



Neuroadaptive Bayesian Optimization

Implications for the Cognitive Sciences

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Human Cognitive and Brain Sciences



Overview

1. Motivation
2. The framework
 - 2.1 Bayesian optimization
 - 2.2 Validation study
3. Application studies
 - 3.1 Human brain mapping
 - 3.2 Non-invasive brain stimulation
 - 3.3 Biomarker discovery
4. Implications & Discussion

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Aims of cognitive neuroscience

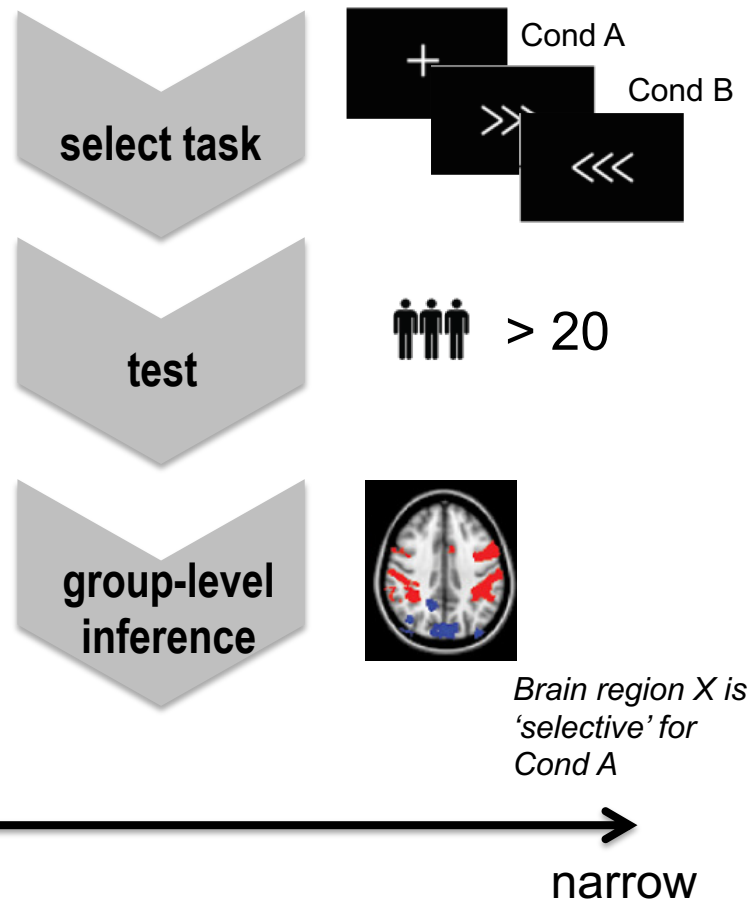
Research questions

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?

Standard approach



Aims of cognitive neuroscience

Lorenz et al. *TiCS* 2017

Human-brain mapping

- Over-specified inferences about functional-anatomical mappings
 - right IFG Hampshire & Sharp *TiCS* 2015
 - dACC Wager et al. *PNAS* 2016

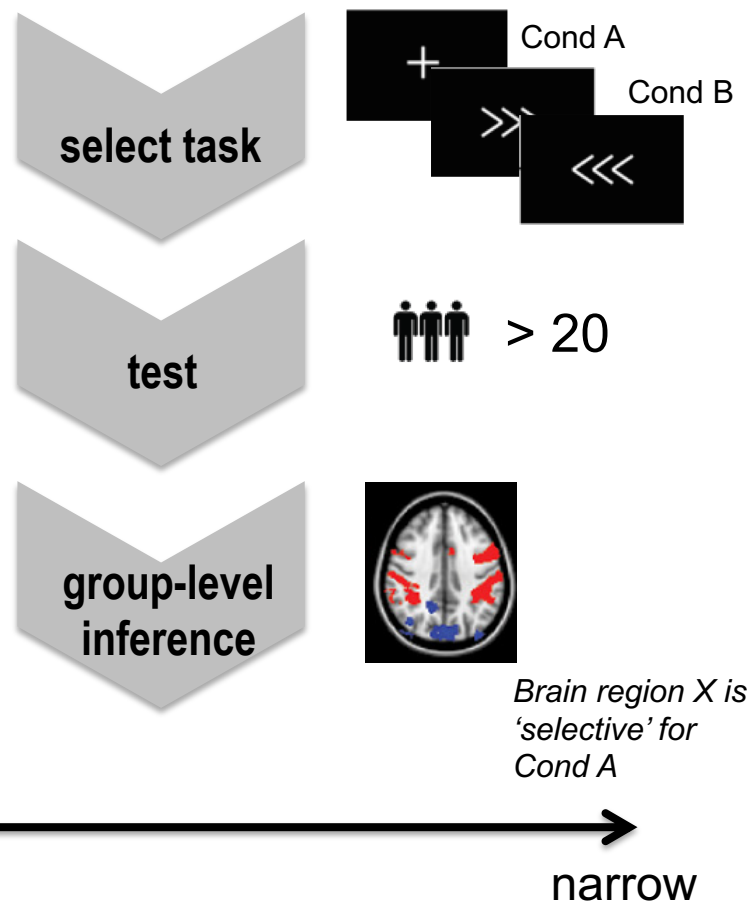
Biomarker discovery

- 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

Standard approach



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The framework

neuroadaptive paradigms

open-loop

- stimuli manually adapted to subject
e.g. [37]
- subject modulates brain response
e.g. simple NF [36], early BCIs [34,35], communication with vegetative state patients [38]

informed open-loop

- stimuli is triggered by brain state
e.g. TMS is triggered by EEG signal [26]

closed-loop

machine learning

supervised

passive learning

- BCIs
e.g. [43]
- advanced NF
e.g. [39-42]
- neural selectivity
e.g. [44]

active learning

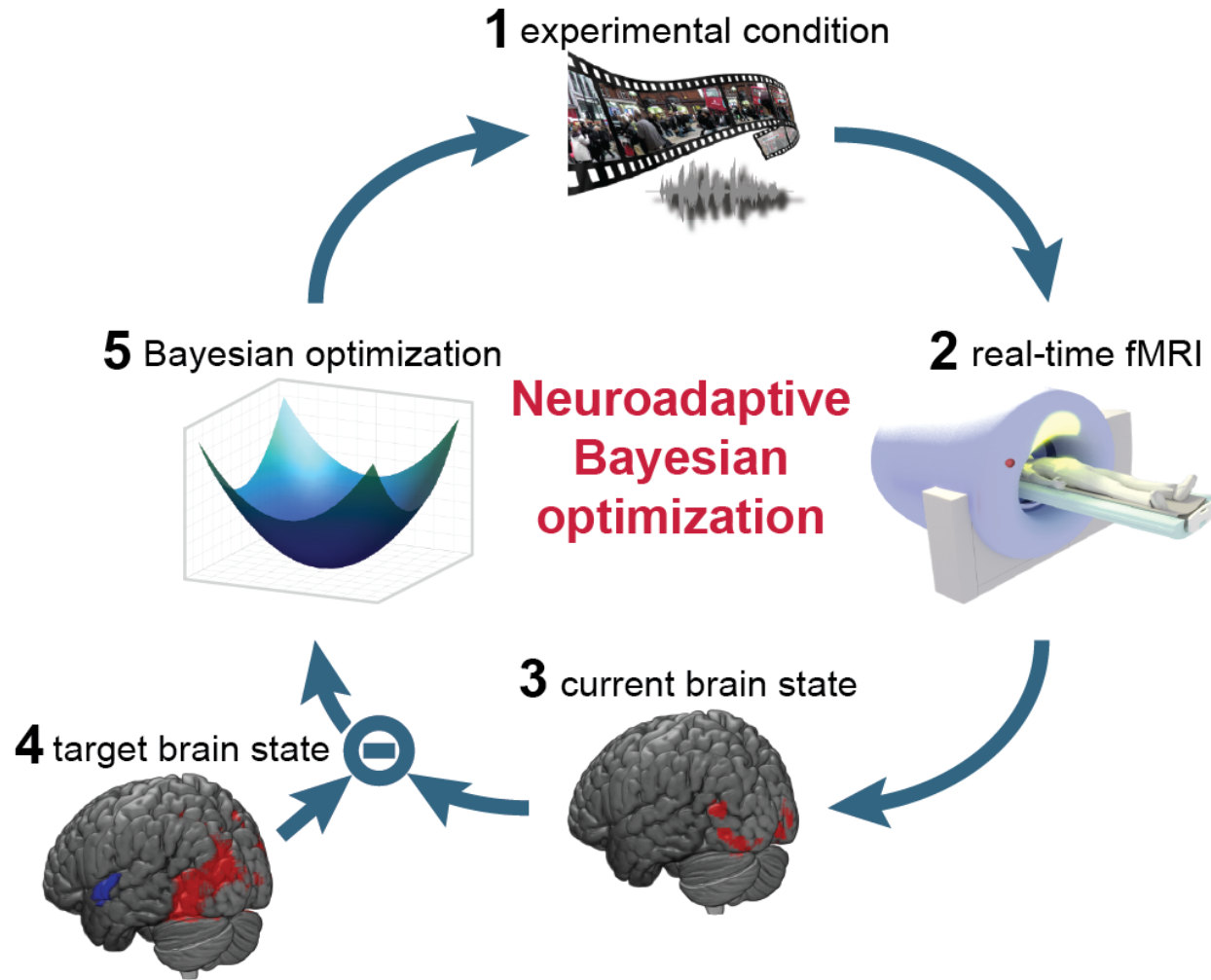
- receptive fields
e.g. [32]

Bayesian optimization

e.g. [47, 51]

