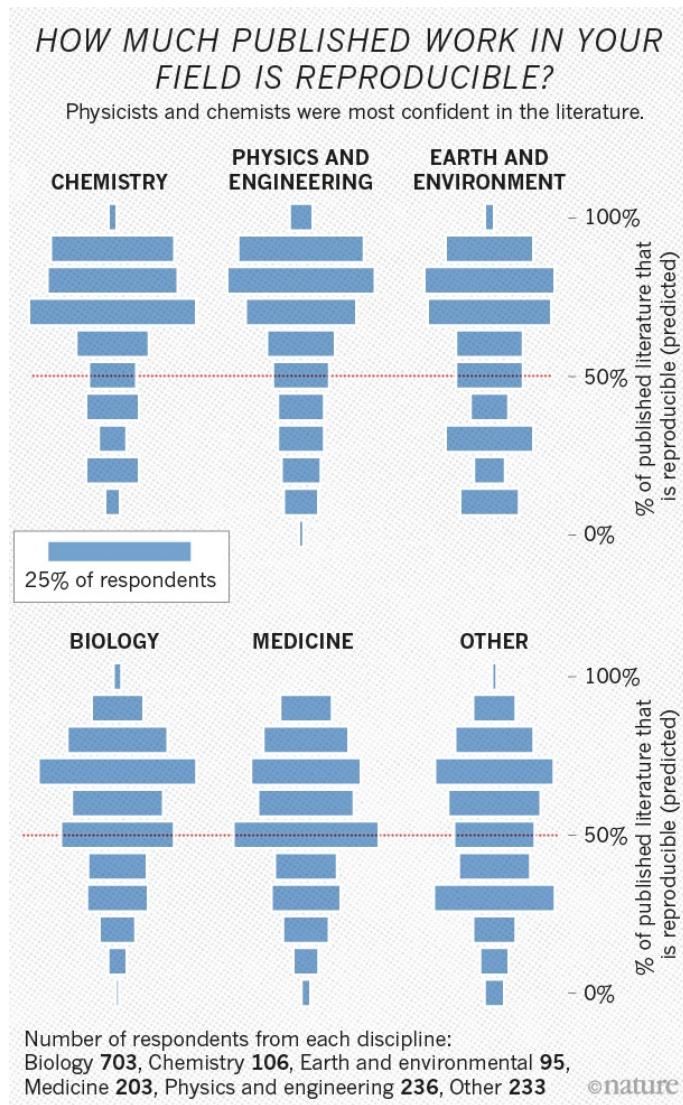


Overview

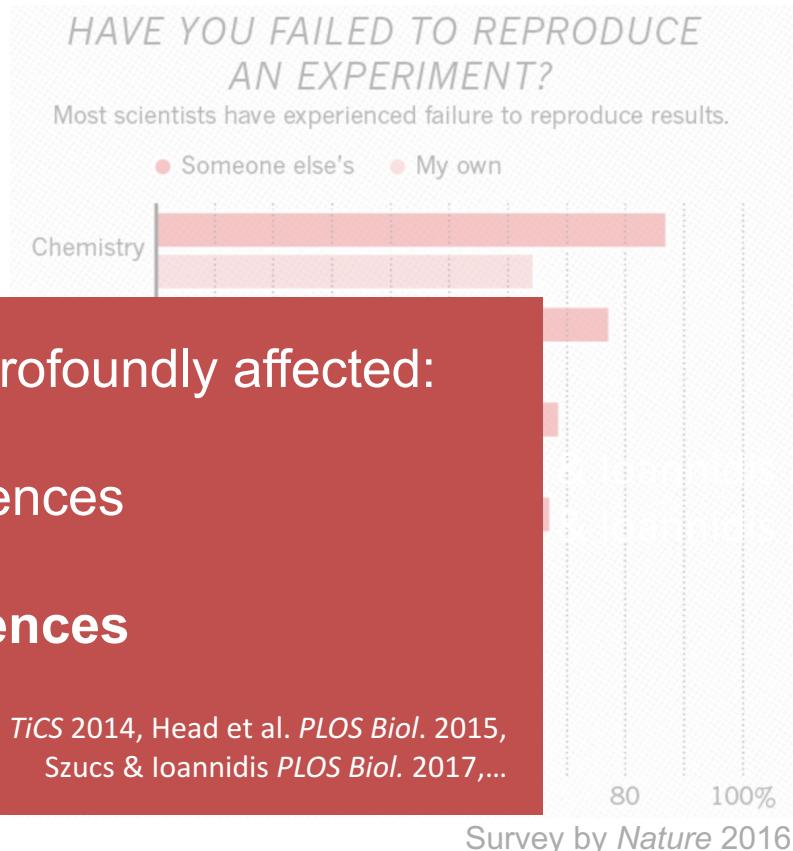
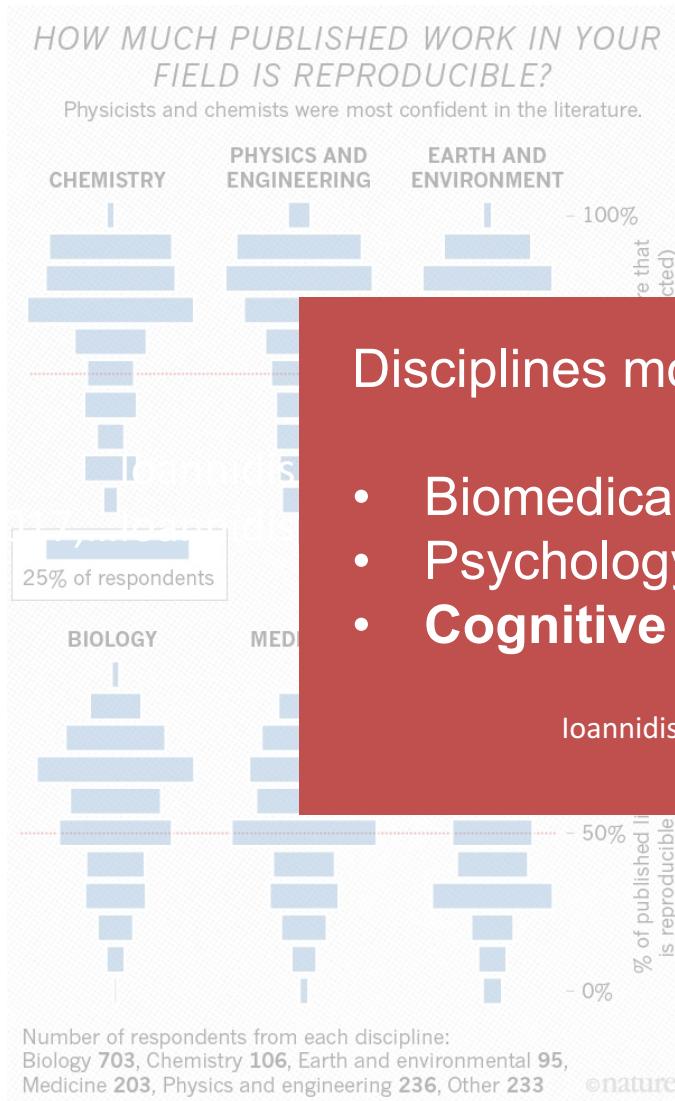
- 1. Motivation**
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Reproducibility crisis



Reproducibility crisis



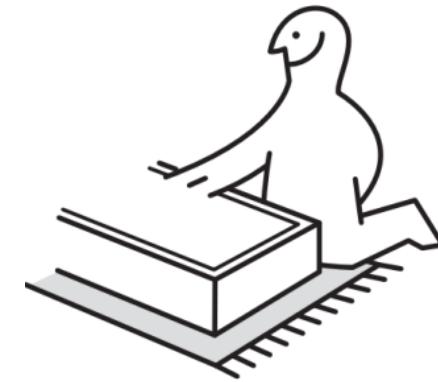
Disciplines most profoundly affected:

- Biomedical sciences
- Psychology
- **Cognitive Sciences**

Ioannidis et al. *TiCS* 2014, Head et al. *PLOS Biol.* 2015,
Szucs & Ioannidis *PLOS Biol.* 2017, ...

Reproducibility crisis in Cognitive Sciences

- Cognitive biases
 - IKEA-effect
 - Texas sharp-shooter effect
- Bad research practices
 - P-hacking
 - HARKing
 - File-drawer effect
- Limitations of methodology
 - Underpowered studies
 - “Narrow” experimental designs



Reproducibility crisis in Cognitive Sciences

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 - “Narrow” experimental designs



Aims of cognitive neuroscience

Research questions

What are the fundamental aspects of cognition?

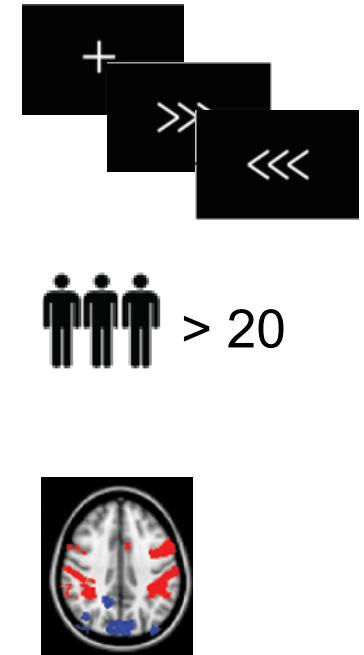
What are the fundamental roles of distinct networks in the brain?

How can cognitive processes be modulated or enhanced?

broad

narrow

Standard approach



Aims of cognitive neuroscience

Human-brain mapping

- Over-specified inferences about functional-anatomical mappings
- Inflated test statistics
(Westfall et al. *Wellcome Open Research* 2017)

Biomarker discovery

- Which exact task conditions will be sensitive to certain patient group?
(Sprooten et al. *Human Brain Mapping* 2017)

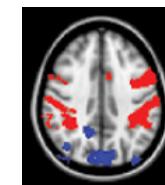
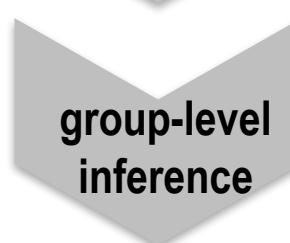
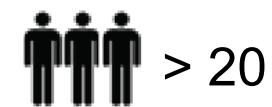
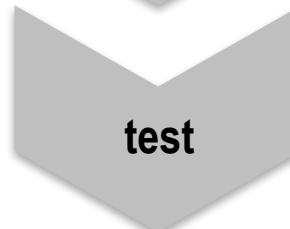
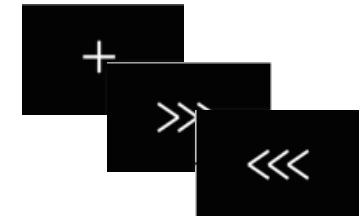
Non-invasive brain stimulation

- Many *free* parameters, confusion surrounding efficacy

broad

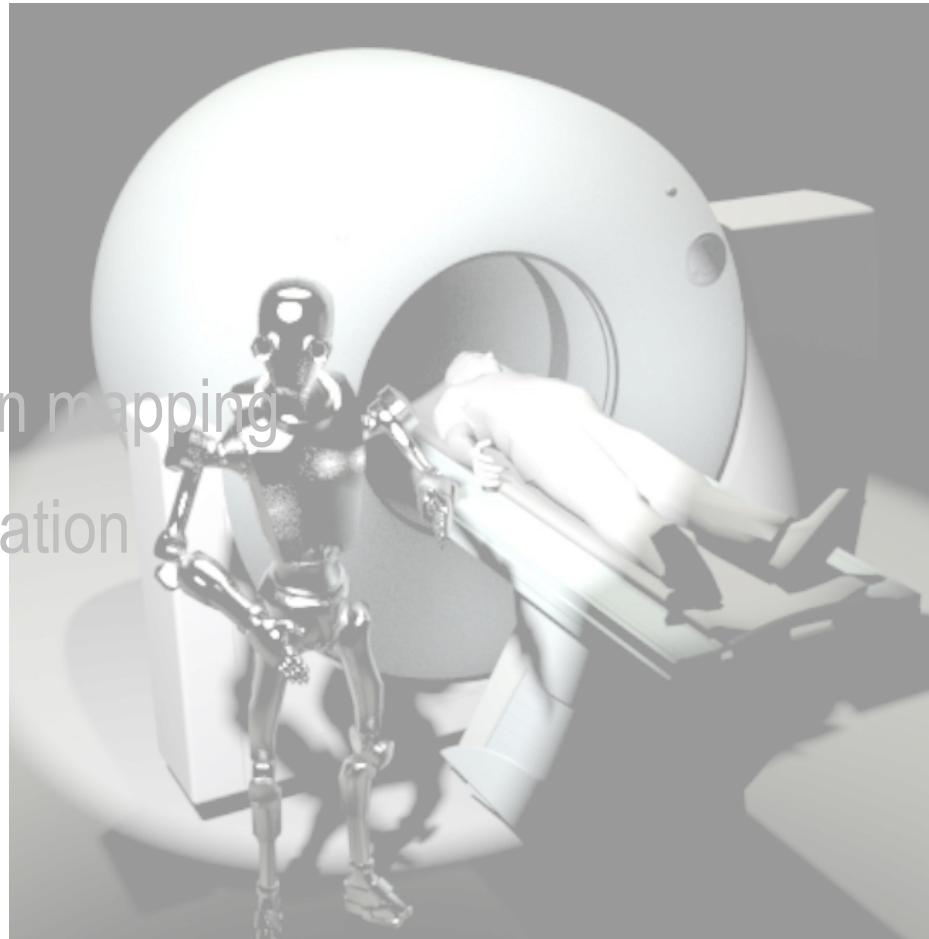
narrow

Standard approach



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The framework

neuroadaptive paradigms

open-loop

- stimuli manually adapted to subject
- subject modulates brain response
e.g. neurofeedback,
communication with vegetative state patients