Lorenz et al. *TiCS* 2017

The framework

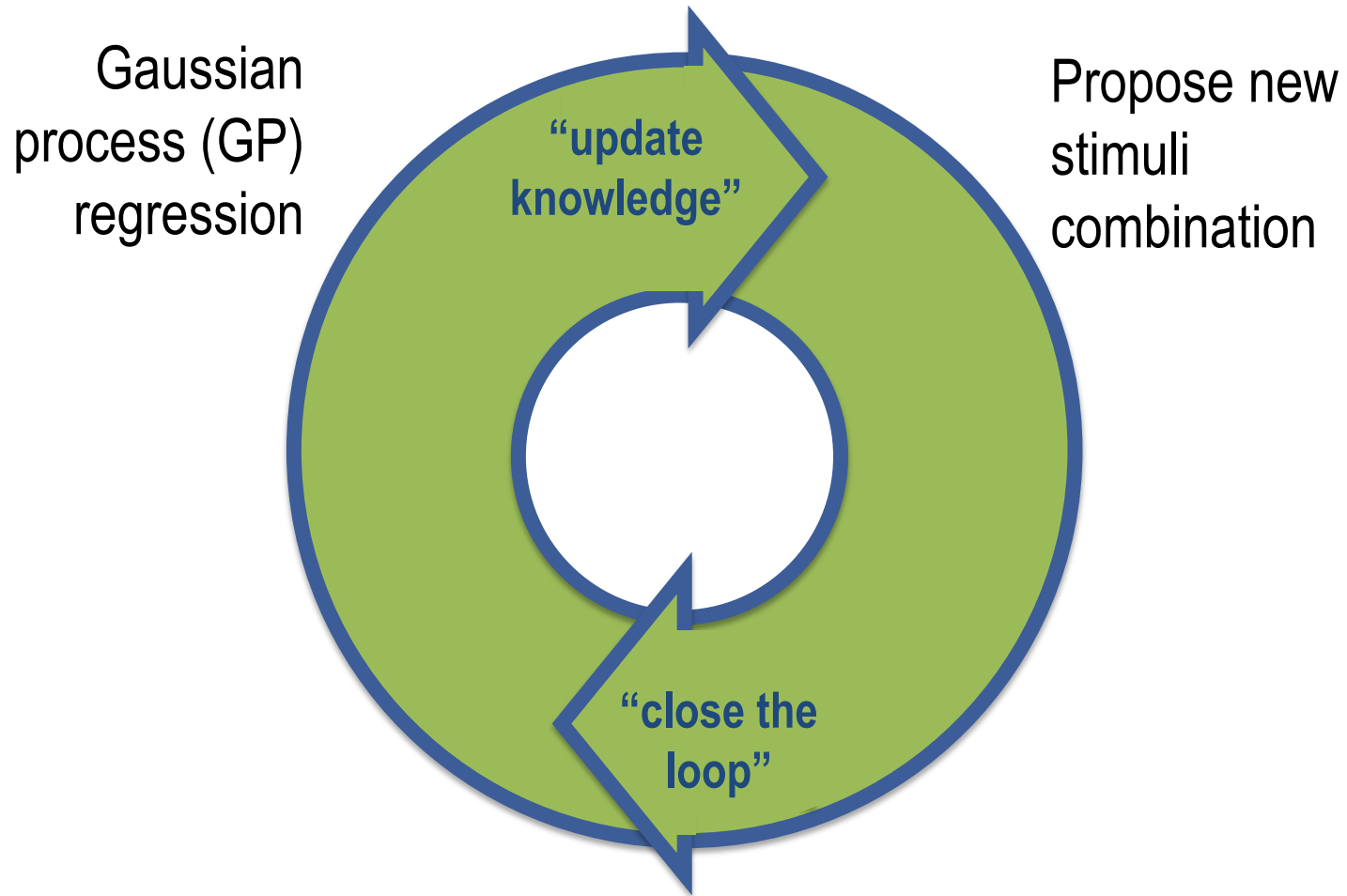


Lorenz et al. *NeuroImage* 2016

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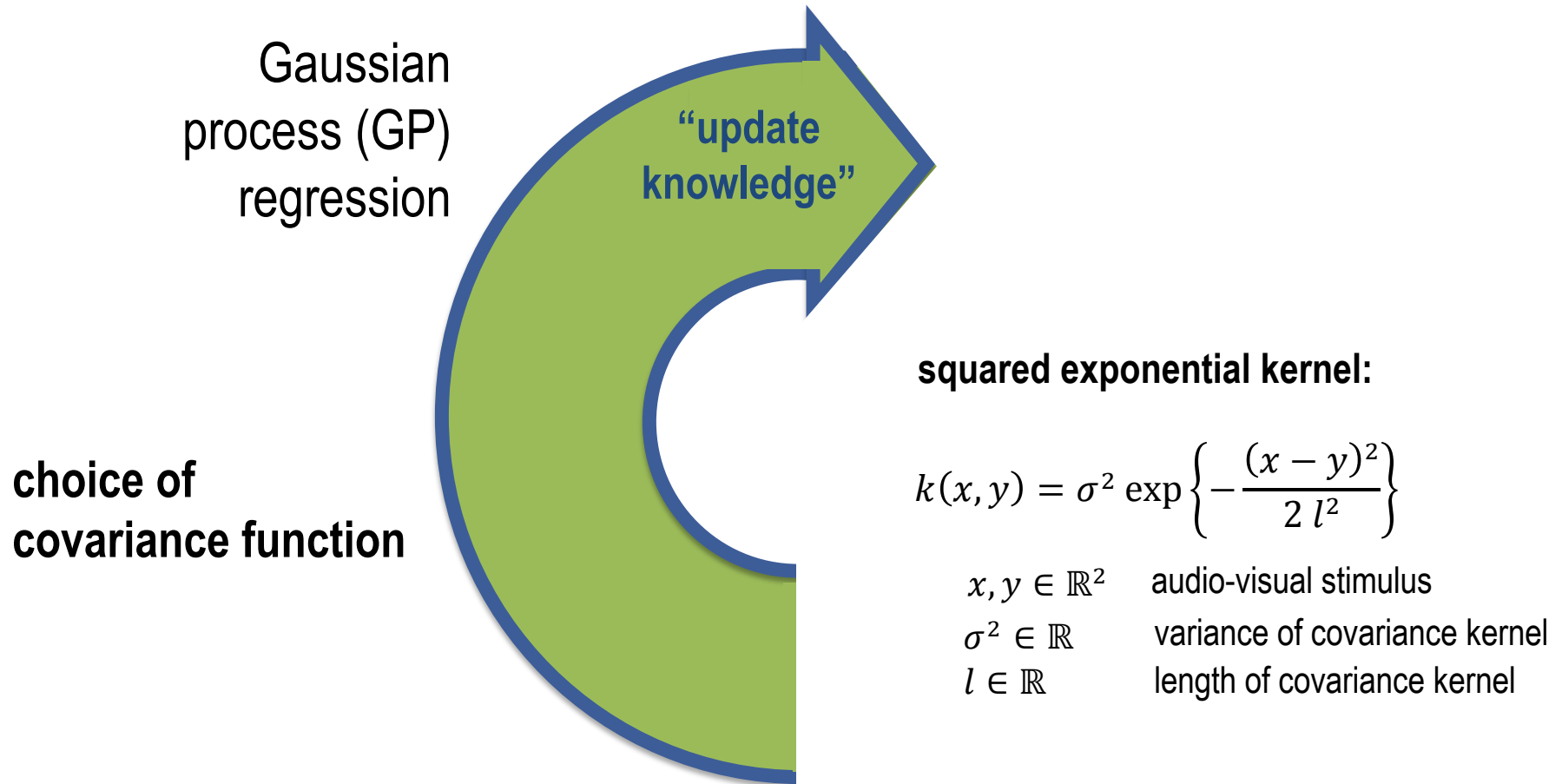
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Bayesian optimization



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

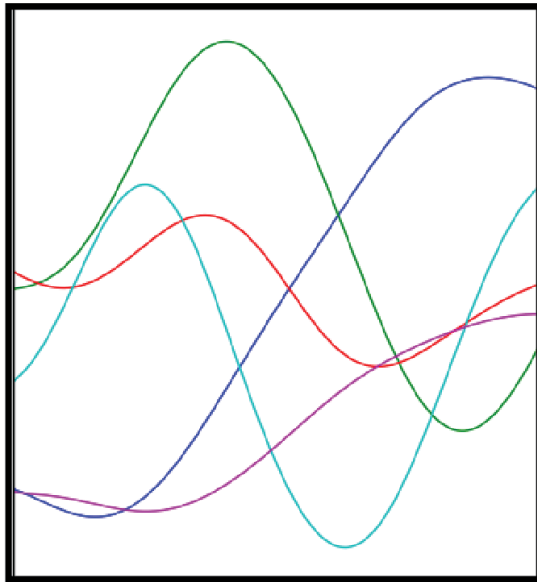
Bayesian optimization



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

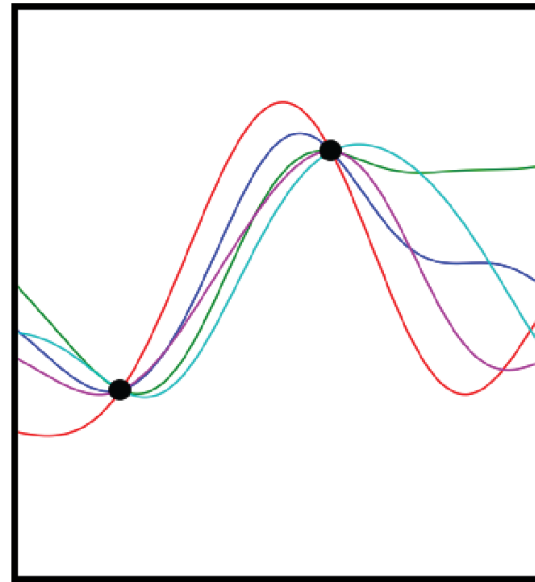
GP regression (1D – example)

prior



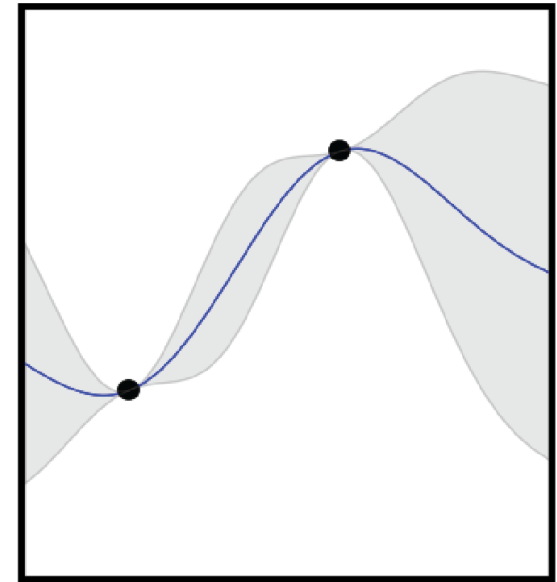
experiment space

posterior



experiment space

prediction



experiment space

Bayesian optimization

Expected improvement acquisition function:

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

$m(x)$: predicted mean

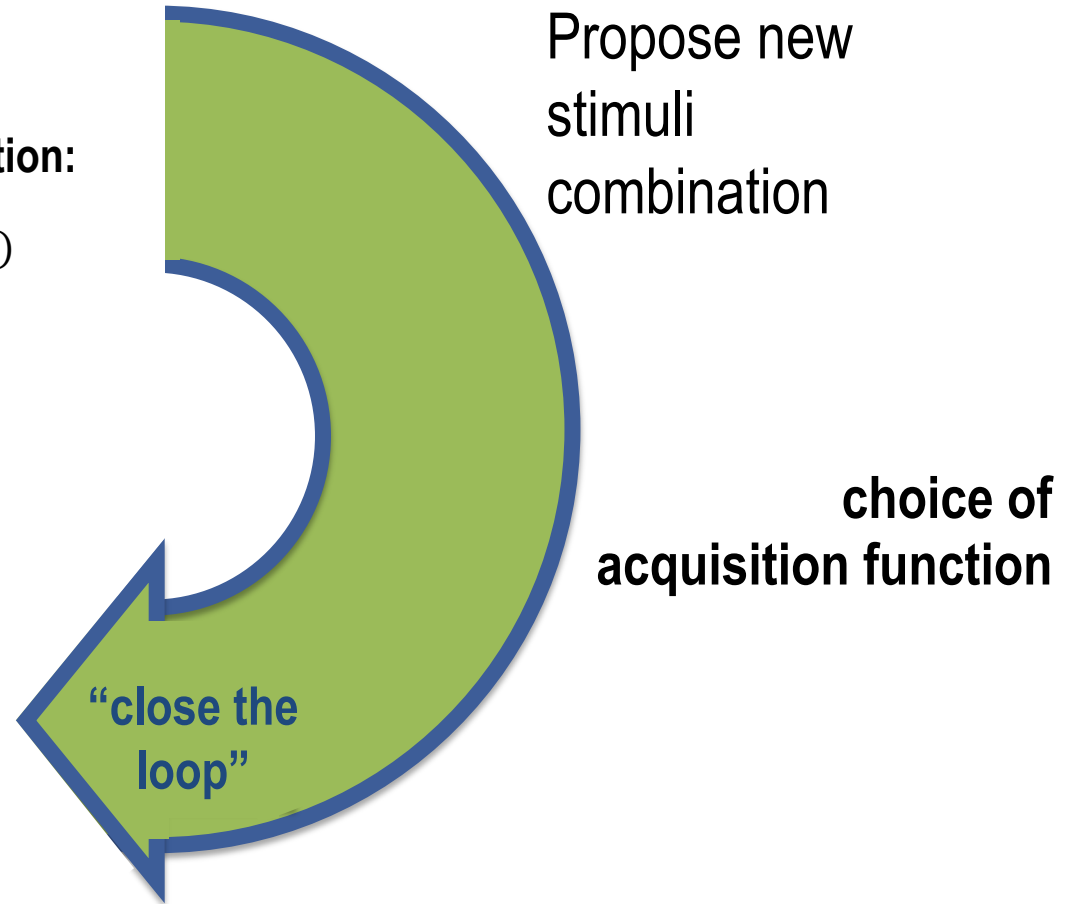
$var(x)$: predicted variance

f_{max} : maximum predicted value

$q()$: cumulative distribution function

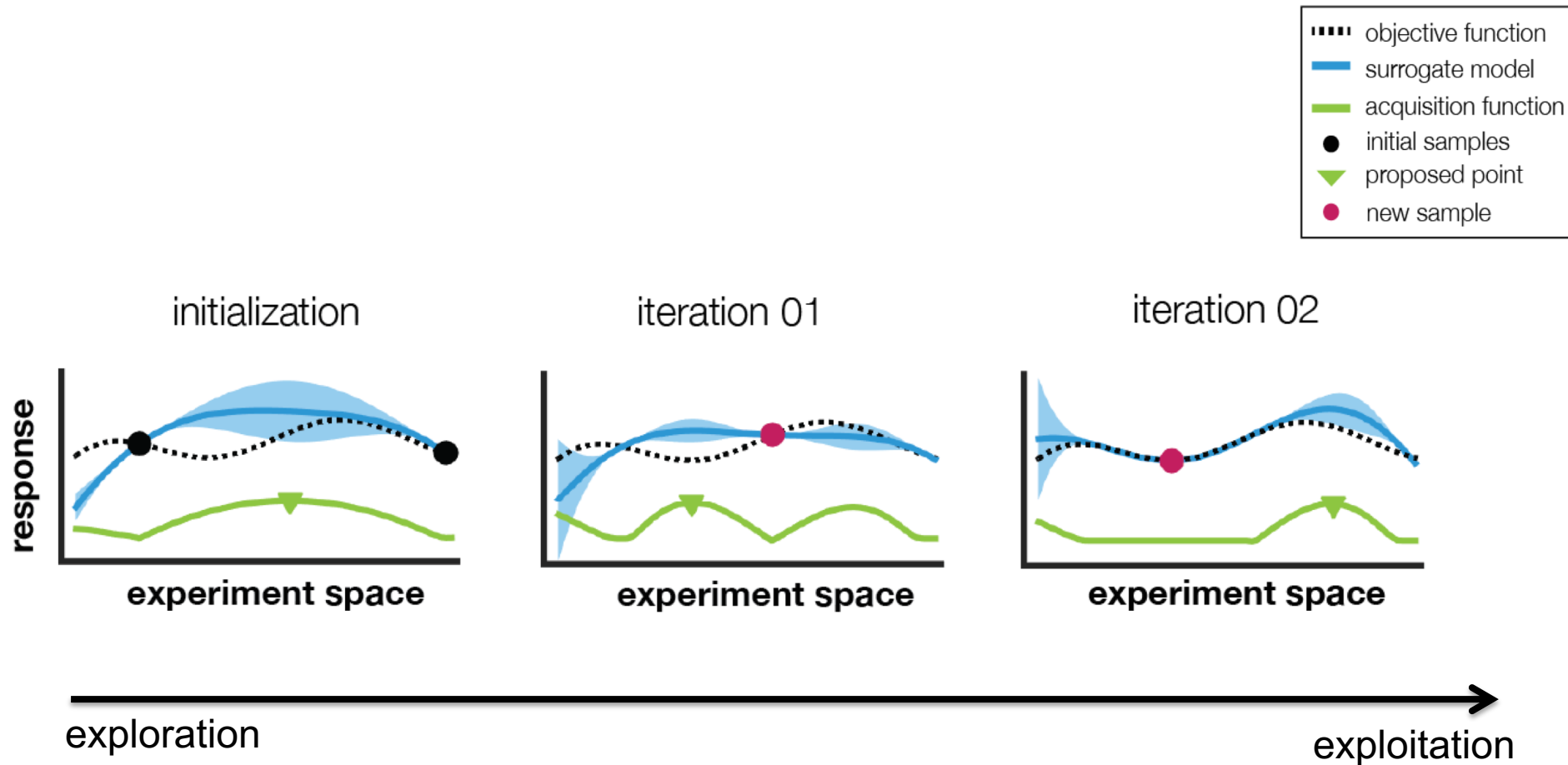
$p()$: probability density function

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



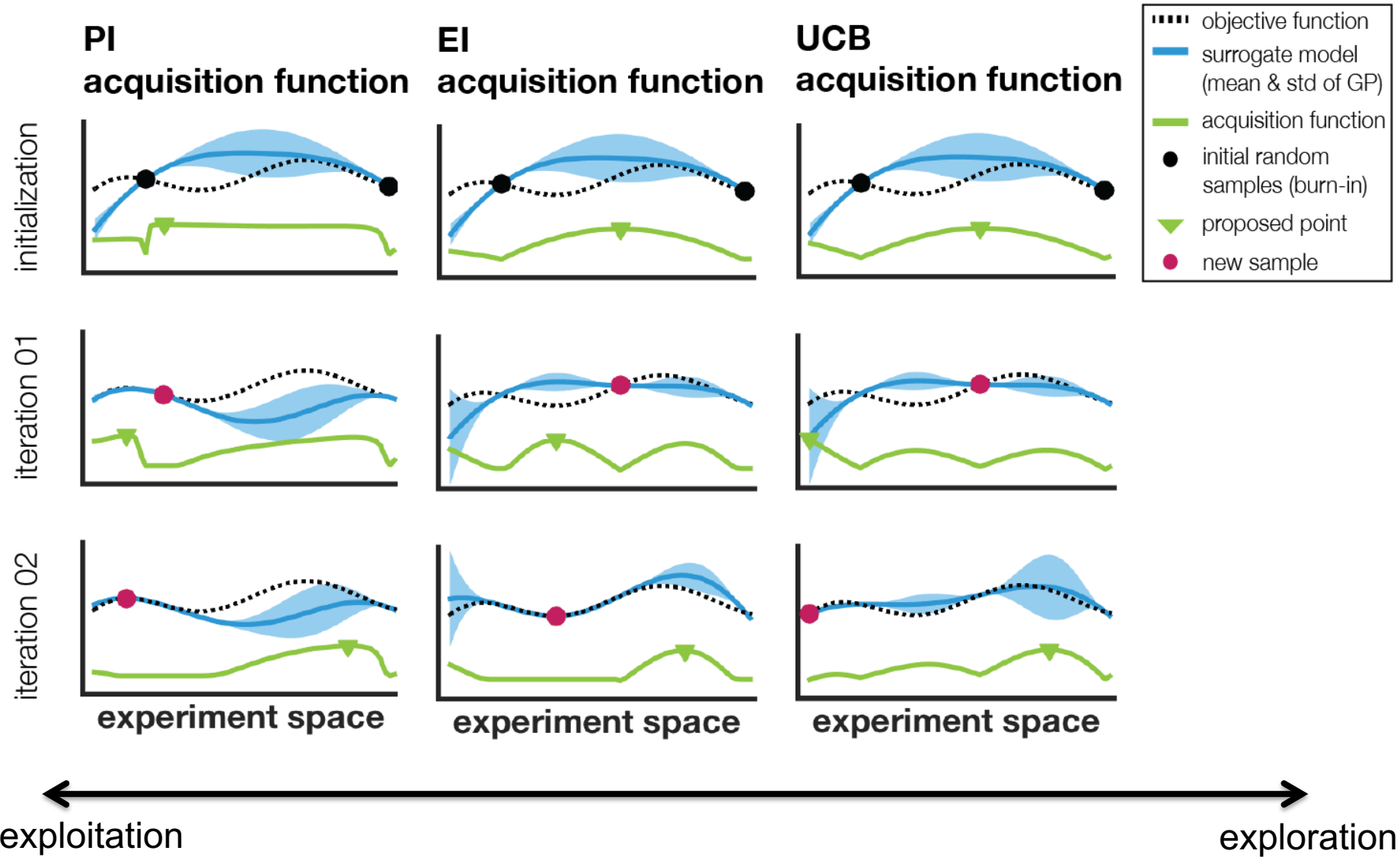
Rasmussen & Williams 2006
Brochu et al. *arXiv* 2010

Bayesian optimization (1D – example)



Lorenz et al. *TiCS* 2017

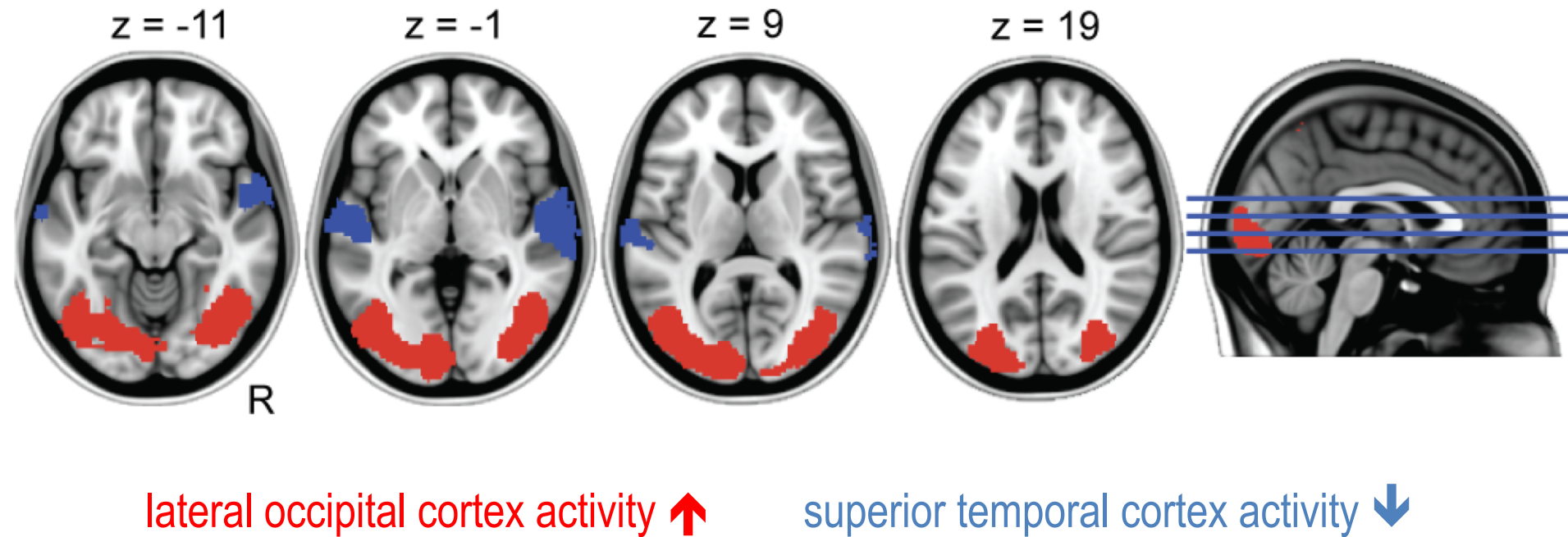
Trade off between exploration and exploitation



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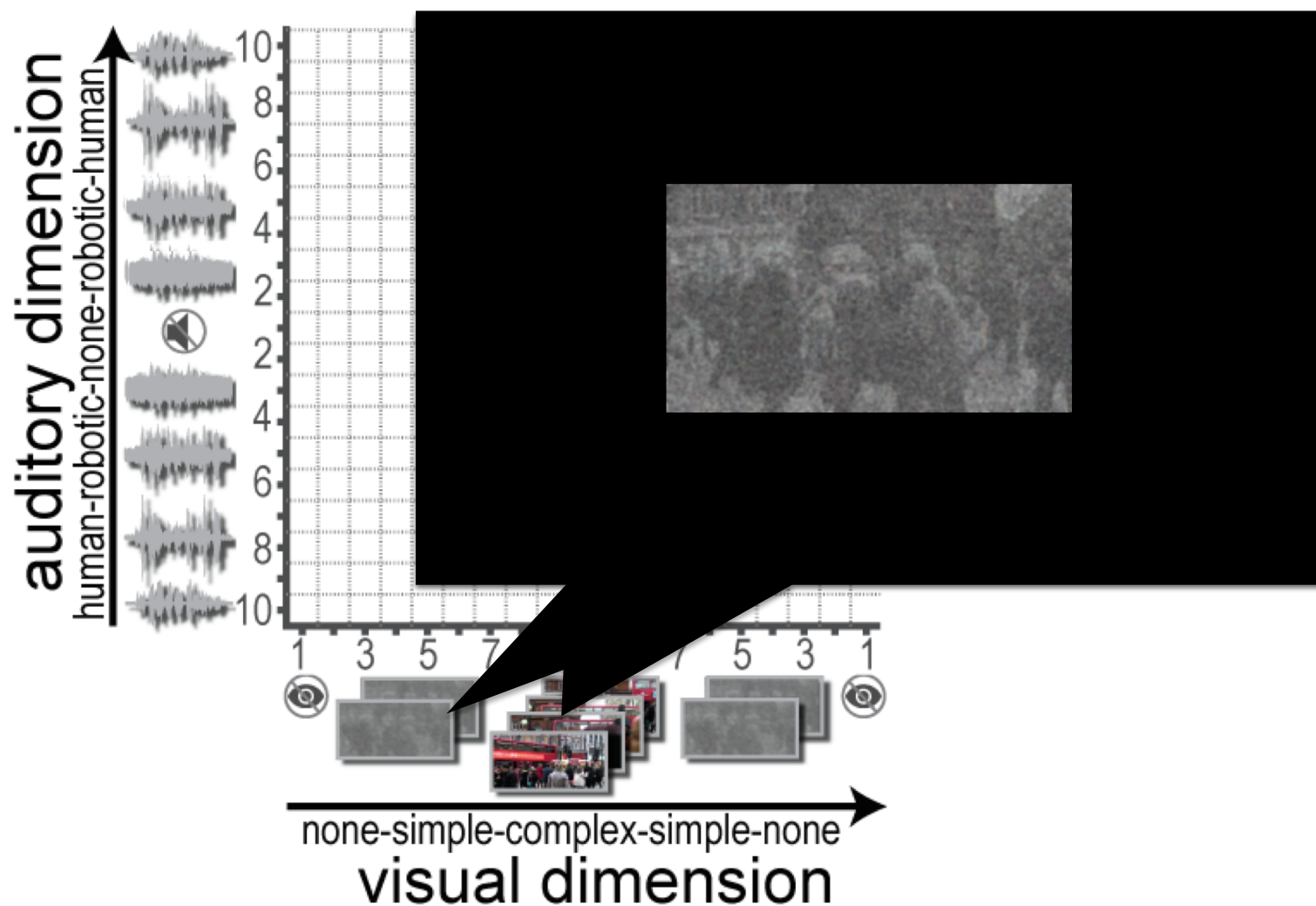
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Target brain state