informed open-loop

- stimuli is triggered by brain state

closed-loop

machine learning

supervised

- ##### passive learning
- BCIs
 - advanced neurofeedback
 - neural selectivity

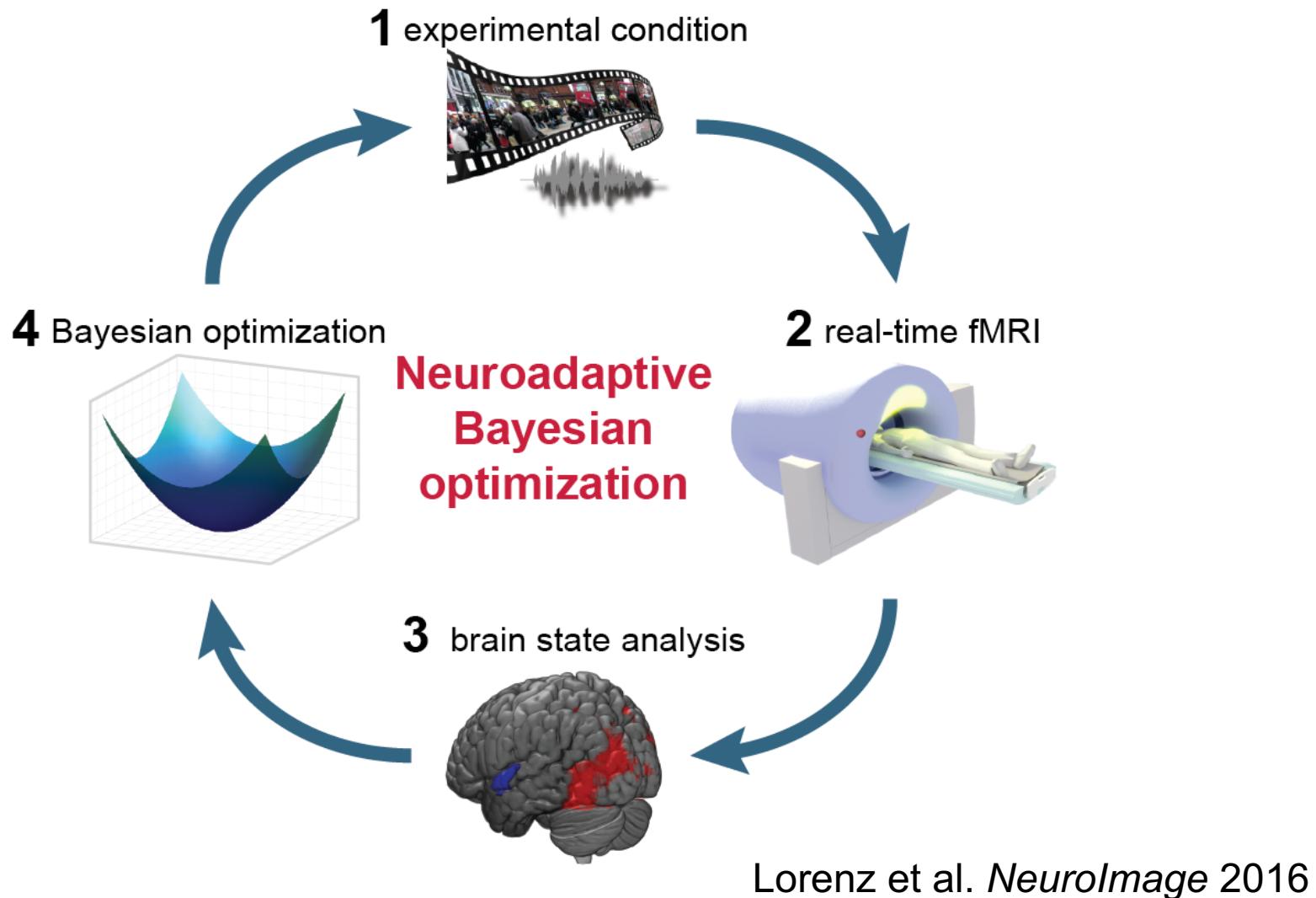
active learning

- tuning curve estimation

Bayesian optimization

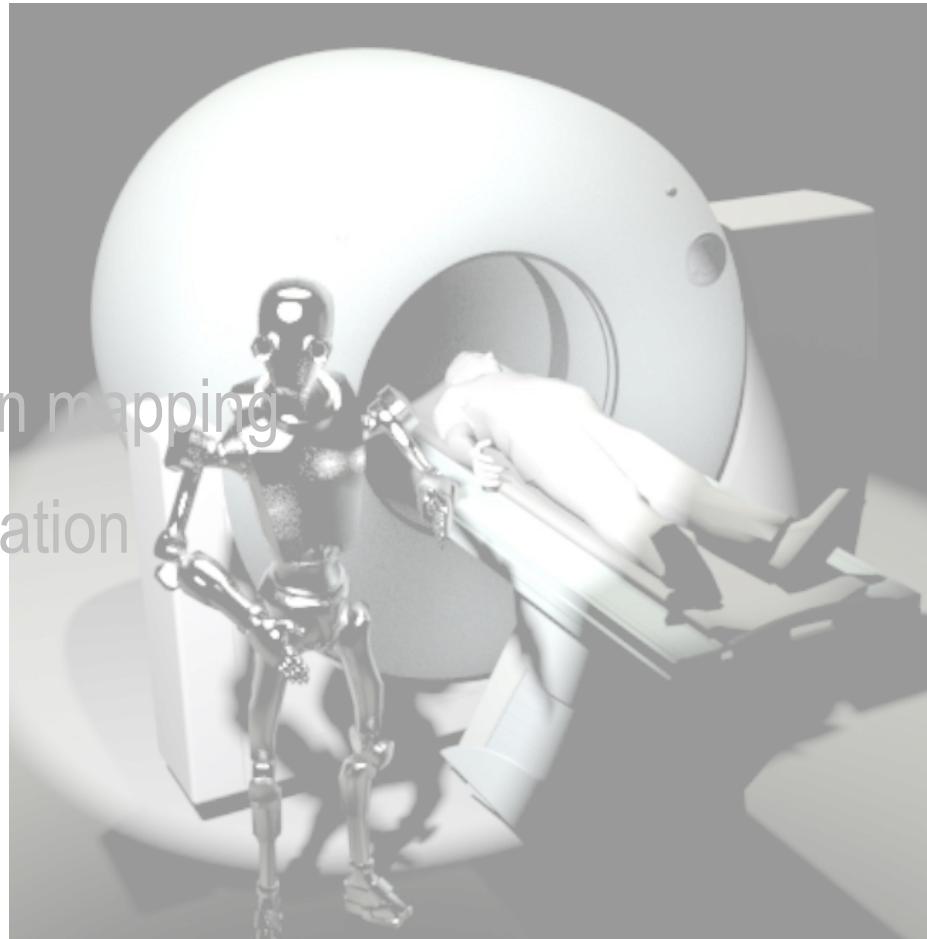
Lorenz et al. *Trends in Cognitive Sciences* 2017

“The Automatic Neuroscientist”

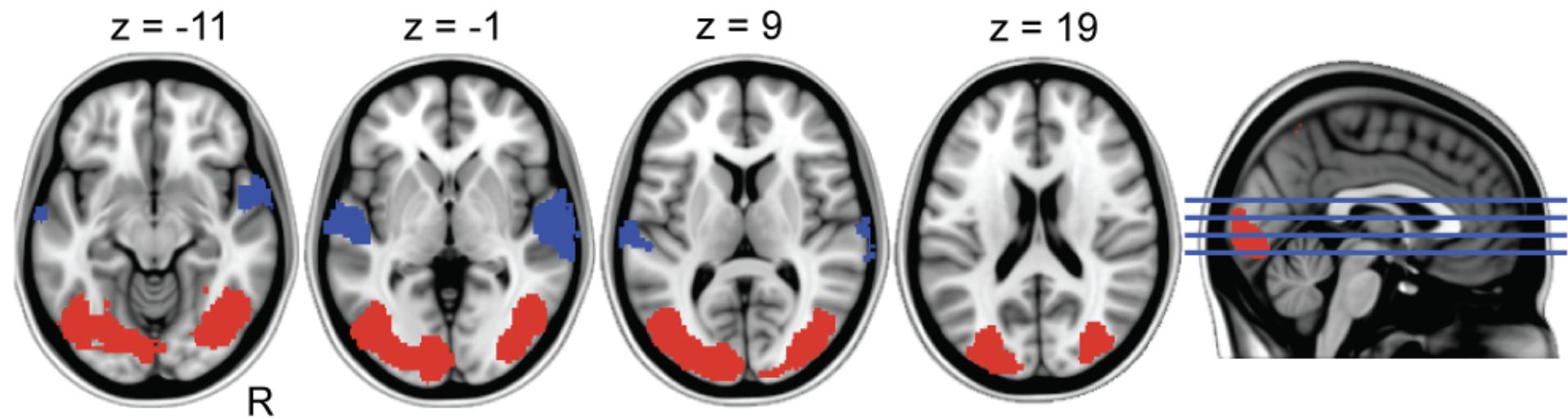


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Target brain state

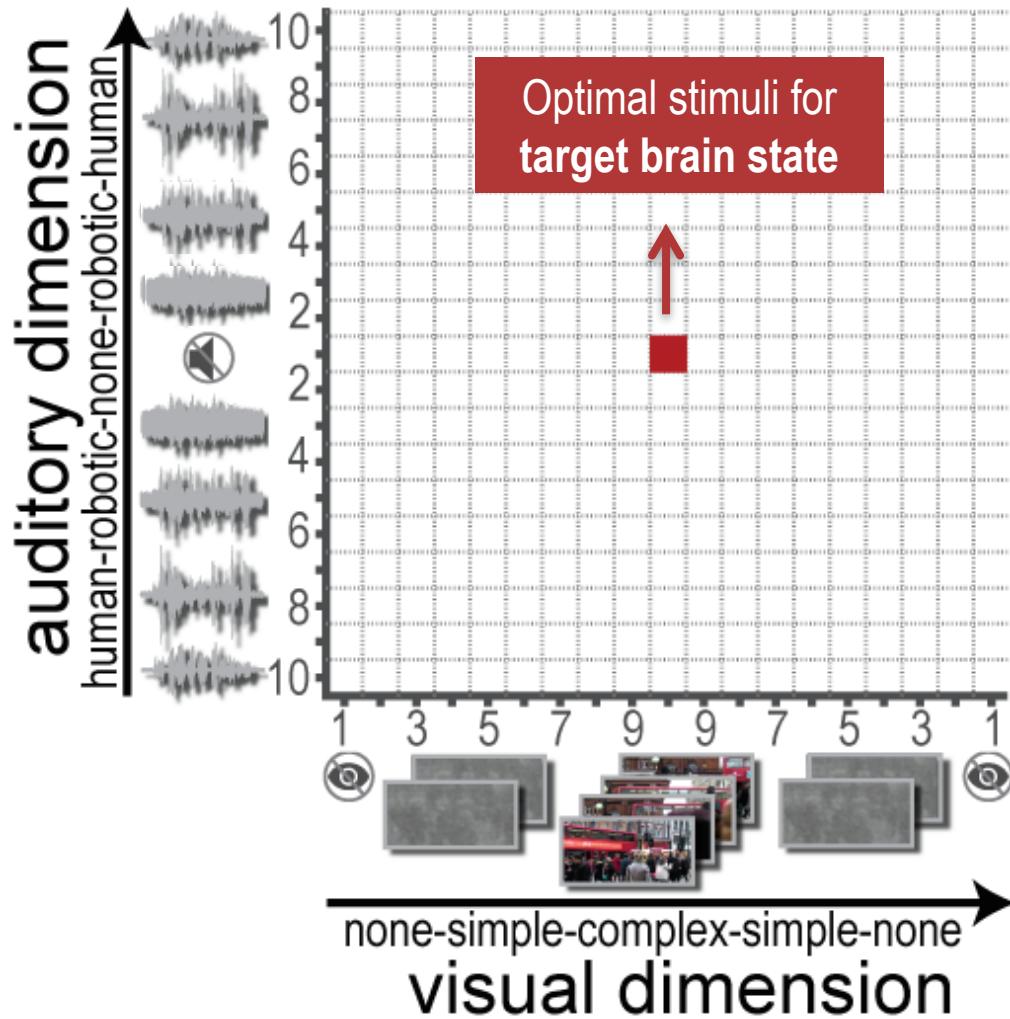


lateral occipital cortex activity \uparrow

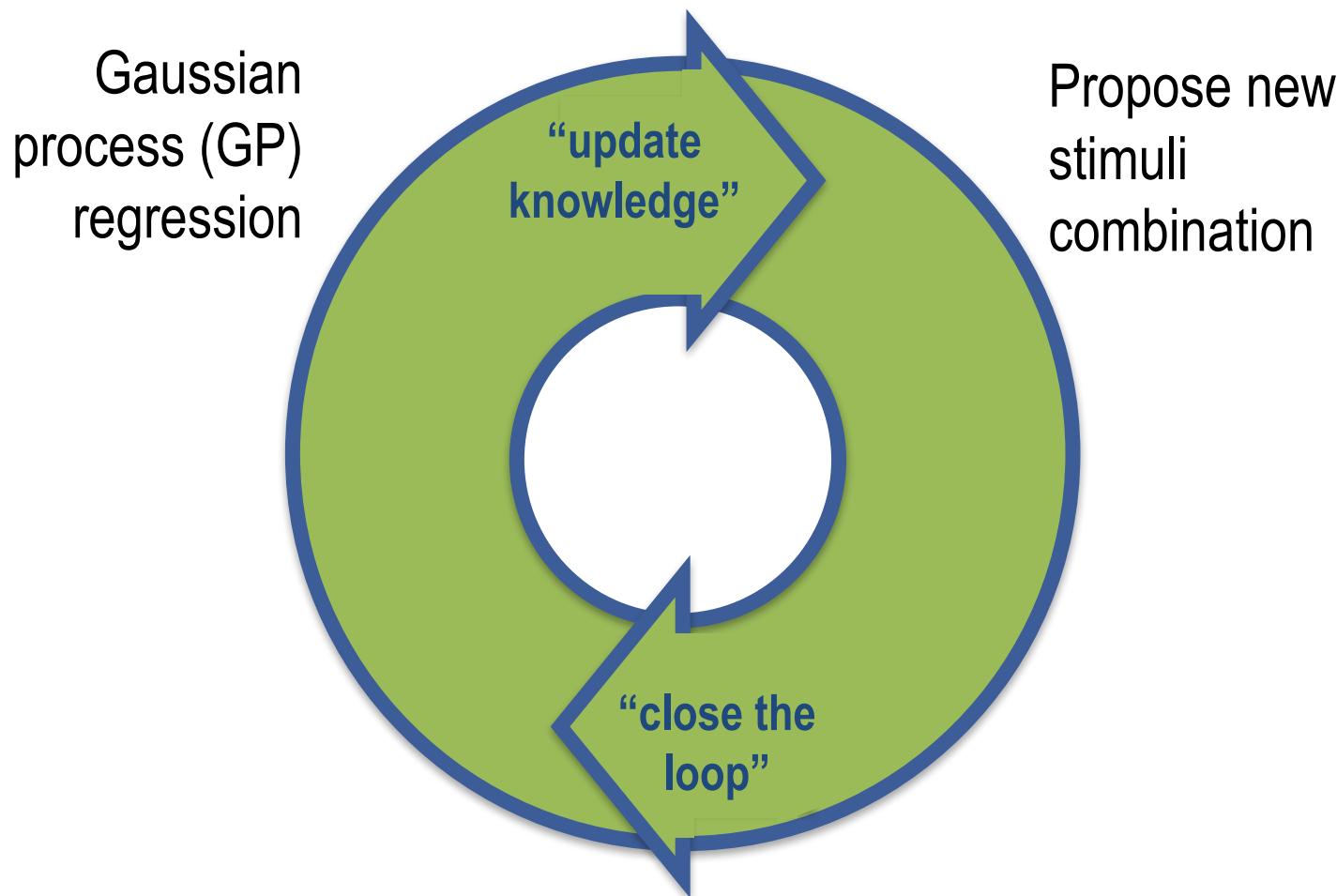
superior temporal cortex activity \downarrow

masks derived from
Braga et al. *NeuroImage* 2013

Experiment space

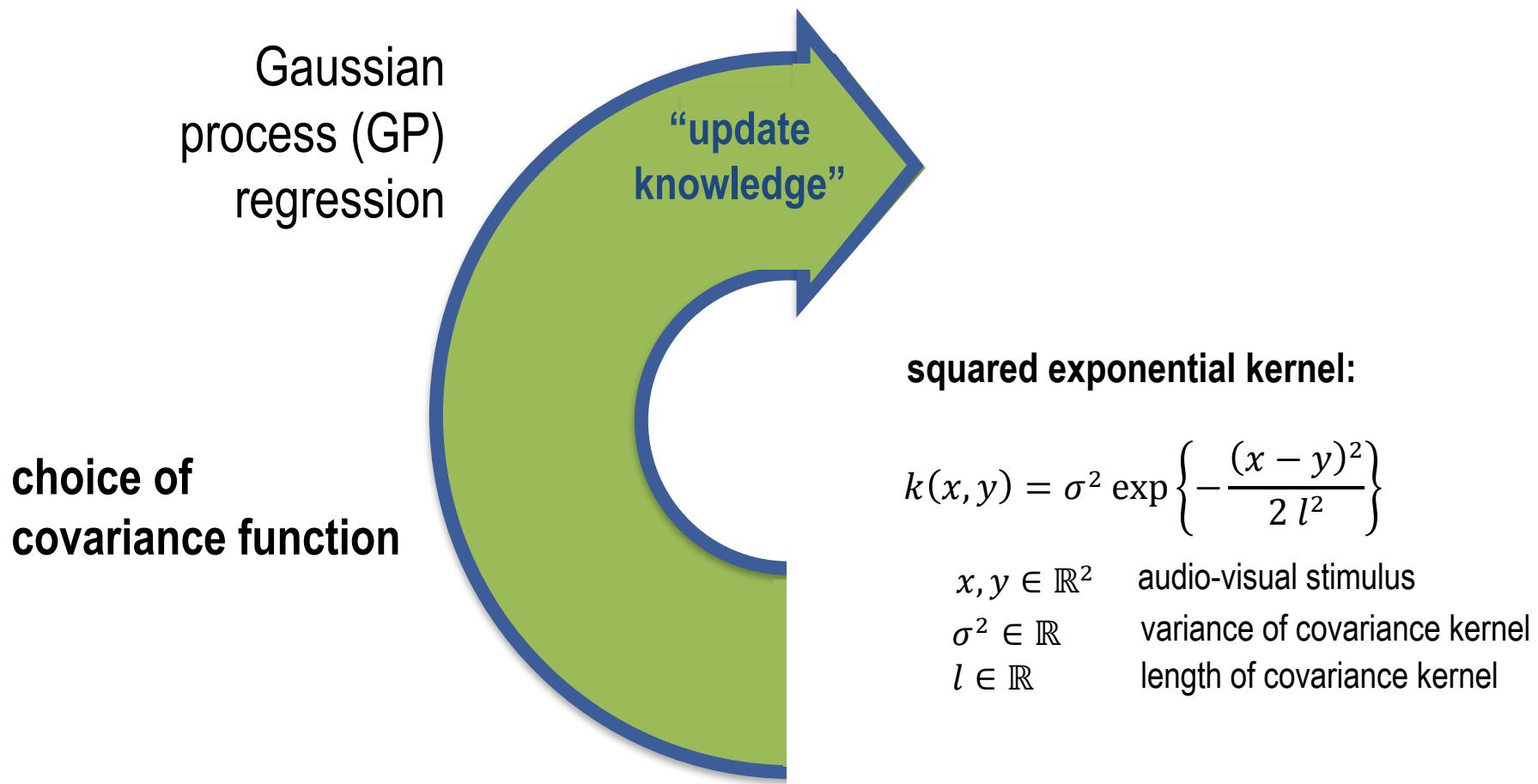


Bayesian optimization



Rasmussen & Williams 2006
Brochu et al. *arXiv* 2010

Bayesian optimization



Rasmussen & Williams 2006
Brochu et al. arXiv 2010

Bayesian optimization

Expected improvement acquisition function:

$$EI(x) = (m(x) - f_{max})q(z) + var(x)p(z)$$

$m(x)$: predicted mean

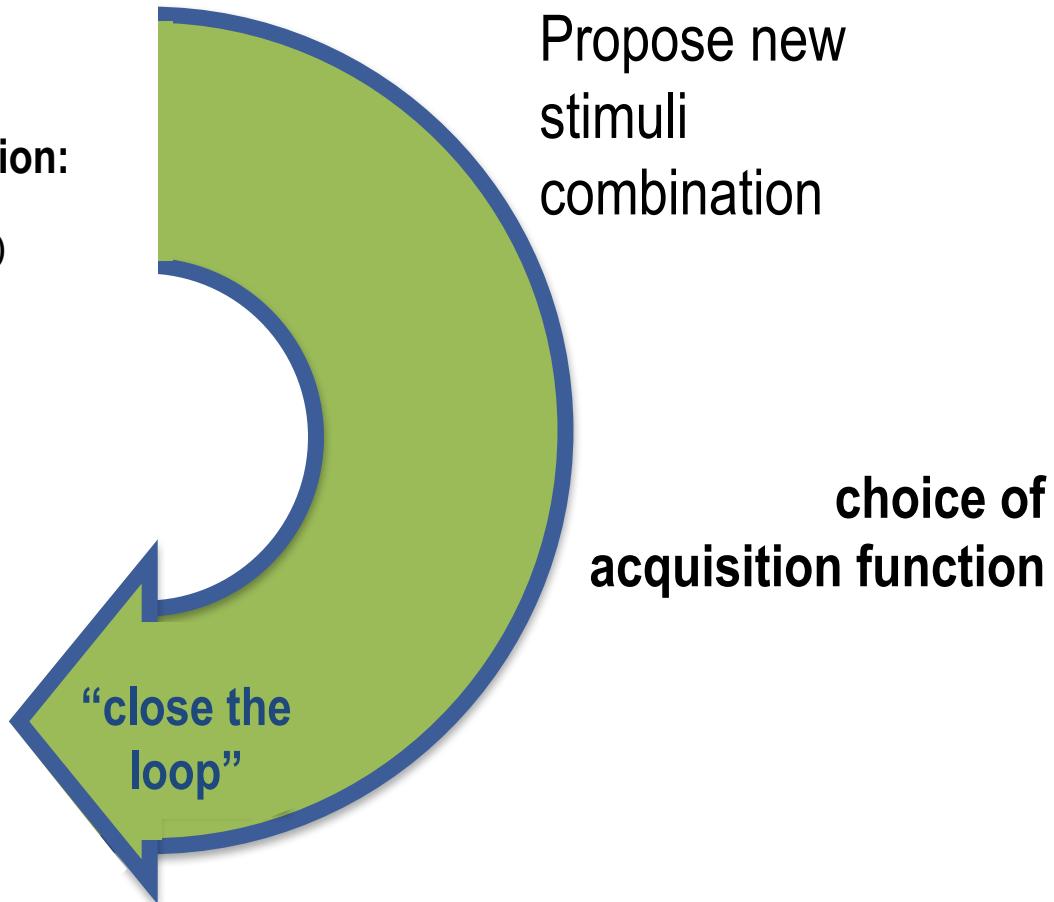
$var(x)$: predicted variance

f_{max} : maximum predicted value

$q(\cdot)$: cumulative distribution function

$p(\cdot)$: probability density function

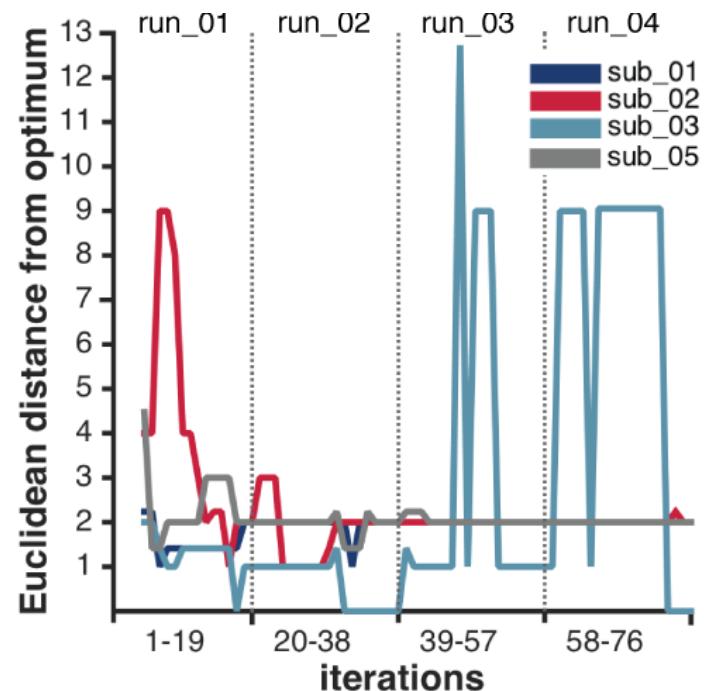
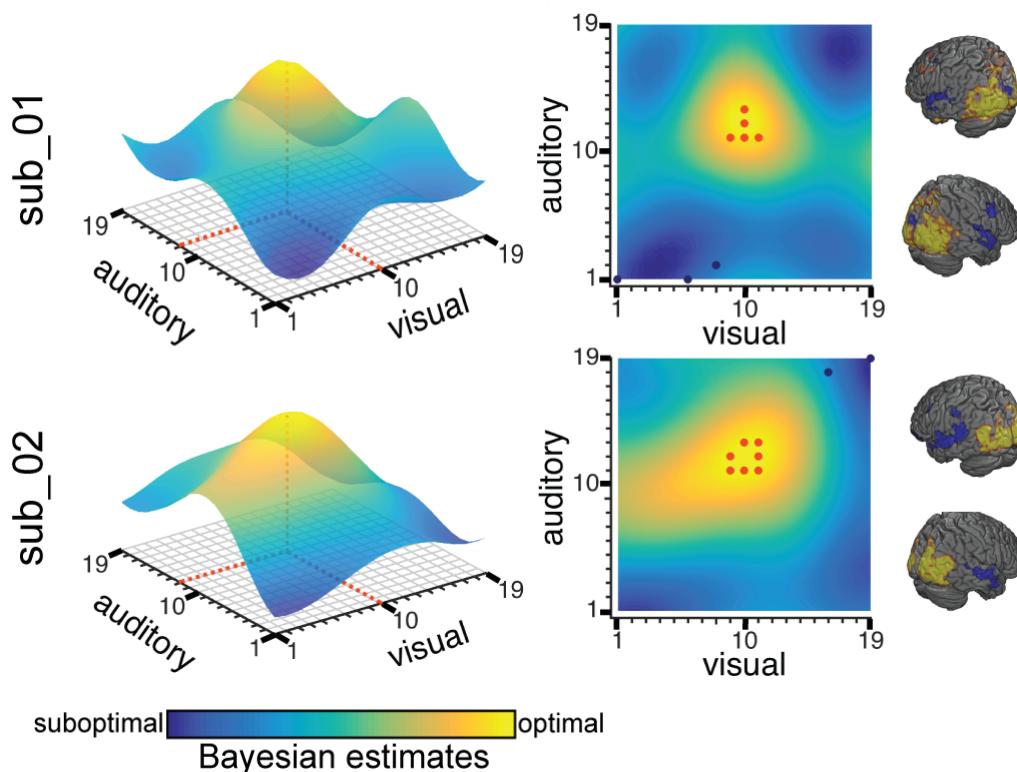
$$z = \frac{m(x) - f_{max}}{var(x)}$$



Rasmussen & Williams 2006
Brochu et al. arXiv 2010

Results

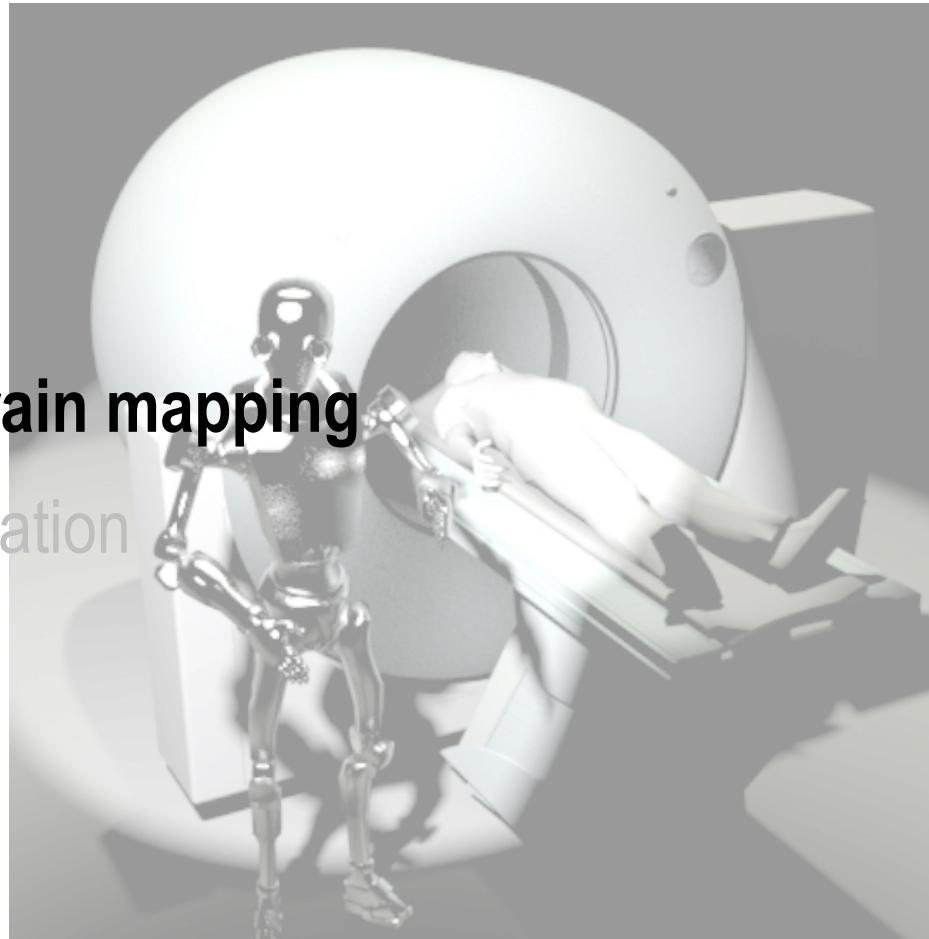
3 people = 5



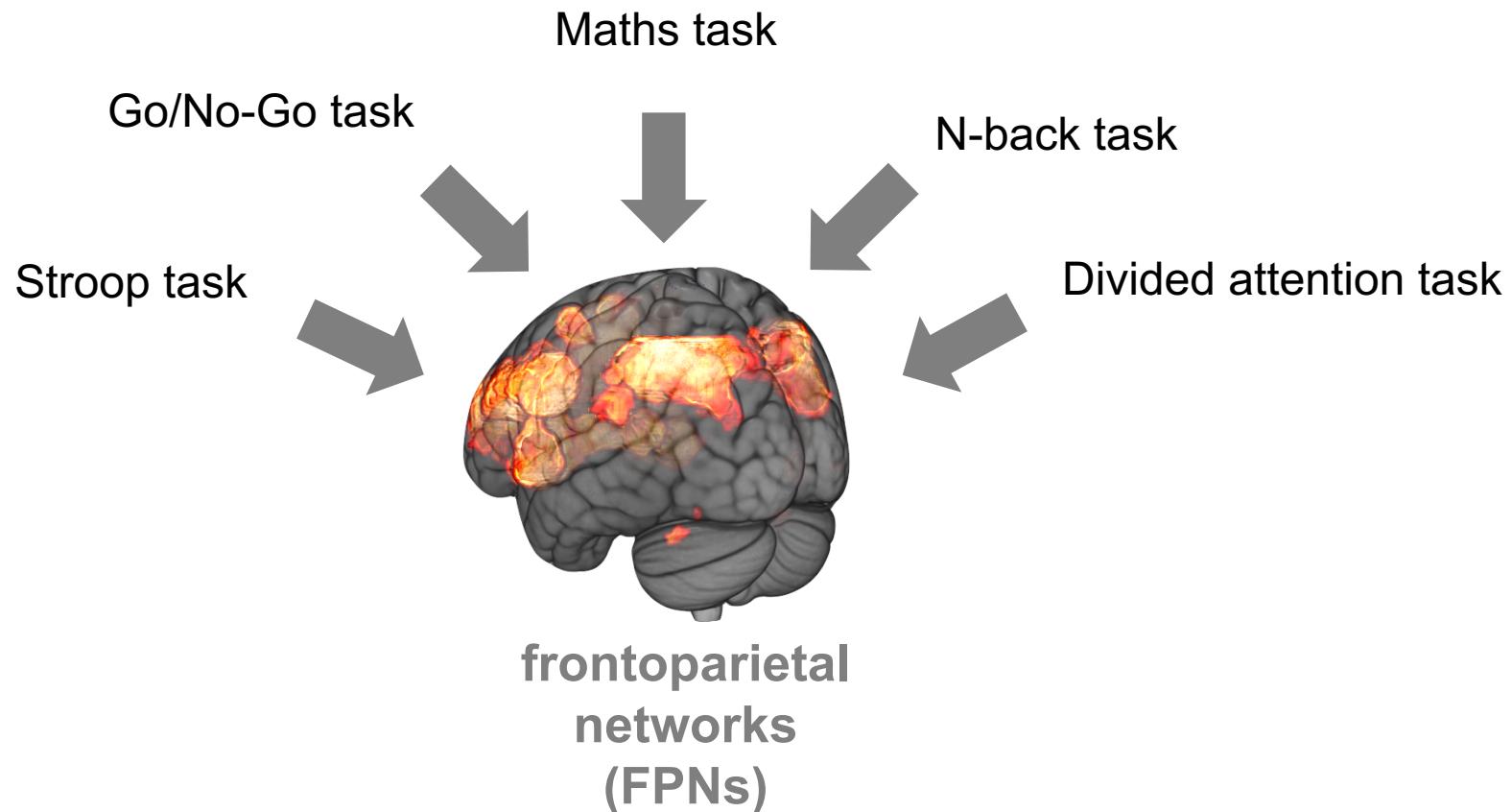
Lorenz et al. *NeuroImage* 2016

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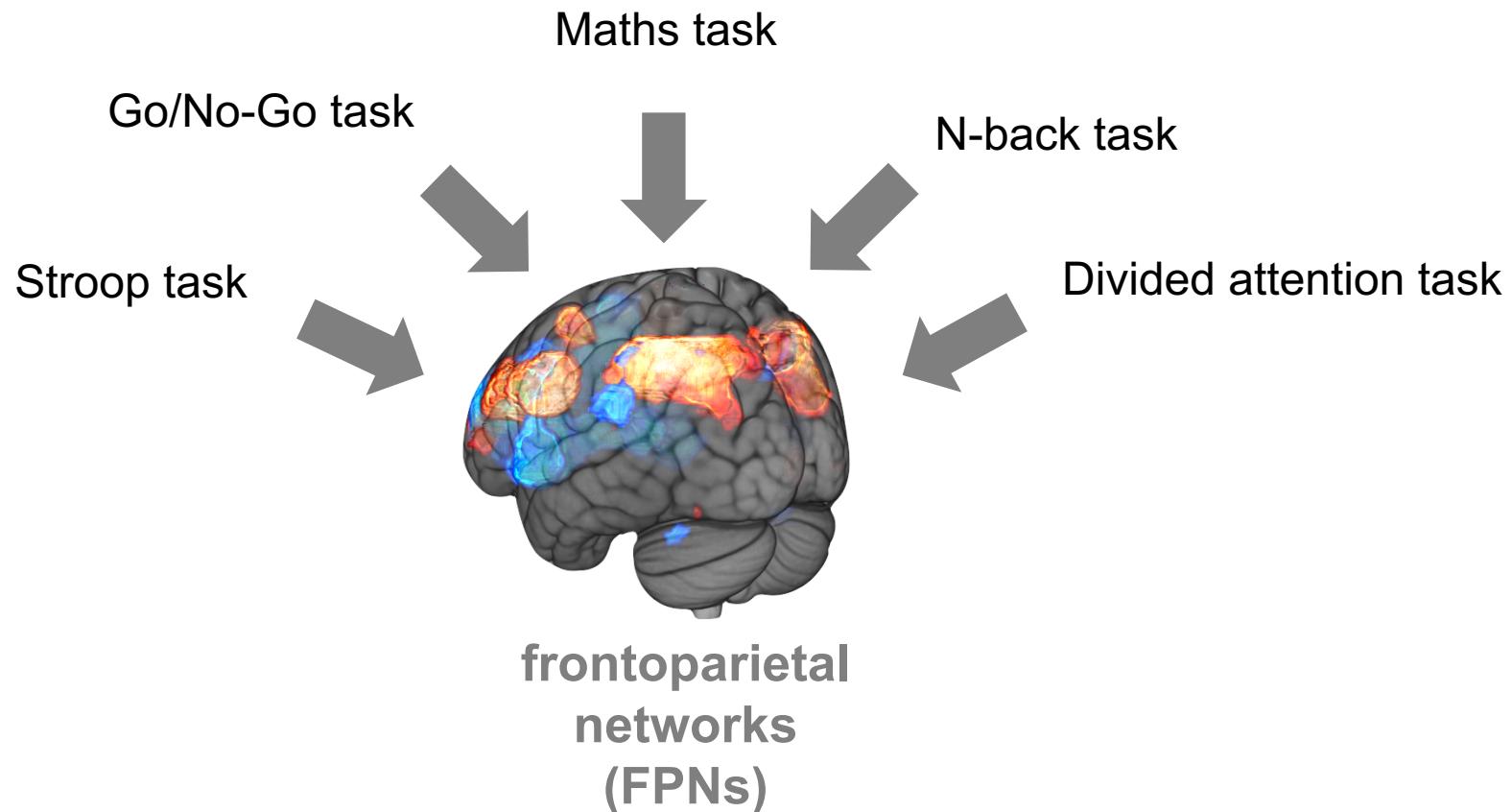