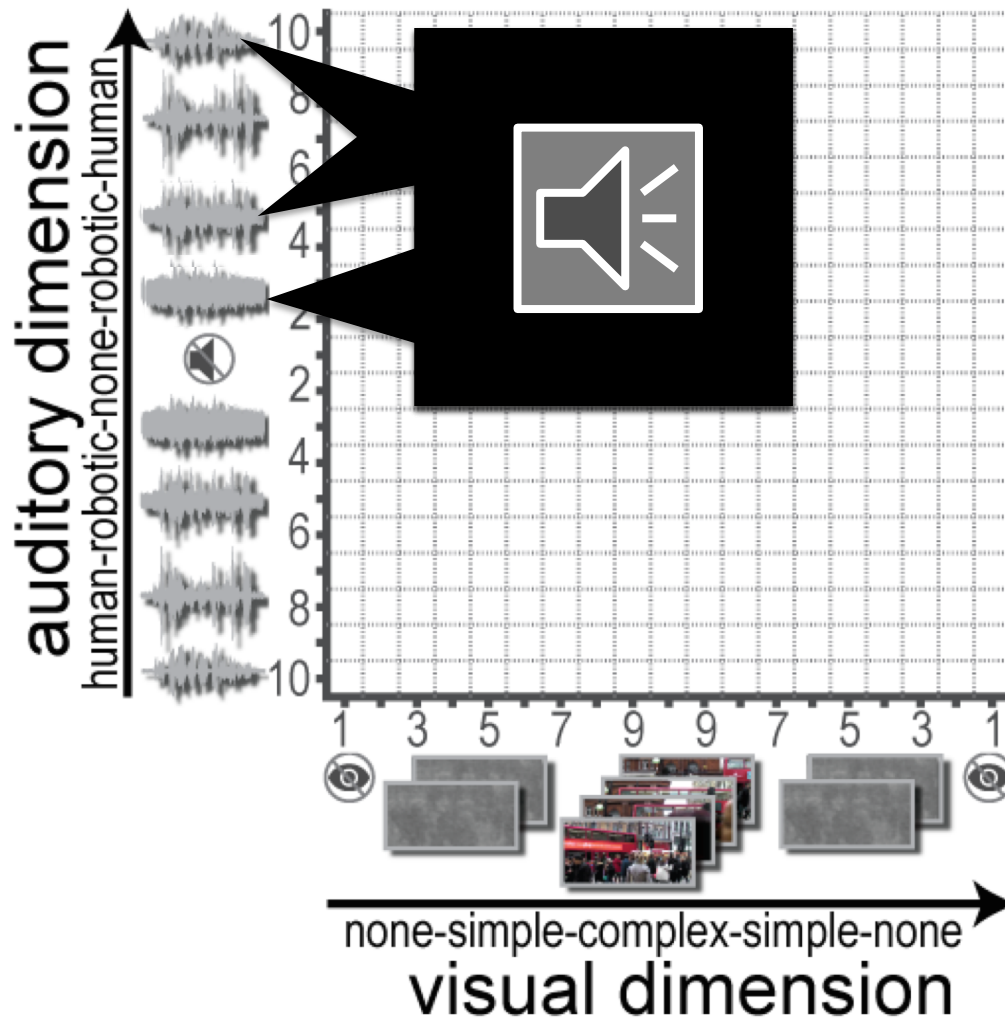


masks derived from Braga et al. *NeuroImage* 2013

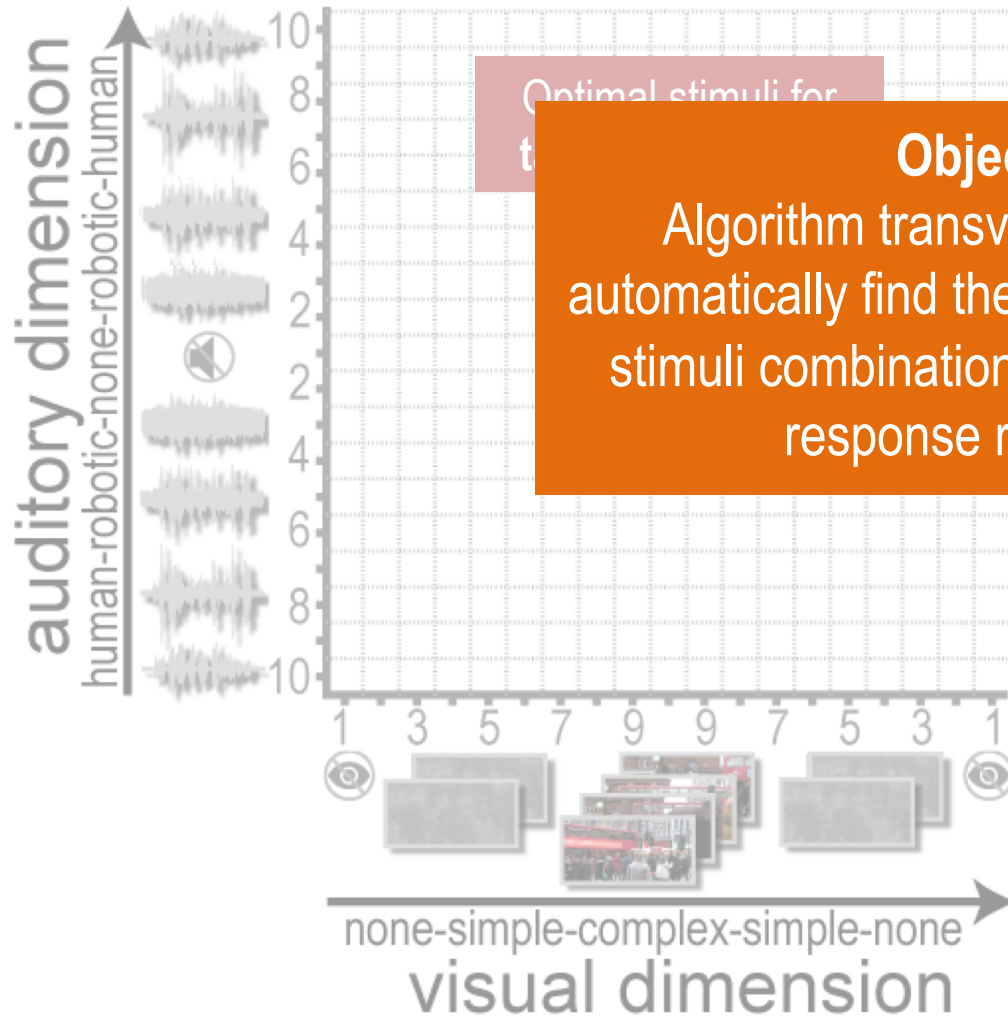
Experiment space



Experiment space



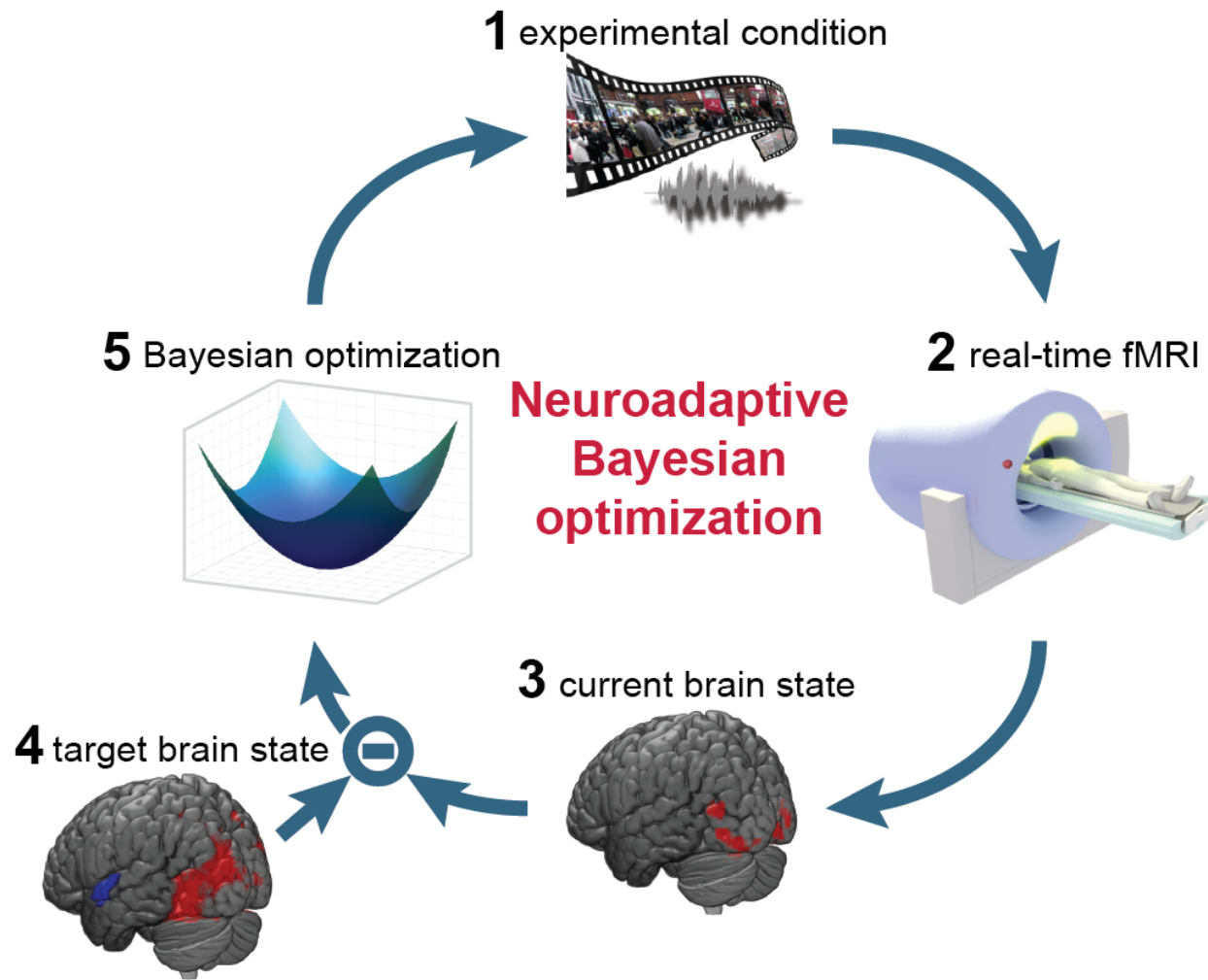
Experiment space



Objective:

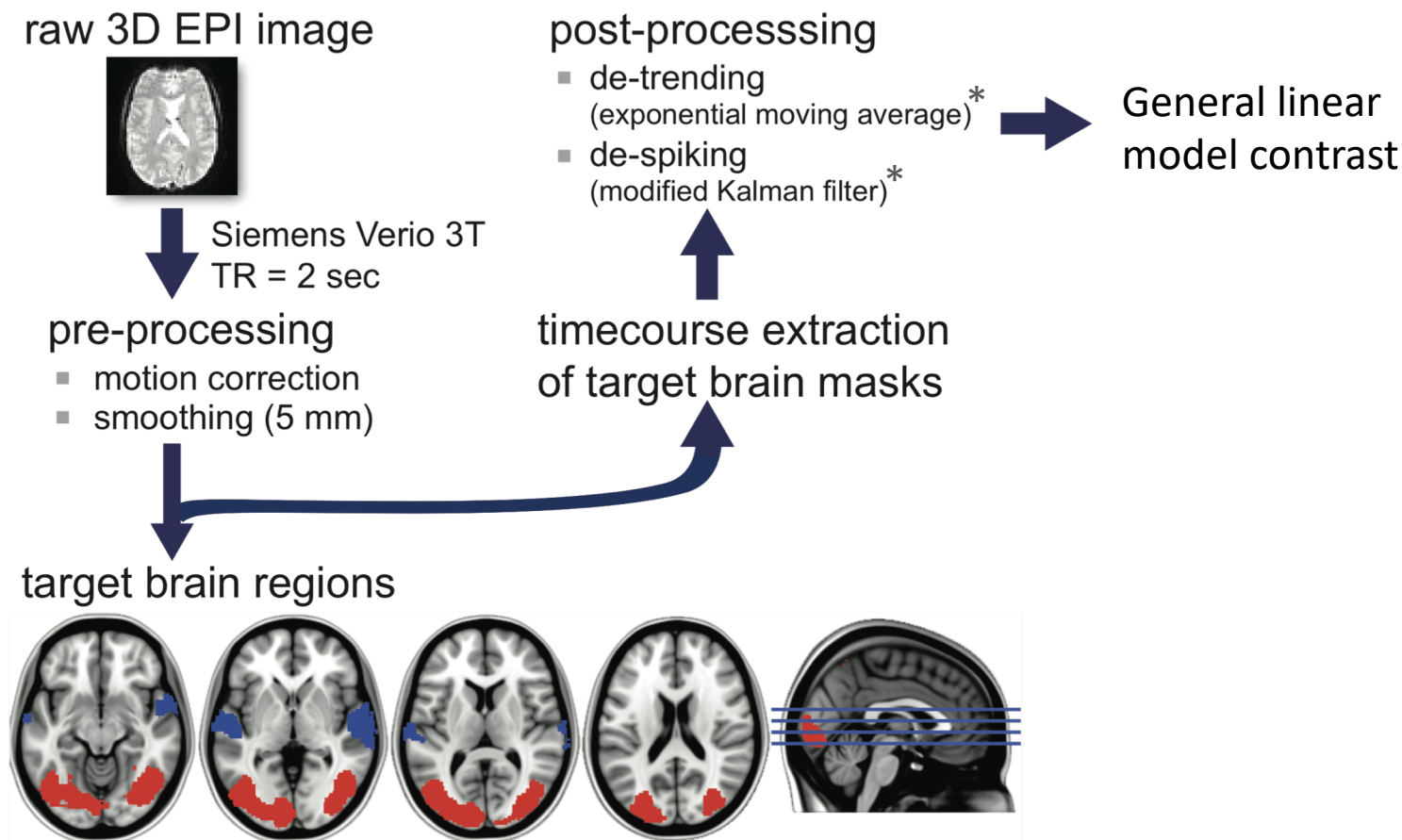
Algorithm transverses the space,
automatically find the optimal audio-visual
stimuli combination & map out stimuli-
response relationship

Method



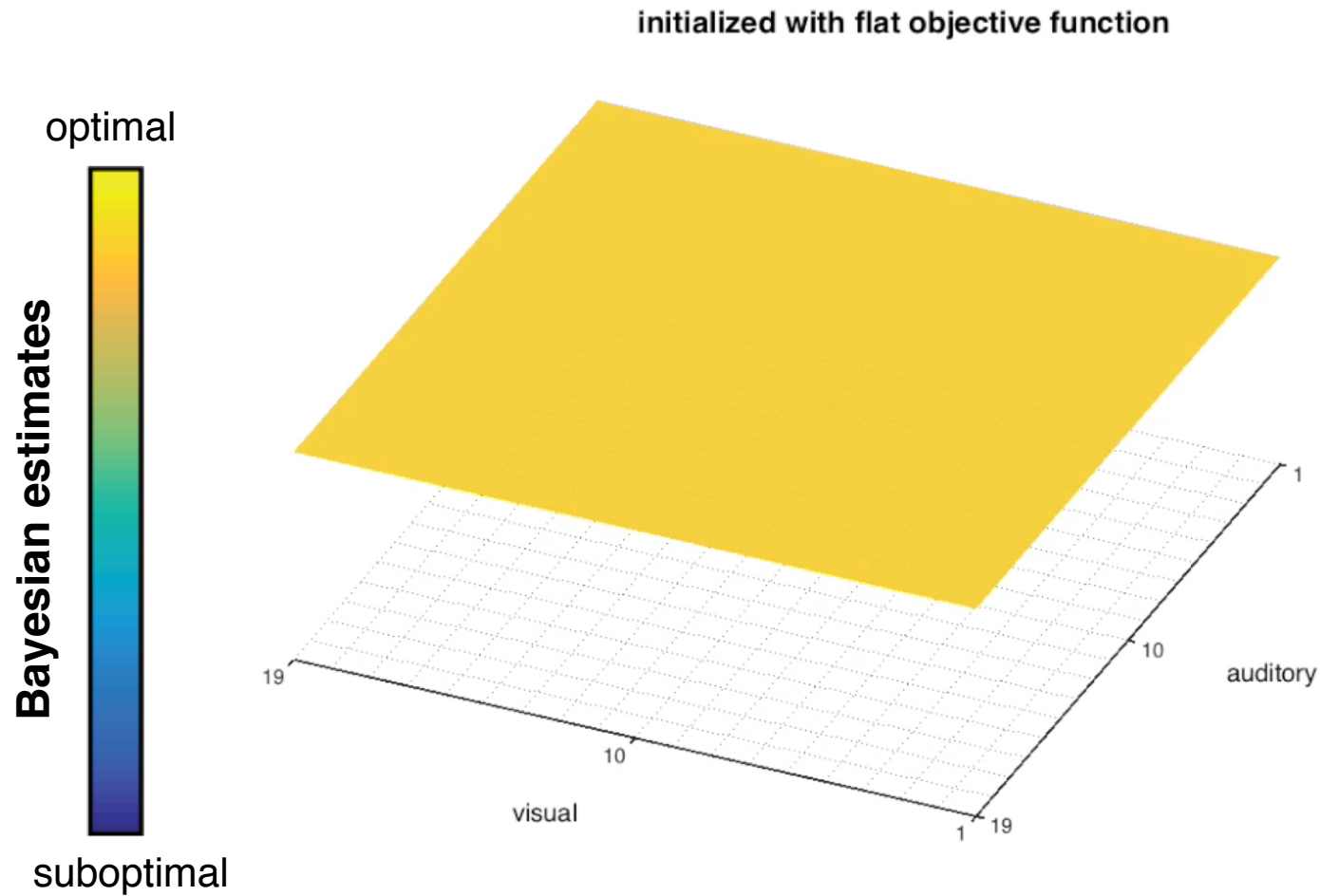
Lorenz et al. *NeuroImage* 2016

Real-time fMRI pipeline



* both algorithms obtained from
Koush et al. *NeuroImage* 2012

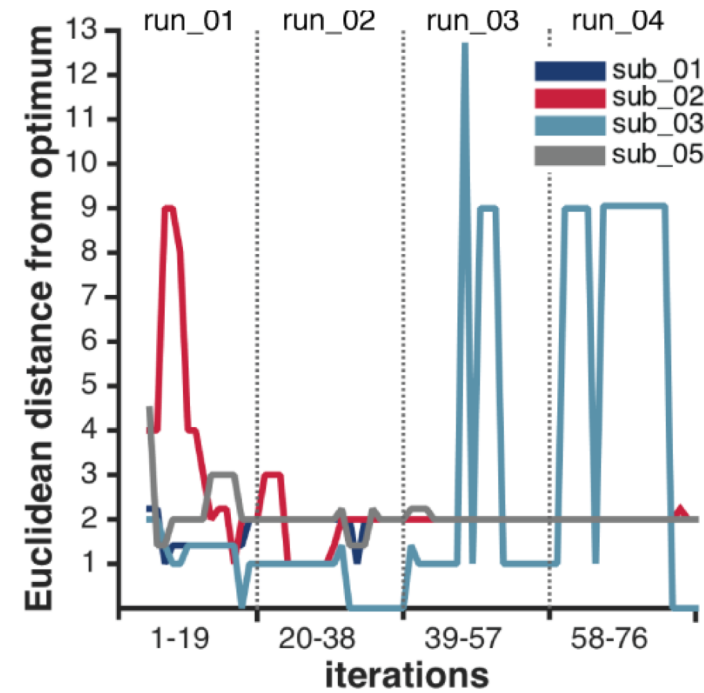
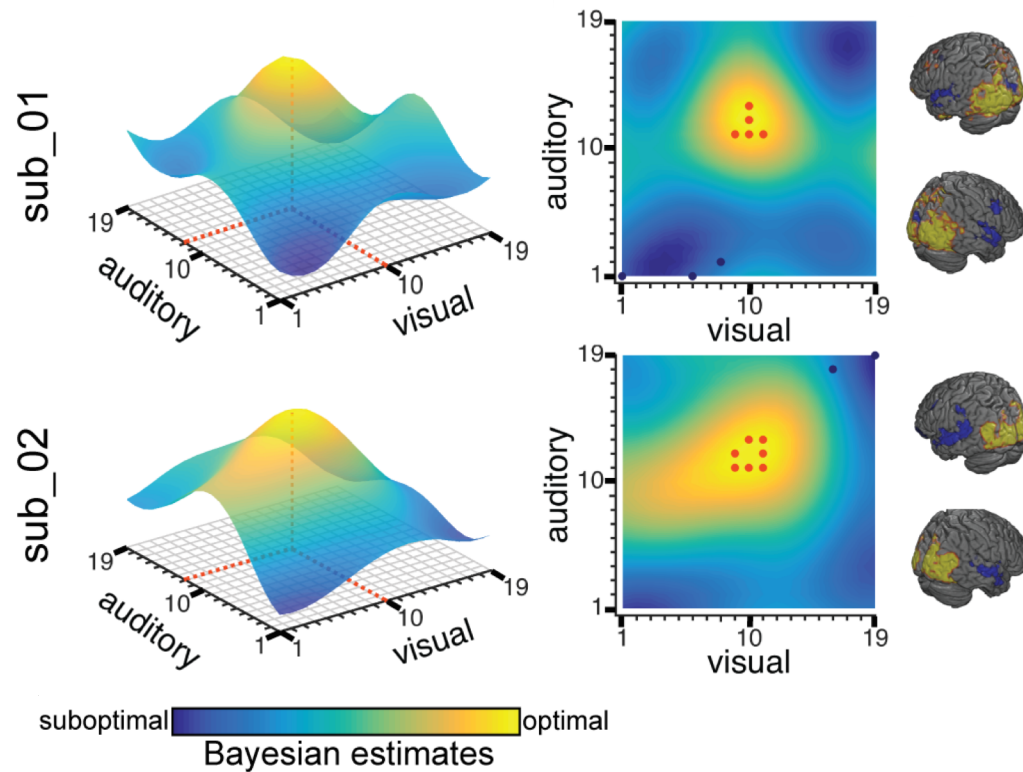
Example of one subject