Motivation



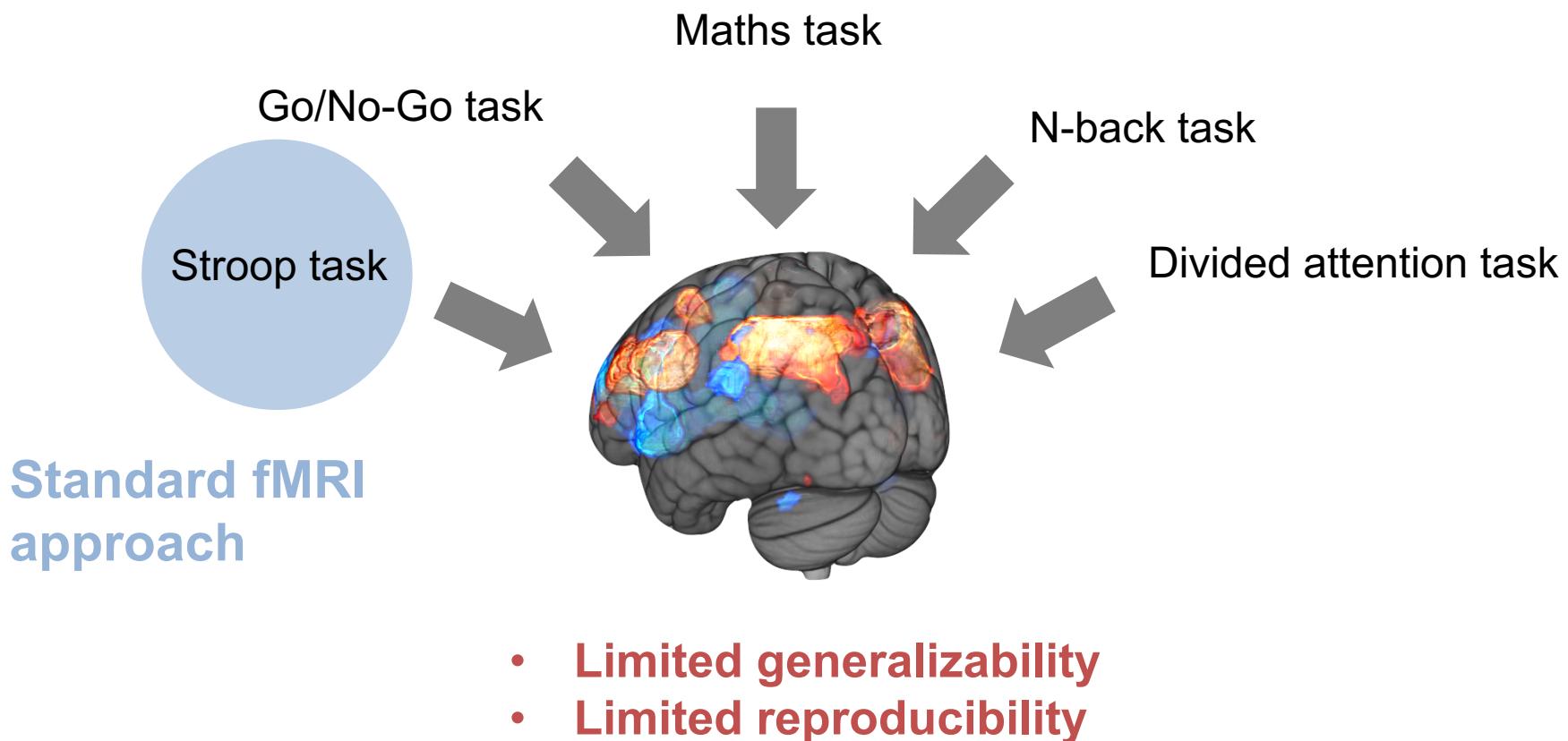
Duncan & Owen *TiNS* 2000
Fedorenko et al. *PNAS* 2013

Motivation



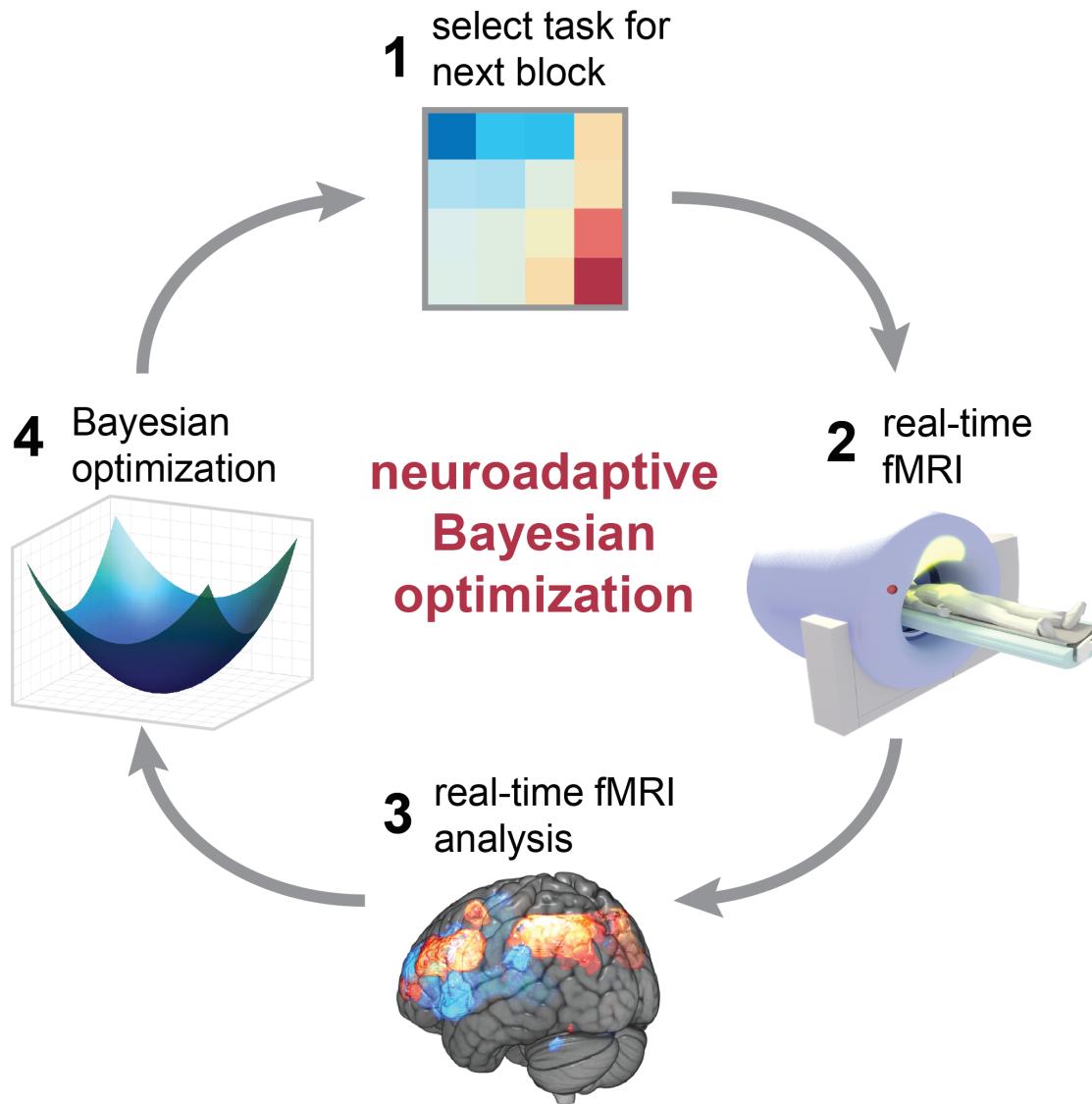
Hampshire et al. *Neuron* 2012

Motivation

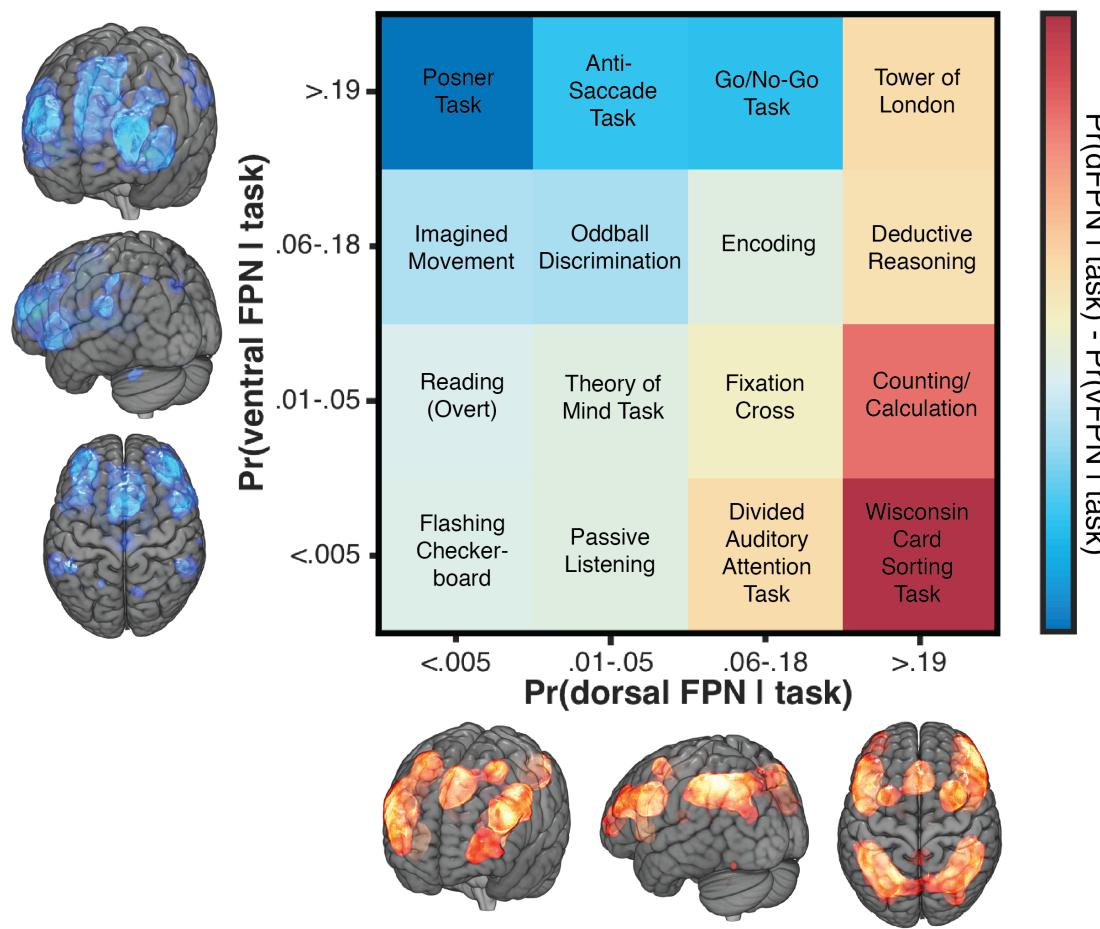
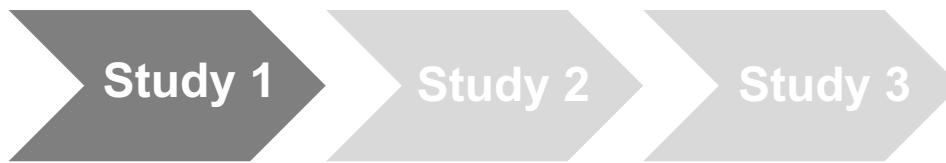


Lorenz et al. *Trends in Cognitive Sciences* 2017
Westfall et al. *Wellcome Open Research* 2017