Lorenz et al. *NeuroImage* 2016

Results

👤👤👤 = 5

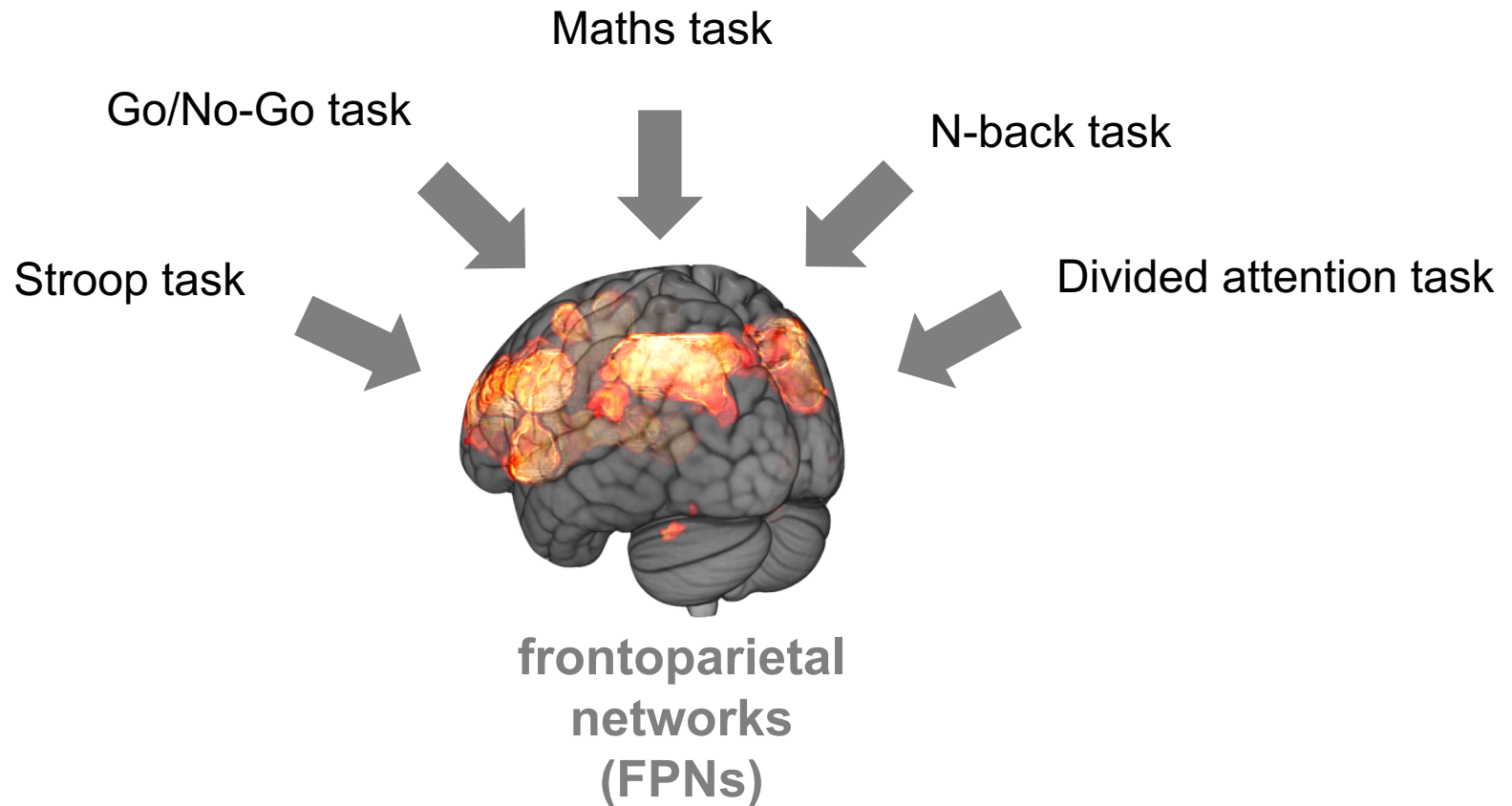


Lorenz et al. *NeuroImage* 2016

Overview

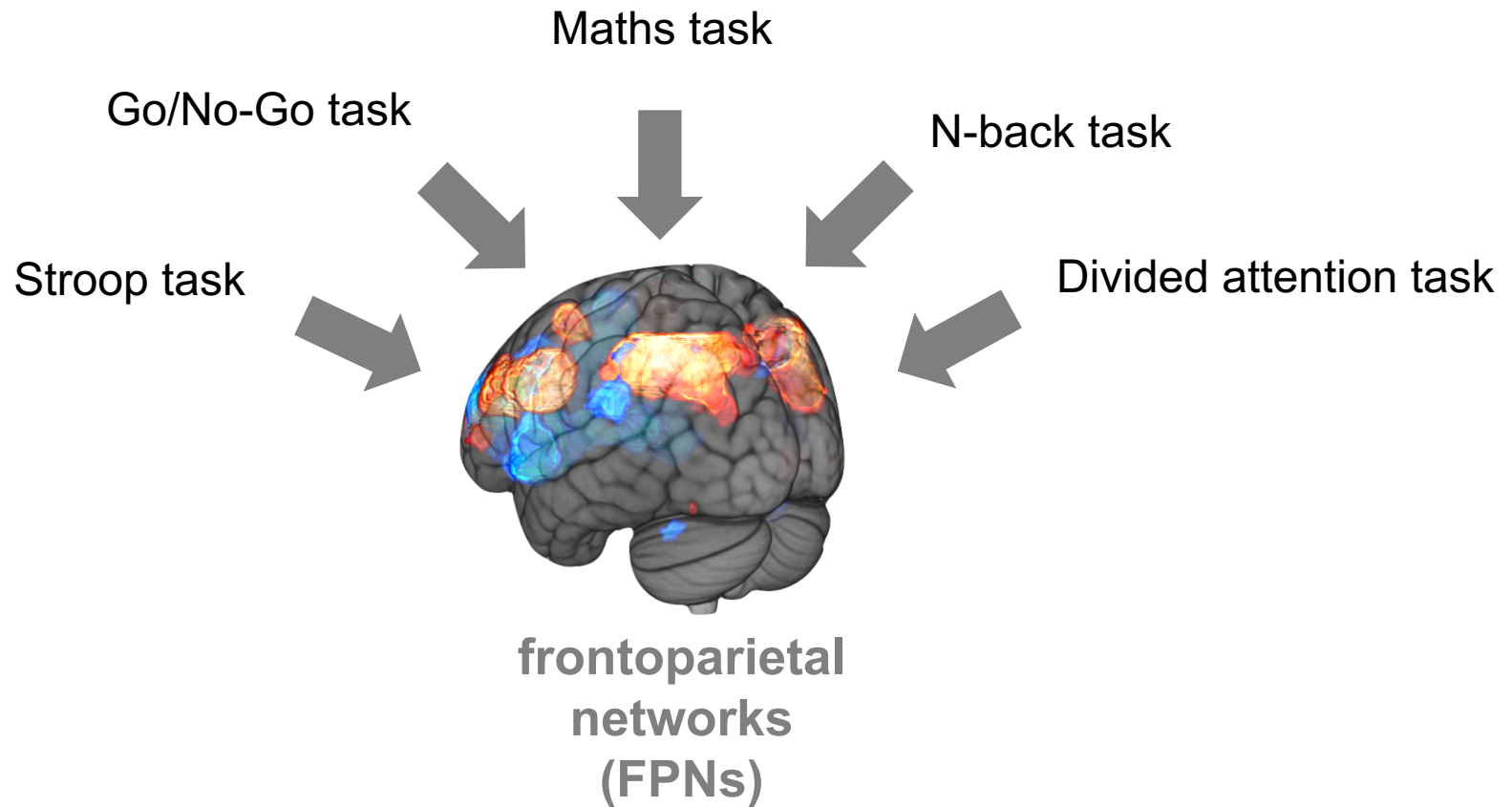
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Many-to-many mapping problem



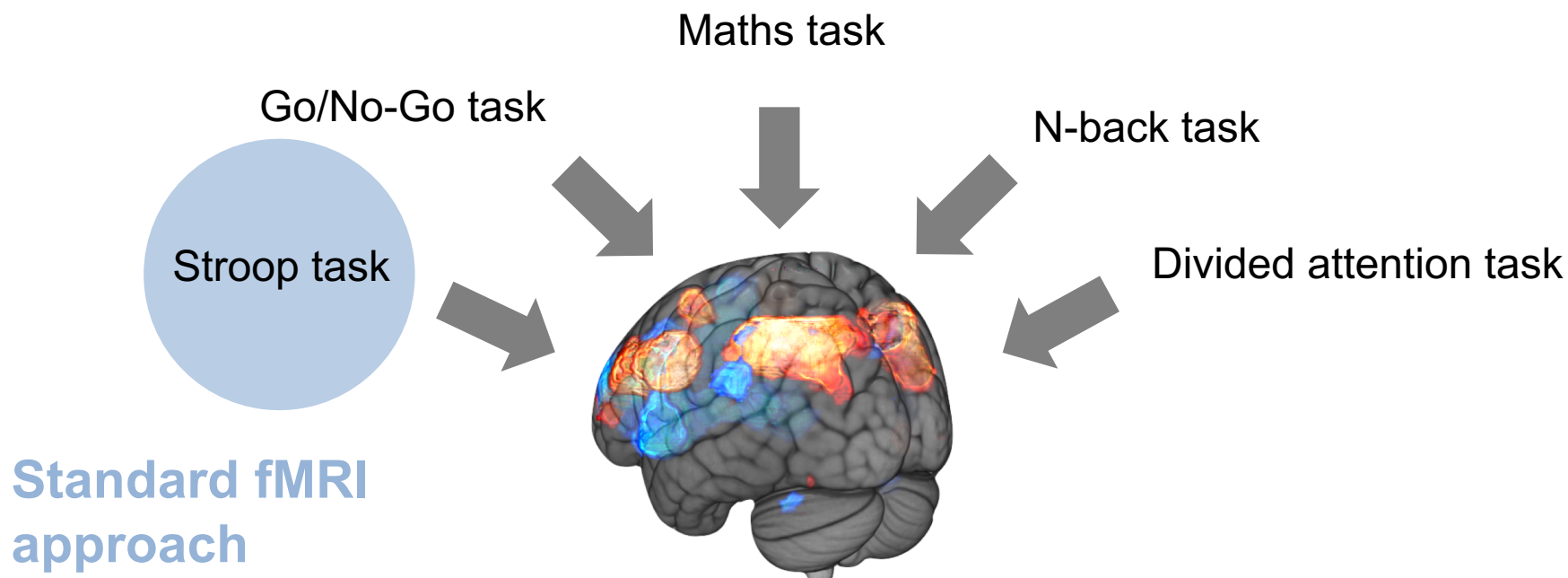
Duncan & Owen *TiNS* 2000
Fedorenko et al. *PNAS* 2013