Searching across cognitive tasks

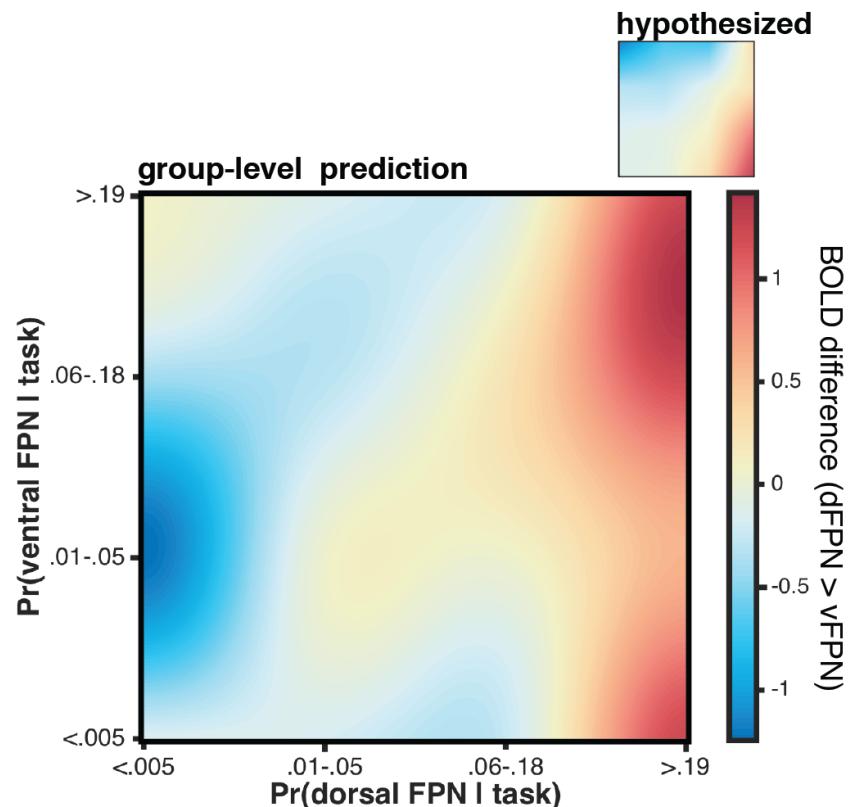
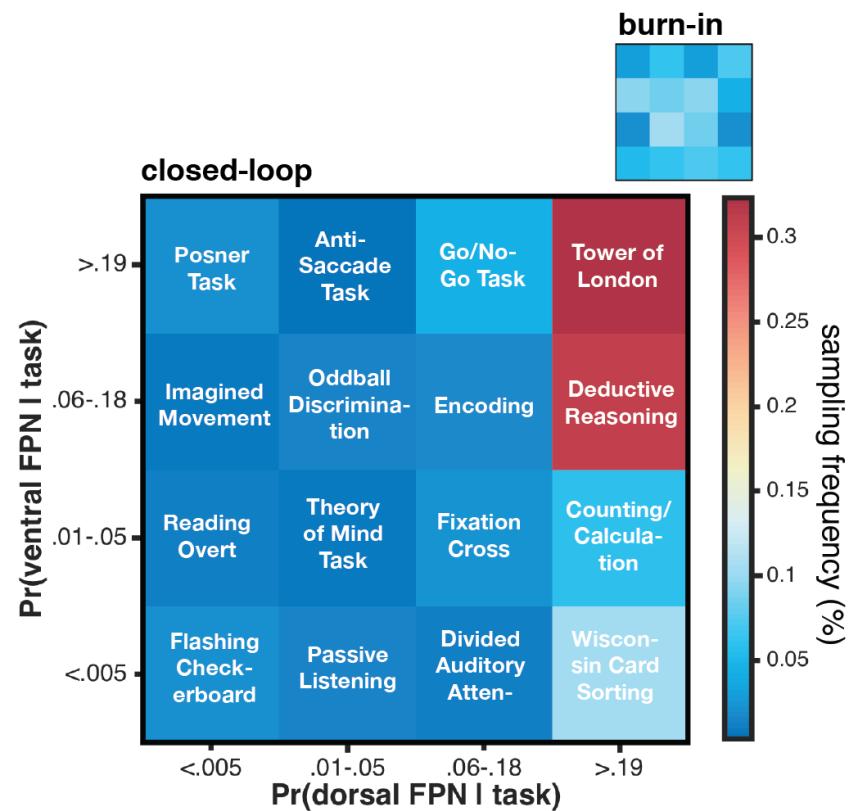
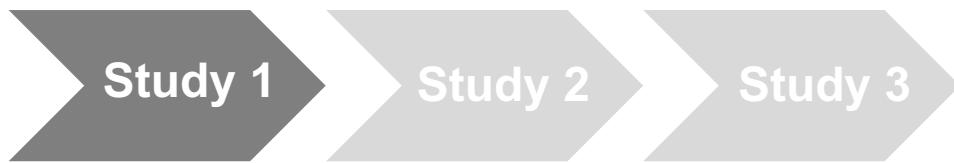


Task space based on meta-analysis



Yeo et al. *Cerebral Cortex* 2015

Find optimal tasks



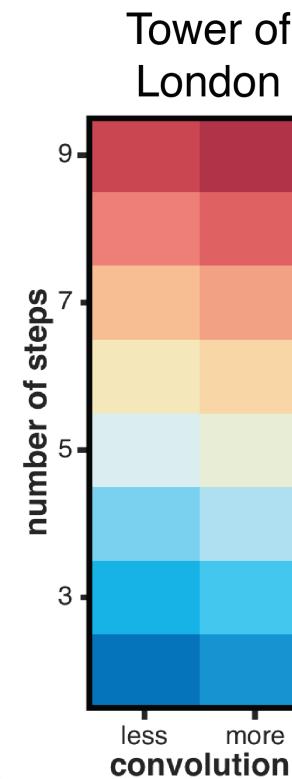
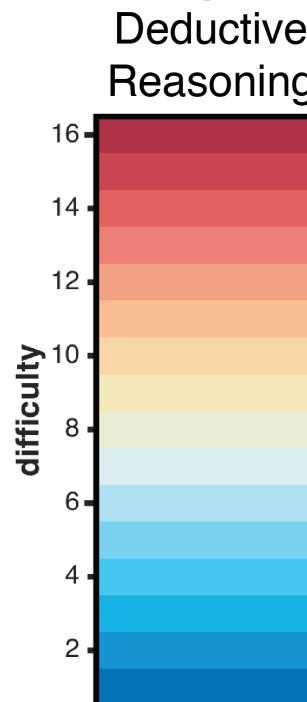
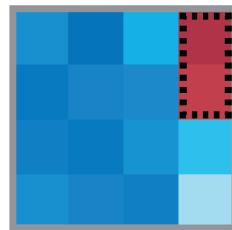
Tower of London & Deductive Reasoning tasks maximally dissociate FPNs

Zoom in task space and fine-tune tasks

Study 1

Study 2

Study 3

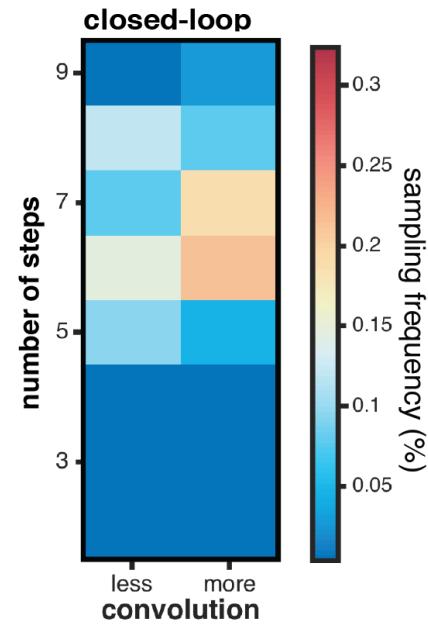
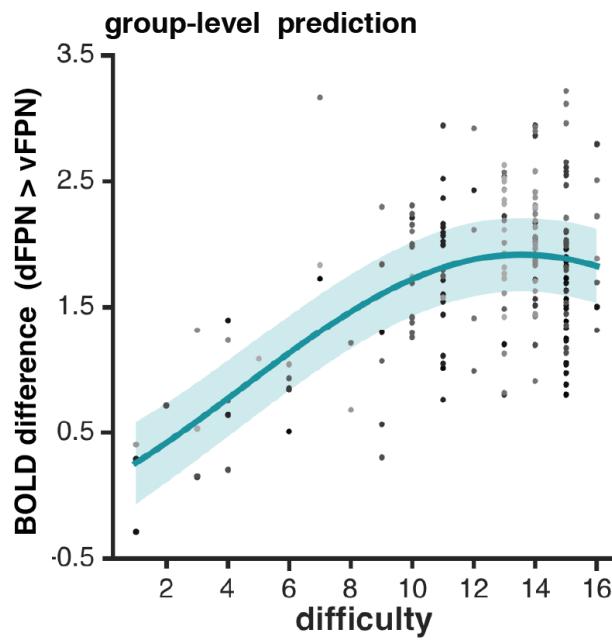
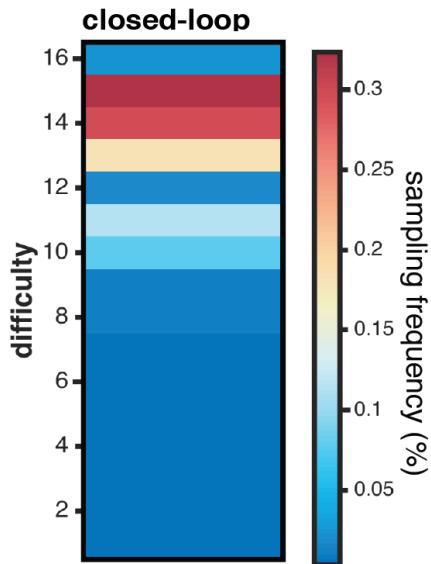