Many-to-many mapping problem



Hampshire et al. *Neuron* 2012

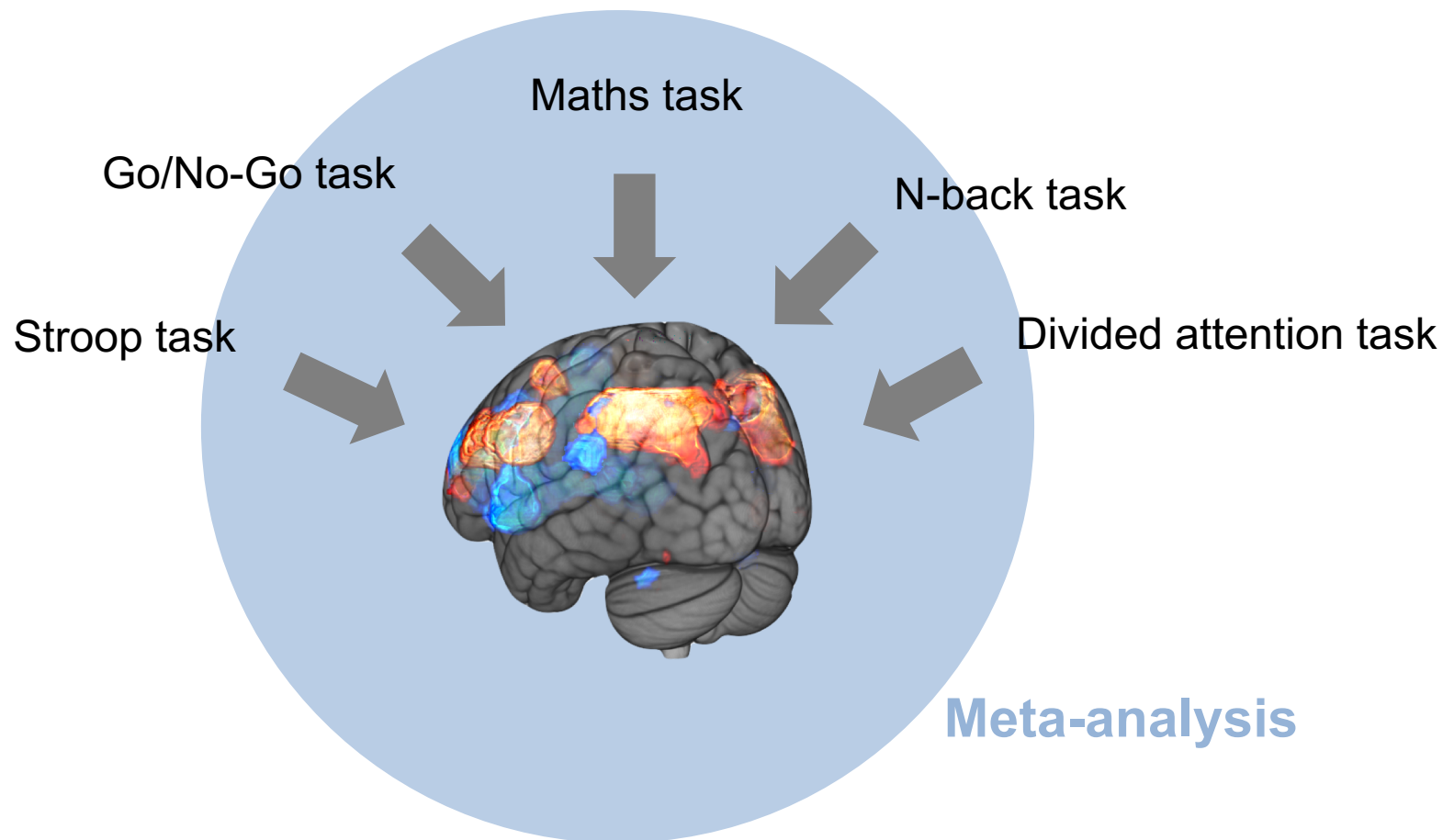
Motivation



- **Limited generalizability**
- **Limited reproducibility**

Lorenz et al. *TiCS* 2017
Westfall et al. *Wellcome Open Research* 2017

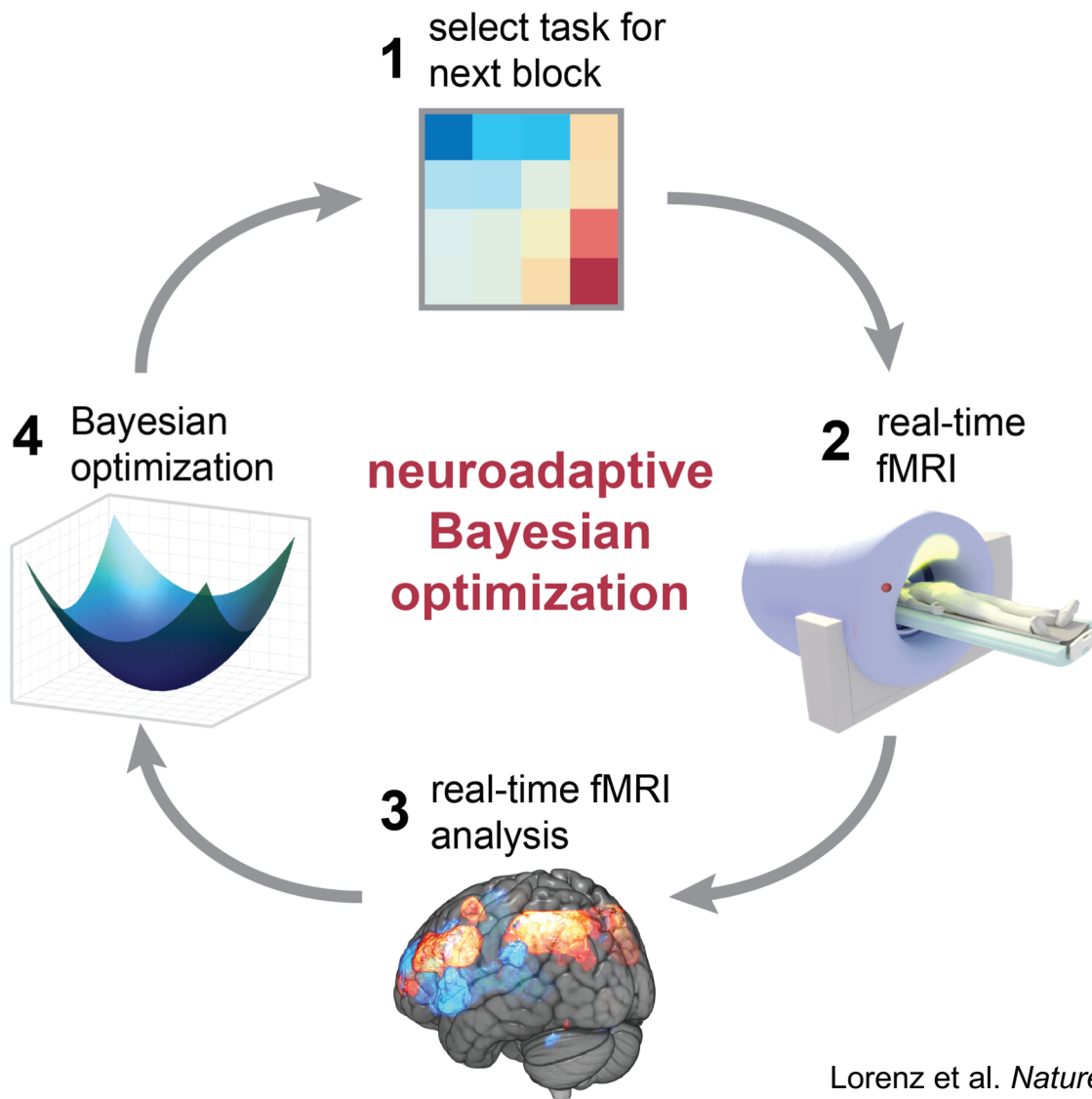
Motivation



- **Broad approximation**
- **Prone to biases**

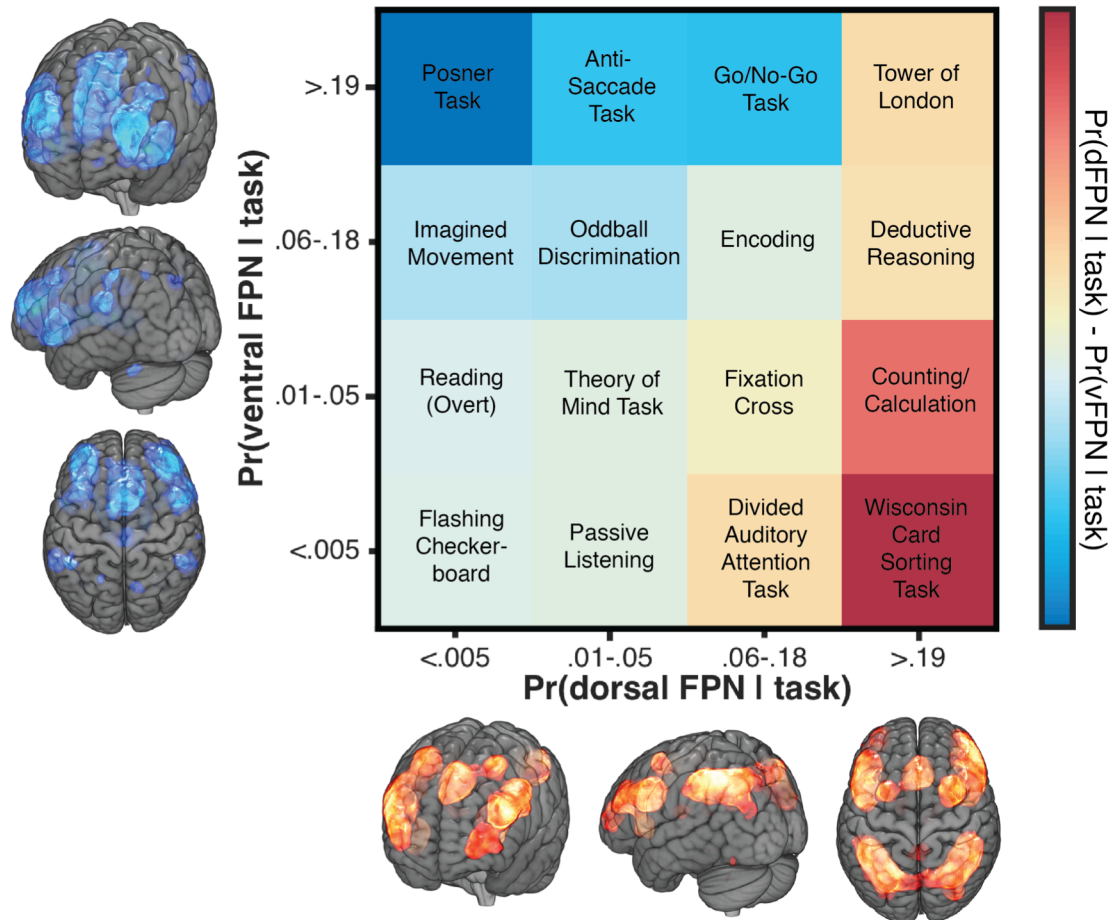
Yarkoni et al. *Nature Methods* 2011
Poldrack & Yarkoni *Annual Review of Psychology* 2016

Searching across cognitive tasks



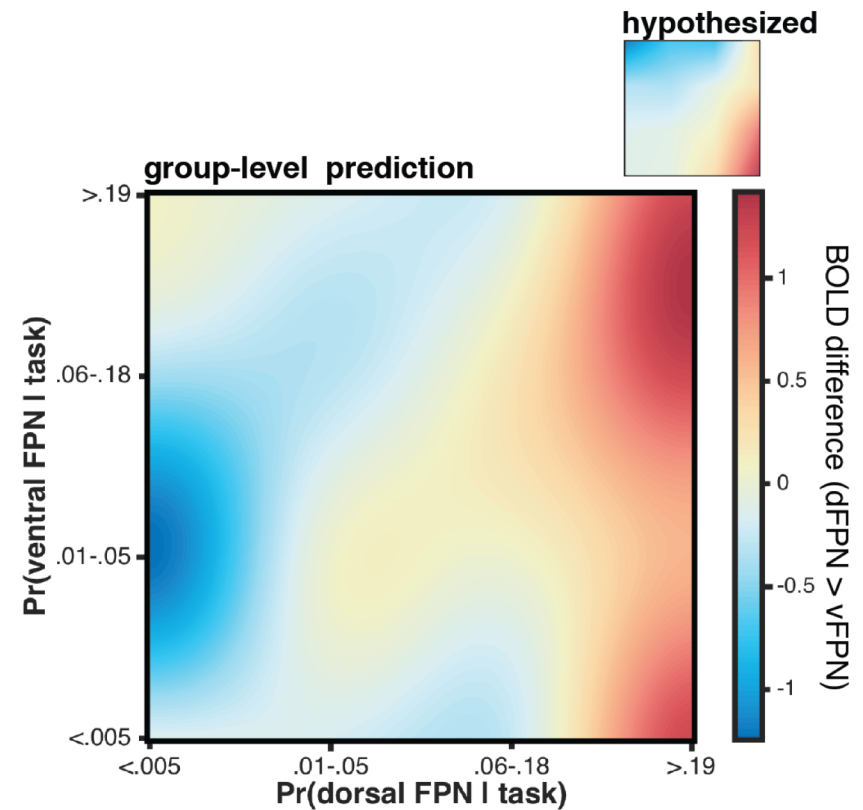
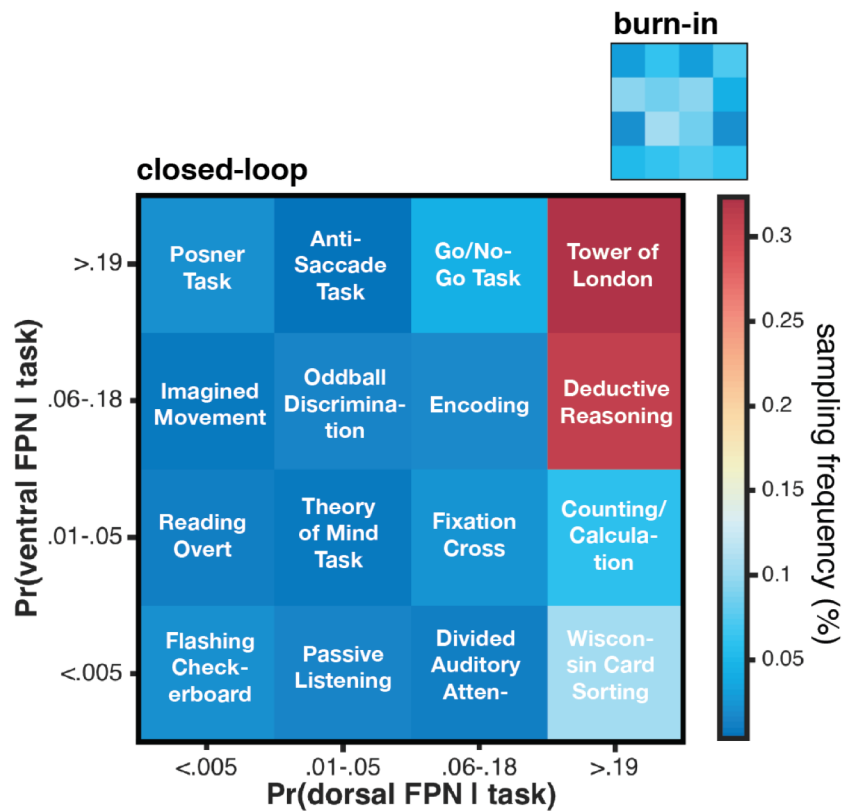
Lorenz et al. *Nature Communications* 2018

Task space based on meta-analysis



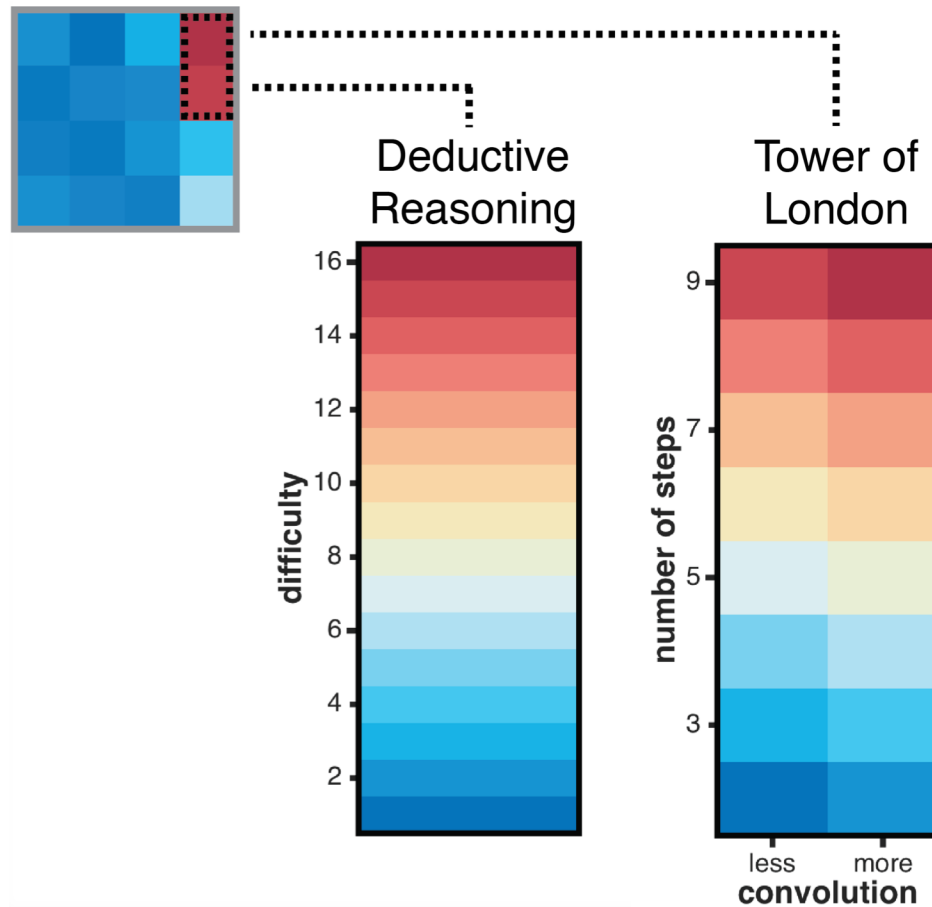
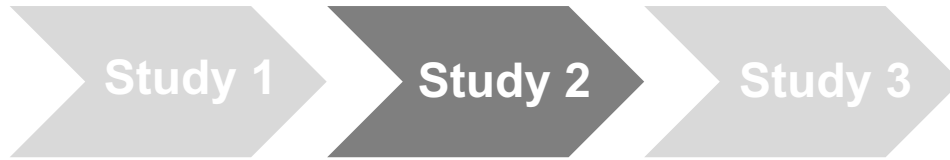
maps & space from
Yeo et al. *Cerebral Cortex* 2015

Find optimal tasks



Tower of London & Deductive Reasoning tasks maximally dissociate FPNs

Zoom in task space and fine-tune tasks

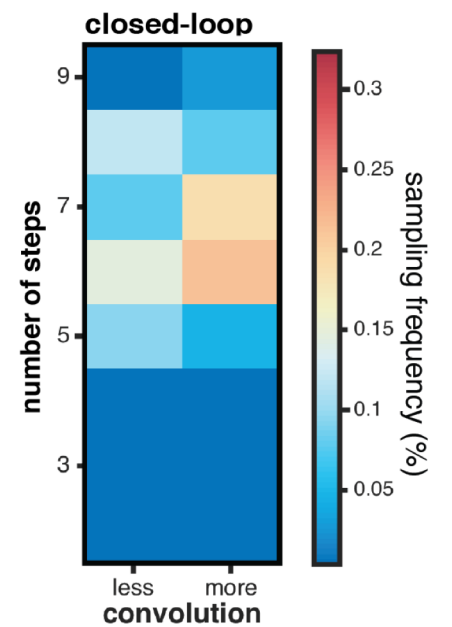
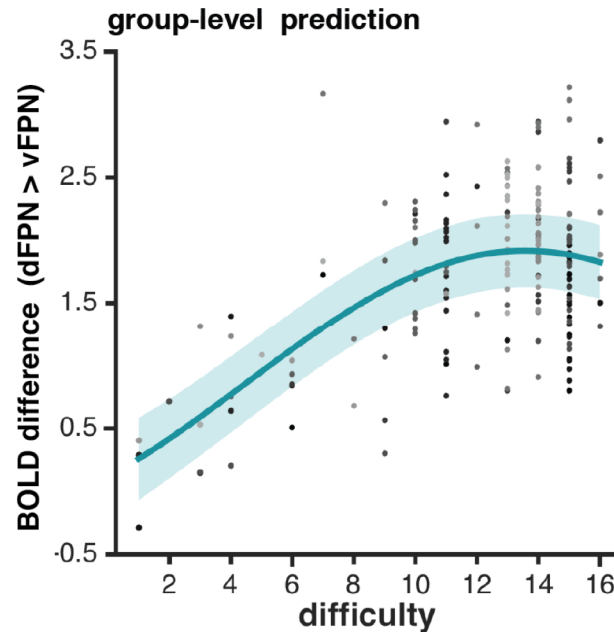
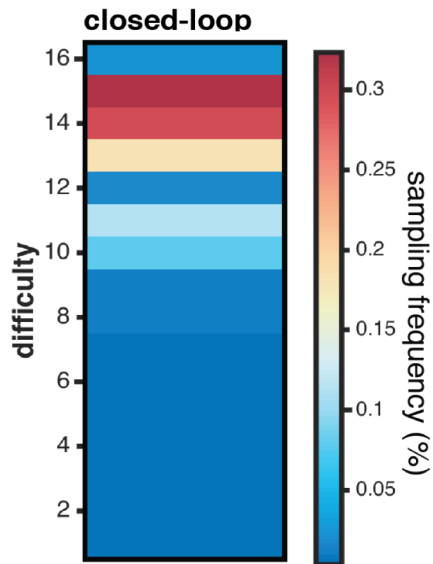


Find optimal task parameters

Study 1

Study 2

Study 3



Deductive Reasoning

Tower of London

Find *unique* functional activation profile

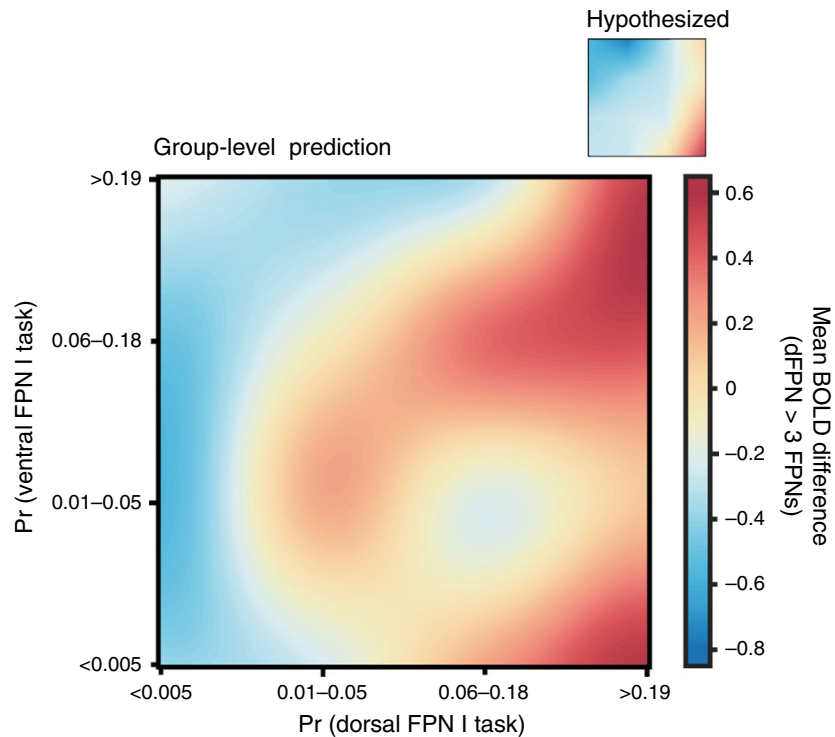
Study 1

Study 2

Study 3

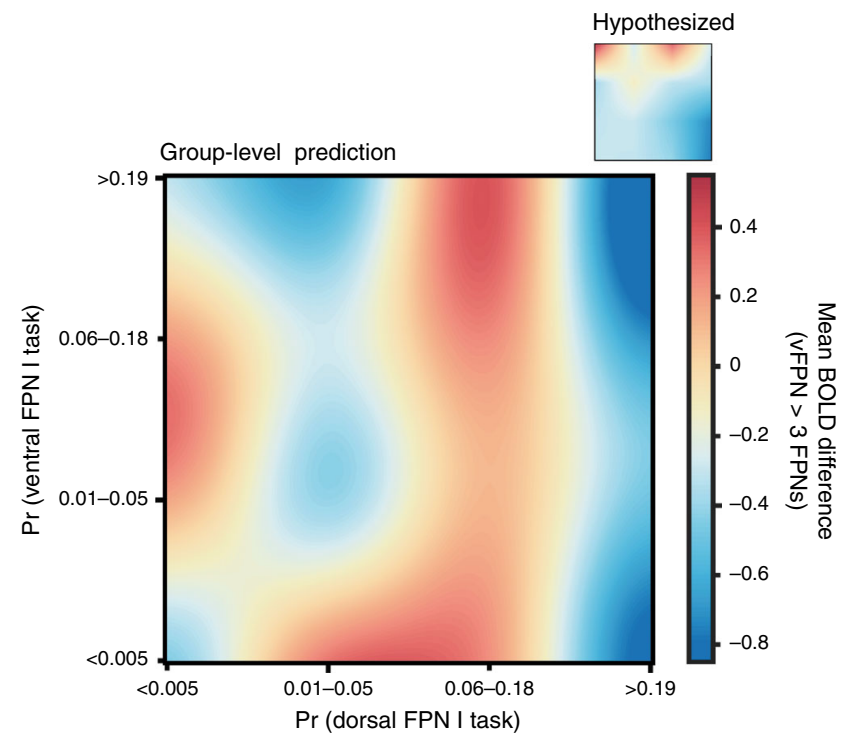


dorsal FPN > 3 other FPNs



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

ventral FPN > 3 other FPNs



Go/No-Go, Divided Auditory Attention, Imagined Movement, Passive Listening & Overt Reading

Summary

- High intra- and inter-subject reliability (subject-level results)
- Results deviate from previous meta-analyses and hypothesized functional labels for these FPNs
- FPNs should be functionally defined according to unique functional activity profile across multivariate task space
- Starting point for neurobiologically-derived cognitive taxonomy

Lorenz et al. *Nature Communications* 2018

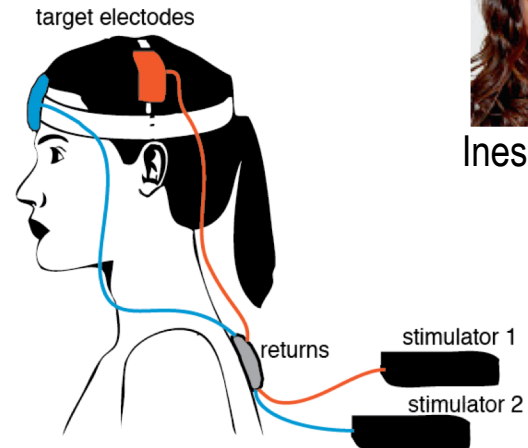
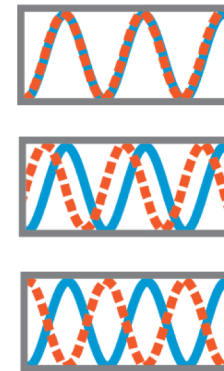
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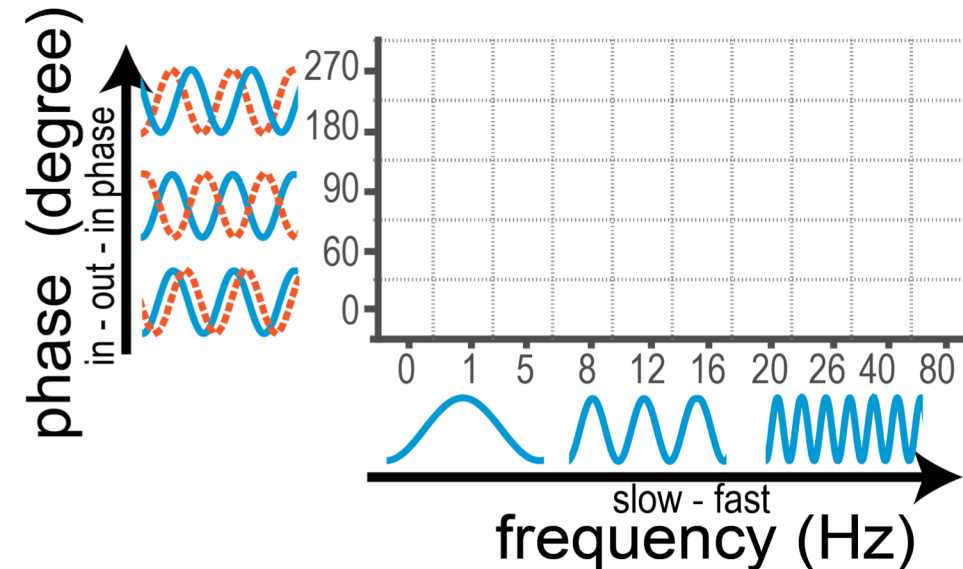
Transcranial alternating current stimulation (tACS)