Find optimal task parameters

Study 1

Study 2

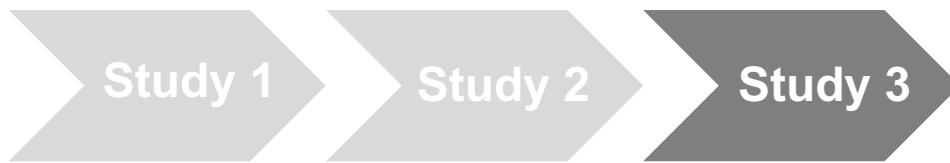
Study 3



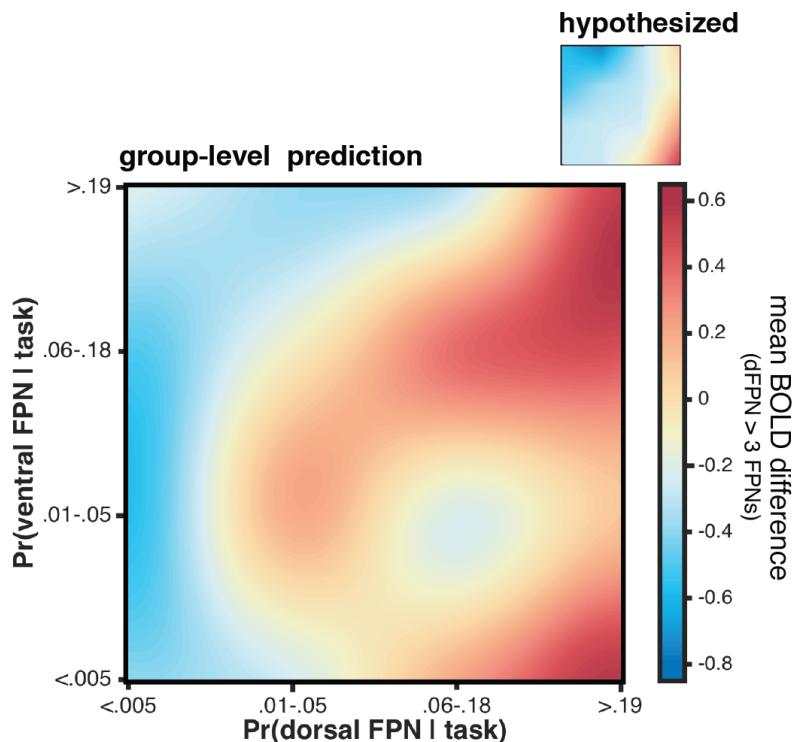
Deductive Reasoning

Tower of London

Find unique functional profile

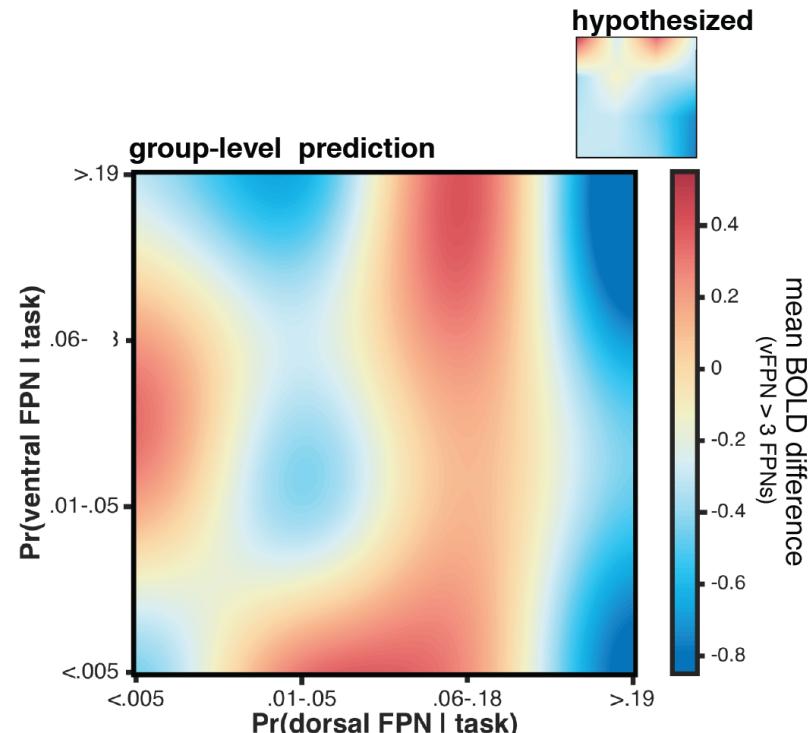


dorsal FPN > 3 other FPNs



Tower of London, Deductive Reasoning, Encoding
& Wisconsin Card Sorting tasks

ventral FPN > 3 other FPNs



Go/No-Go, Divided Auditory Attention,
Passive Listening & Reading tasks

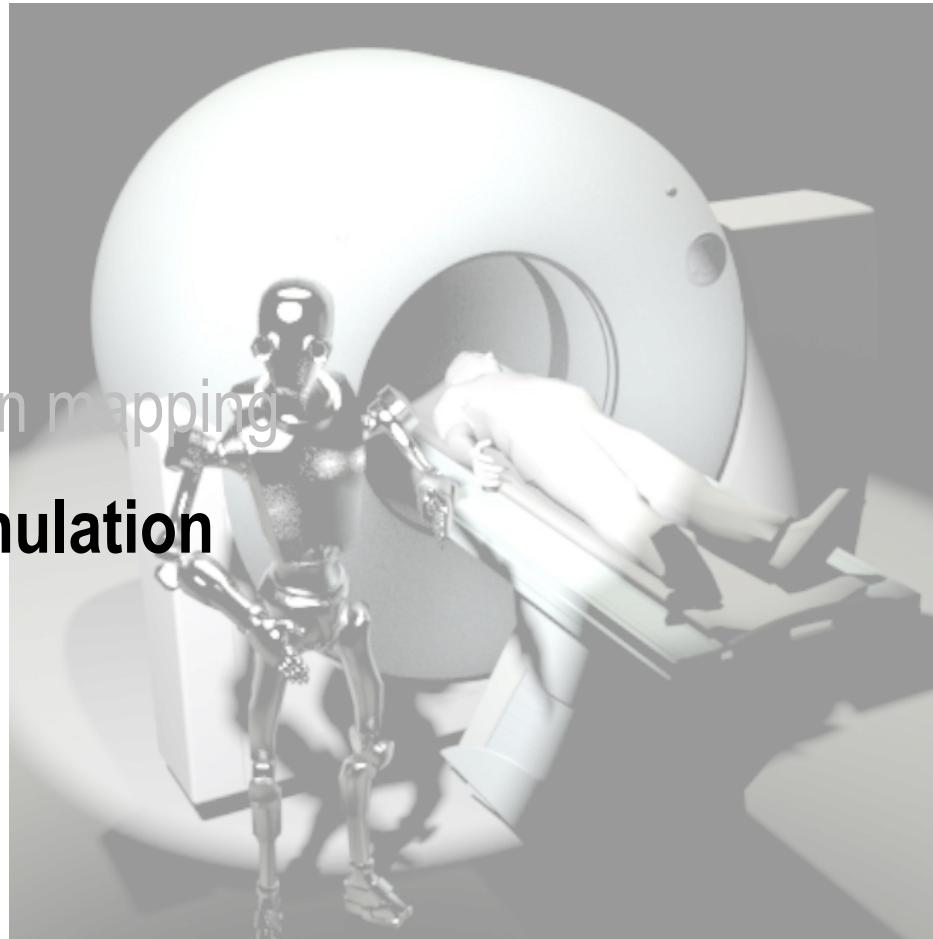
Summary

- High inter-subject reliability
- Functional profile across many tasks is unique to each FPN
- Set of optimal tasks only partially corresponds to meta-analysis and previous functional labels
- Neurally-derived cognitive taxonomy needed
- **Powerful synergy between neuroadaptive Bayesian optimization and meta-analyses**

Lorenz et al. *under revision* (bioRxiv:128678)

Overview

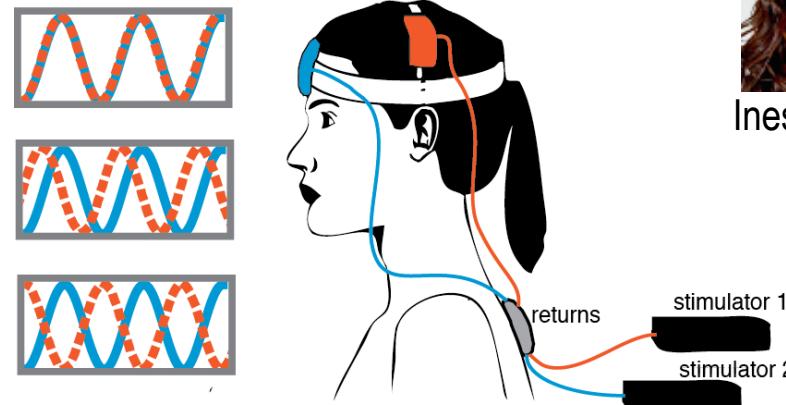
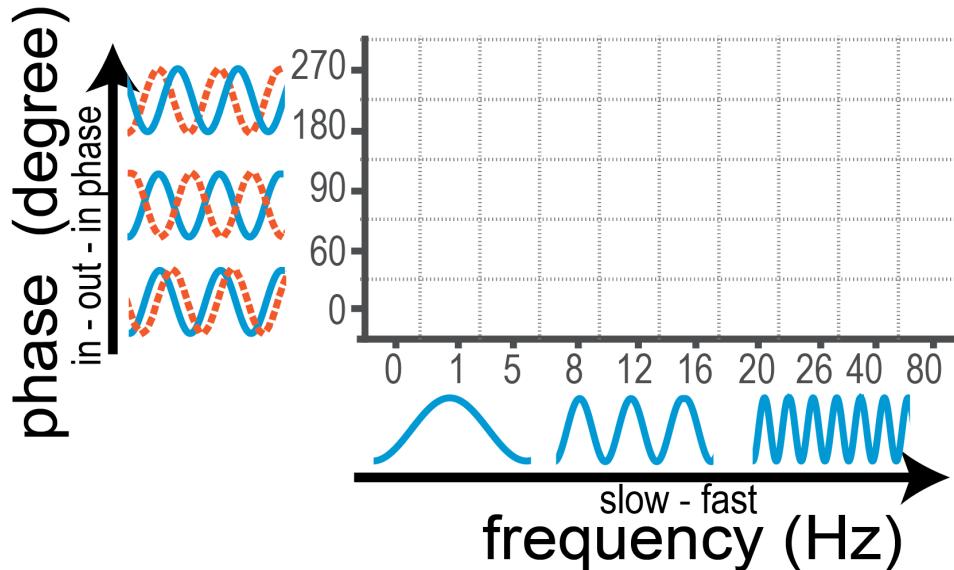
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Transcranial alternating current stimulation (tACS)

■ Status Quo

- Ad hoc definition of frequency and phase
- Cohort testing

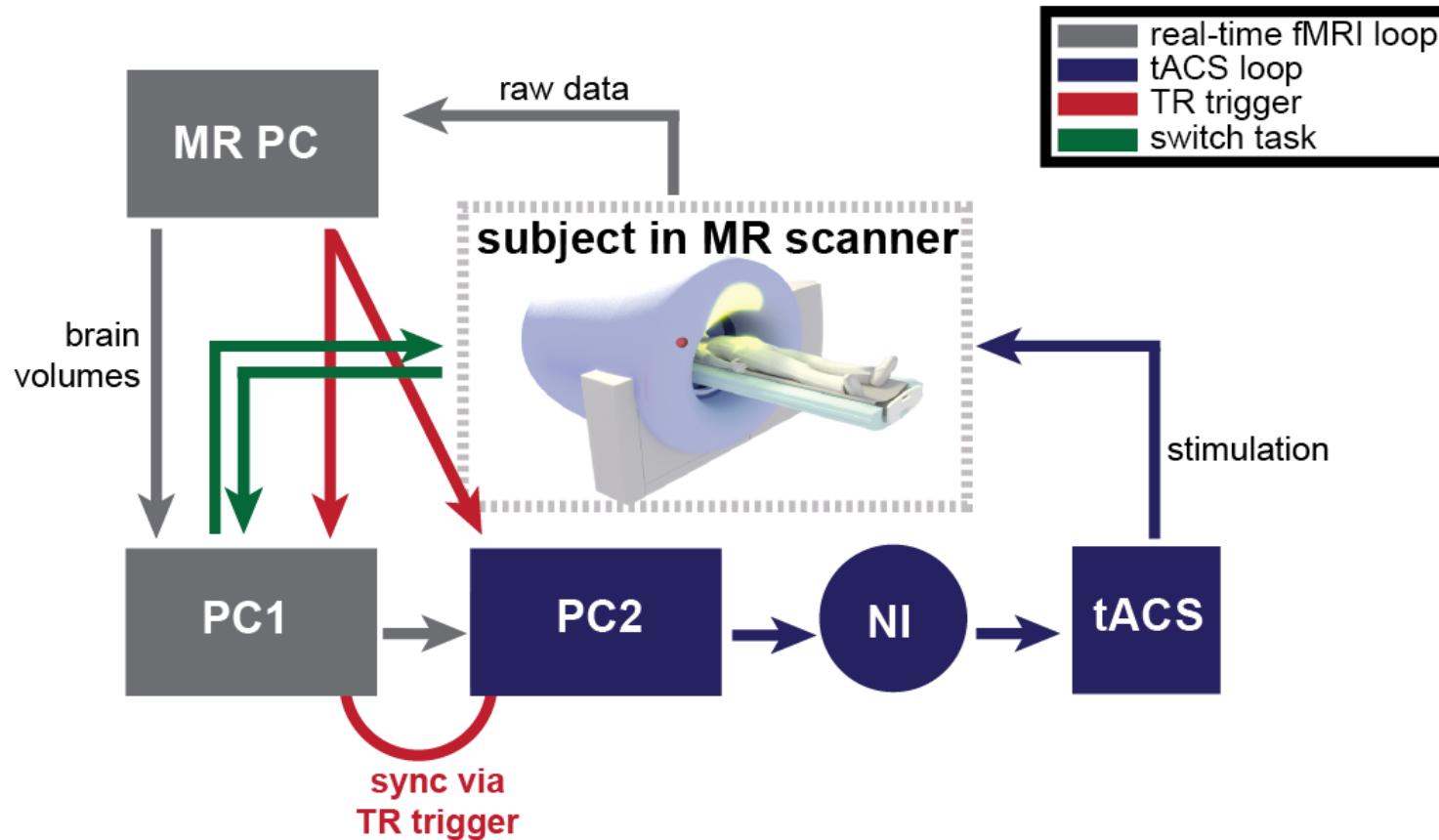


Ines Violante

■ Limitation

1. How to choose frequency and phase?
2. Stimulation parameters may vary due to anatomy or pathology

Concurrent real-time fMRI/tACS

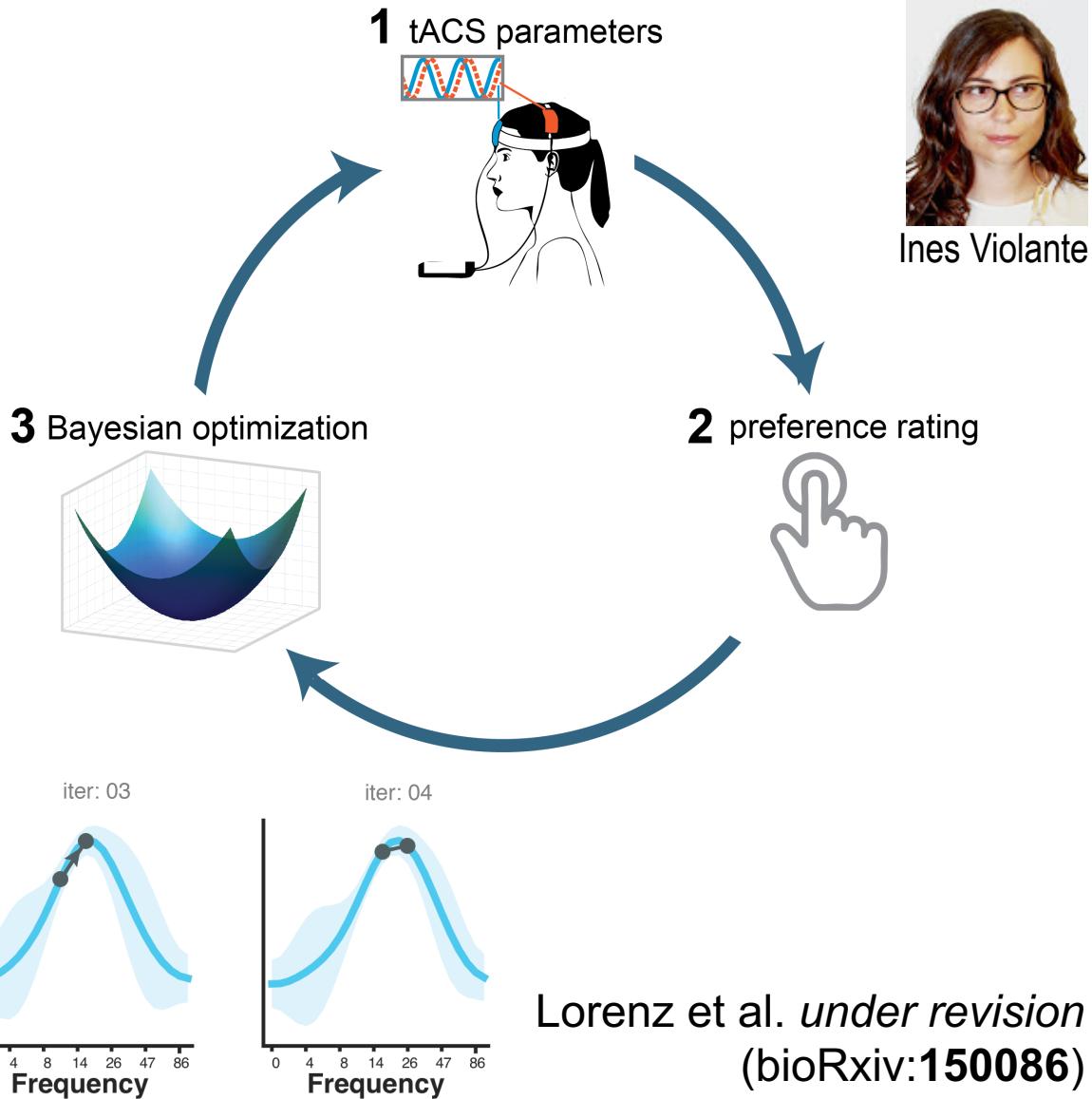


Ines Violante

Lorenz et al. *PRNI* 2016
Lorenz et al. *in preparation*

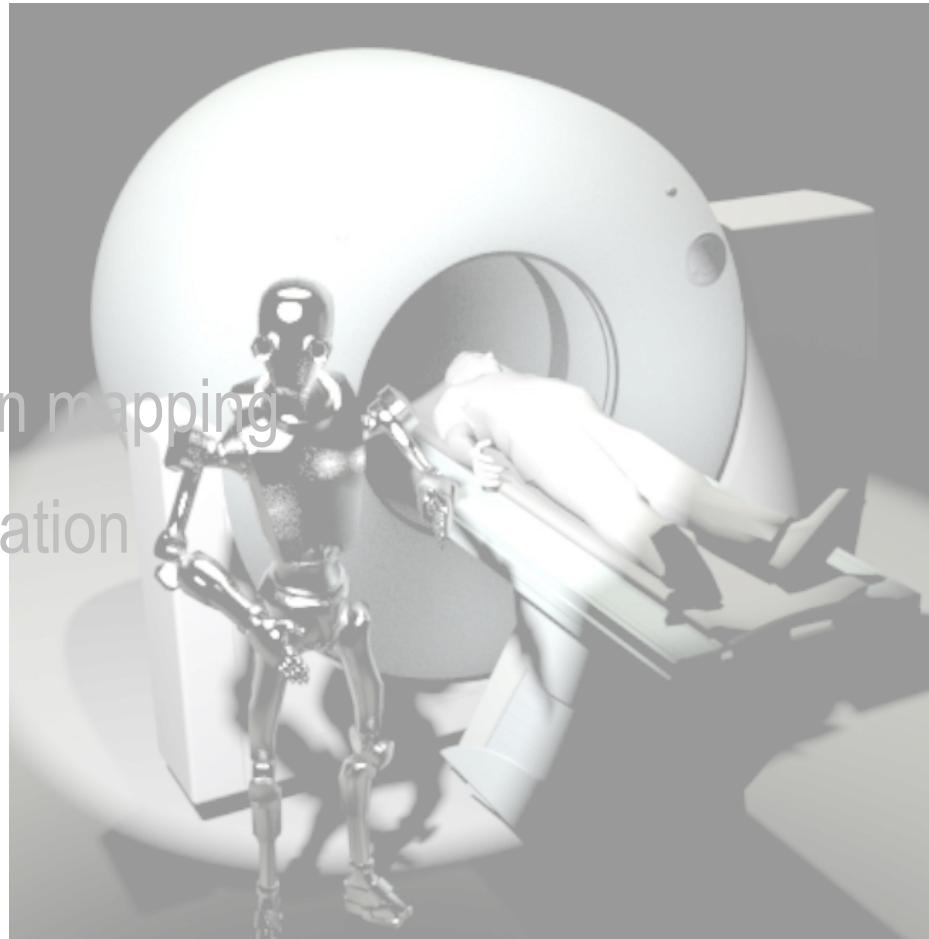
Phosphene perception

- *Phosphenes* = flash-like percepts during brain stimulation
- Major experimental challenge (neuromodulation, alertness)



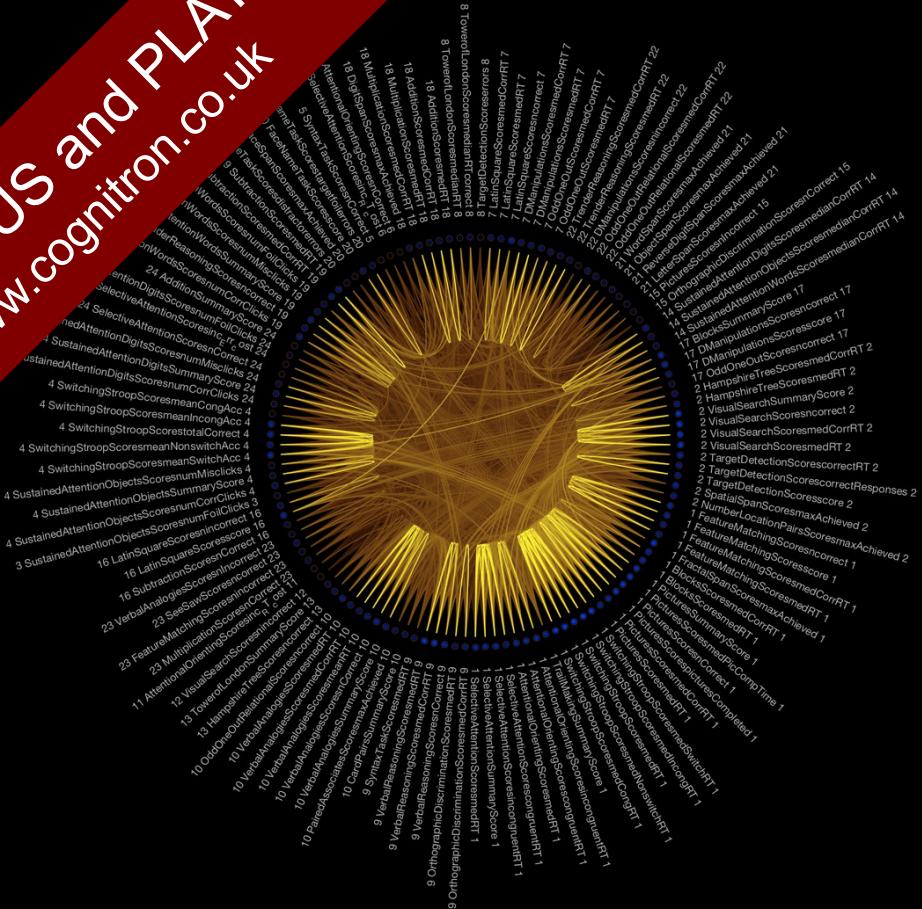
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COGNITRON

AI-web server to dissect human intelligence



A portrait of a man with long, dark hair pulled back, wearing a black button-down shirt. He is positioned in front of a whiteboard that has various handwritten notes and diagrams, including the words "Fingerprints", "Biography", and "DNA".

Adam
Hampshire

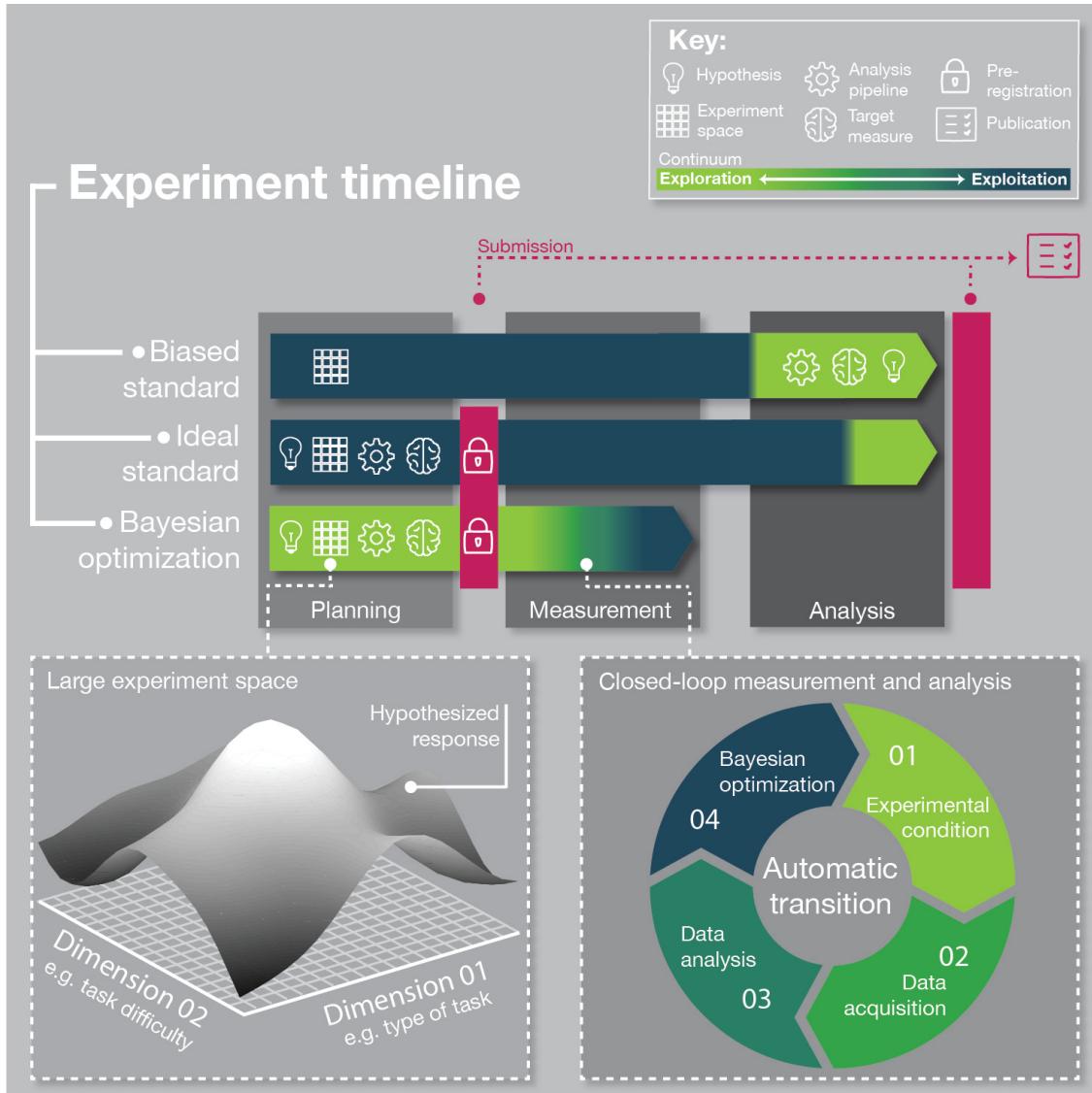
$N > 15,000$

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Implications for improving reproducibility



- More **flexible hypothesis** possible (exploration)
- Improved **specificity & generalizability** of research findings
- Can be combined with **pre-registration**

Lorenz et al. *TiCS* 2017

Future work – need for method development

- Addressing small effect sizes
 - Hierarchical optimization protocol
- Diagnosis: biomarker discovery
 - Novel acquisition functions
- Therapy: tuning to individual patient
 - Statistical inference on objective function/sampling trajectory
- General:
 - Stopping criteria
 - Non-stationarity in time (habituation)

Acknowledgement

Funding

EPSRC

Engineering and Physical Sciences
Research Council

Imperial Biomedical Research Centre

**Imperial College
London**

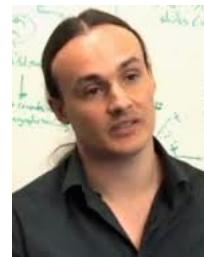
Cognitive, Clinical and Computational
Neuroimaging Laboratory **C³NL**

 **UCL**

Gatsby Computational Neuroscience Unit



Rob



Adam



Ines



Ricardo

**Robert Leech
Adam Hampshire
Ines R. Violante**

Ricardo P. Monti

Resources

- **Code**

- GP regression: <http://github.com/SheffieldML/GPy>
- Acquisition functions: <http://github.com/romylorenz/AcquisitionFunction>

- **Publications**

Lorenz R, Hampshire A, Leech R (2017). **Neuroadaptive Bayesian optimization and hypothesis testing**. *Trends in Cognitive Sciences*, 21(3): 155-167

 Lorenz R, Monti RP, Violante IR, Anagnostopoulos C, Faisal AA, Montana G, Leech R (2016a). **The Automatic Neuroscientist: A framework for optimizing experimental design with closed-loop real-time fMRI**. *NeuroImage*, 129: 320-334

 Lorenz R, Violante IR, Monti RP, Montana G, Hampshire A, Leech R. **Dissociating frontoparietal networks with neuroadaptive Bayesian optimization**. *Under revision* (preprint available on bioRxiv:128678)

 Lorenz R*, Monti RP*, Hampshire A, Koush Y, Anagnostopoulos C, Faisal A, Sharp D, Montana G, Leech R, Violante IR (2016b). **Towards tailoring non-invasive brain stimulation using real-time fMRI and Bayesian optimization**, In *6th International Workshop on Pattern Recognition in Neuroimaging* (free version available on arXiv:1605.01270)

 Lorenz R, Simmons L, Monti RP, Arthur J, Limal S, Leech R, Violante IR. **Assessing tACS-induced phosphene perception using adaptive Bayesian optimization**. *Under revision* (preprint available on bioRxiv: 150086)

Lorenz R, Monti RP, Koush Y, Sharp D, Montana G, Hampshire A, Leech R, Violante IR. **Towards tailoring non-invasive stimulation using neuroadaptive Bayesian optimization**. *In preparation*.

Questions/Feedback?



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@romy_lorenz

general

cognition

brain stimulation