■ Status Quo

- Ad hoc definition of frequency and phase
- Cohort testing



Ines Violante



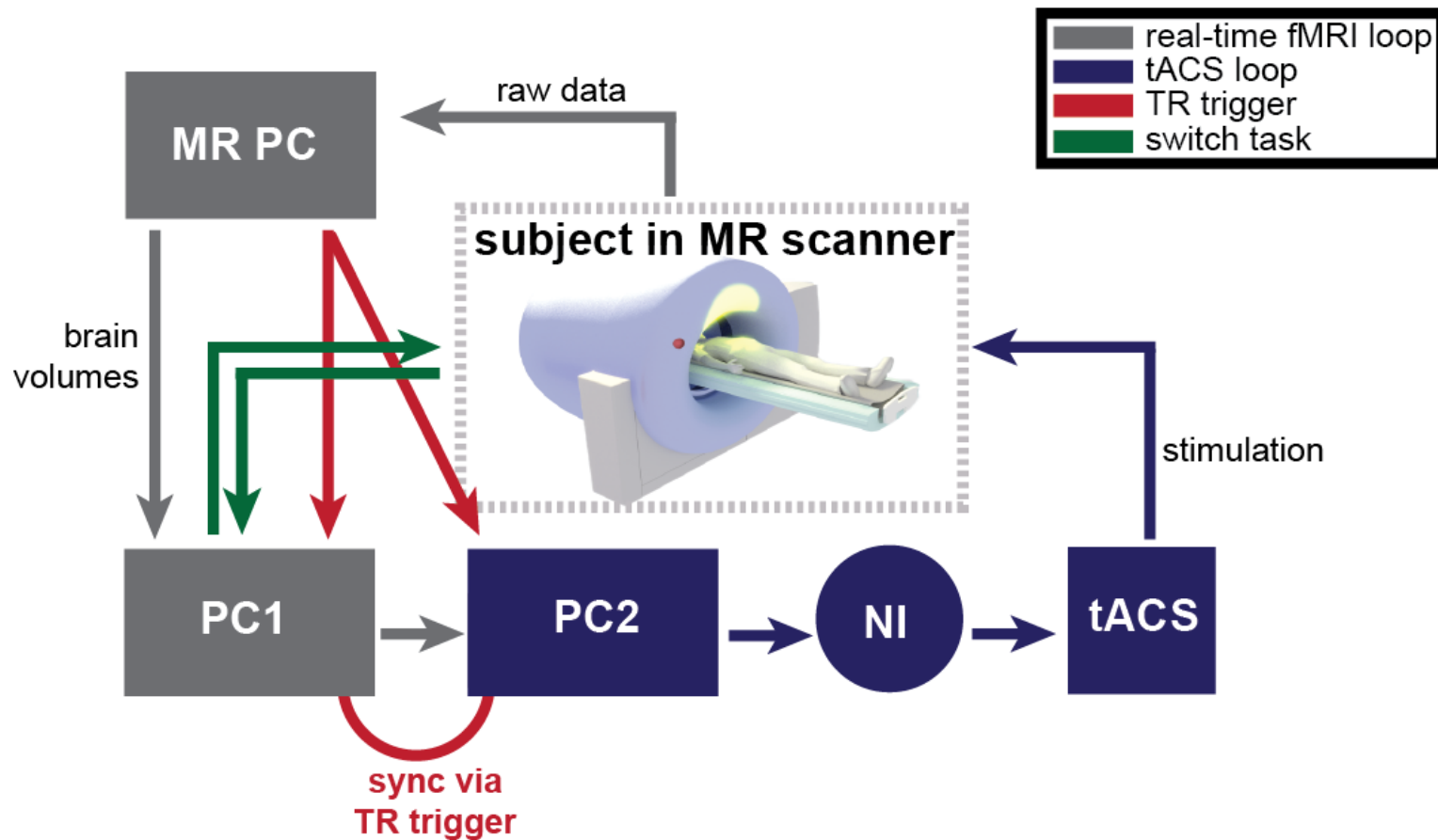
■ Limitation

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

Concurrent real-time fMRI/tACS



Ines Violante

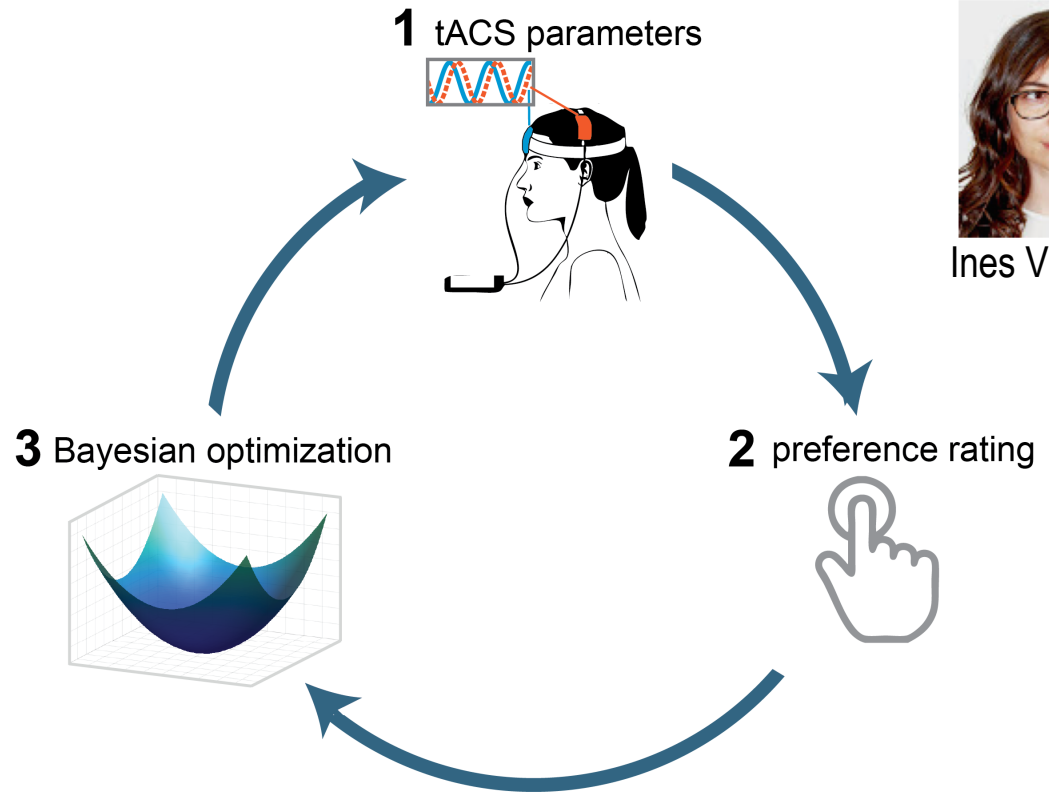


Lorenz et al. *PRNI* 2016
Lorenz et al. *in preparation*

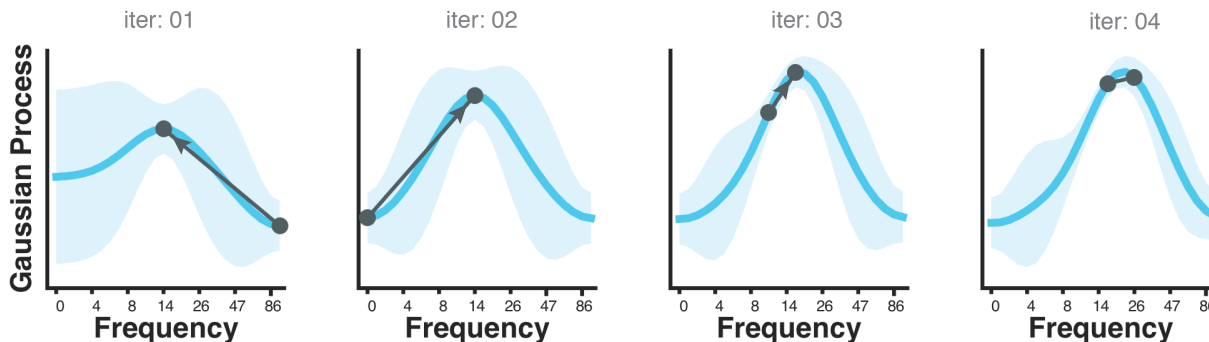
Phosphenes perception

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

Schutter et al. *NeuroImage* 2016

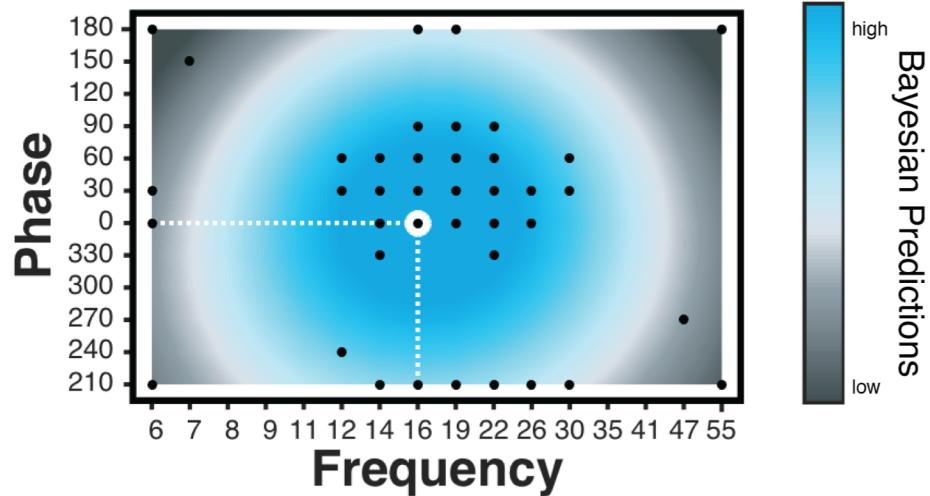
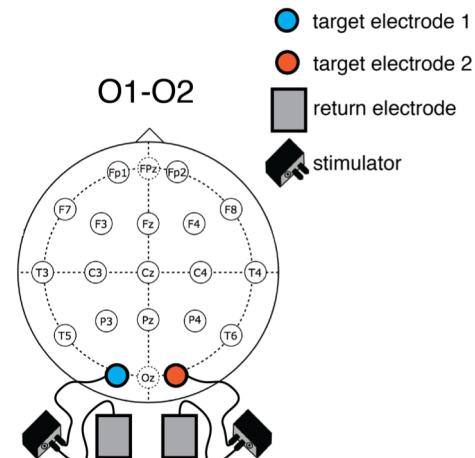
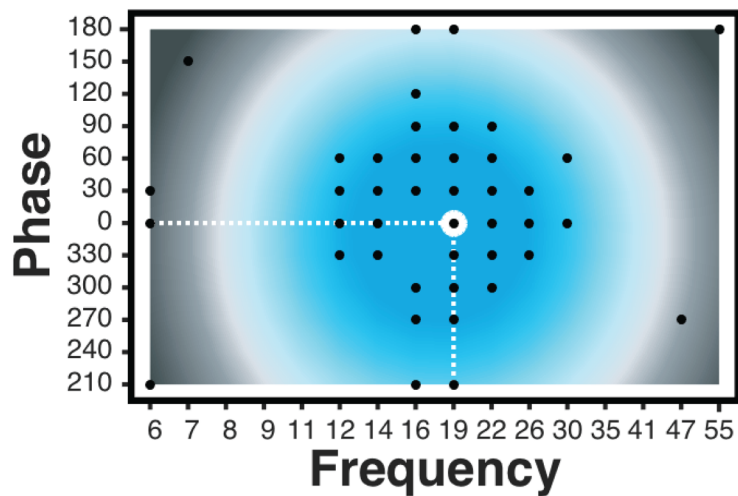
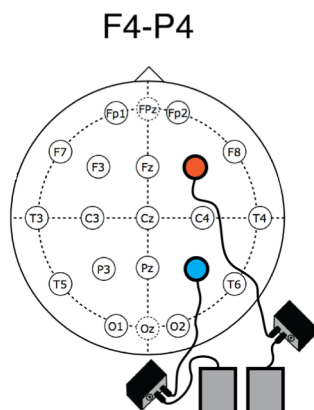


Ines Violante



Lorenz et al. *in submission*
(preprint on bioRxiv:150086)

Phosphenes perception



Ines Violante

Lorenz et al. *in submission*
(preprint on bioRxiv:150086)

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Biomarker discovery

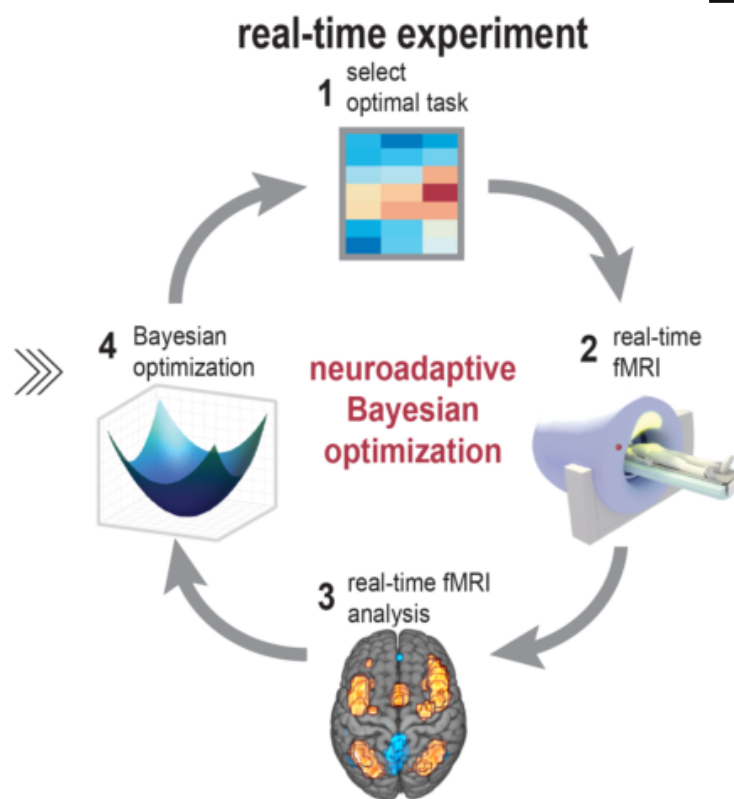
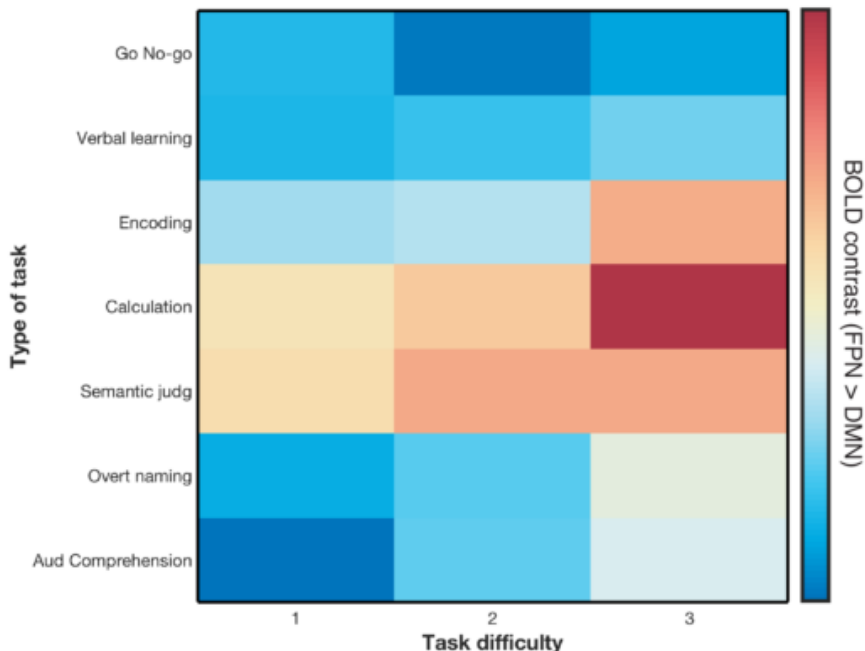


14 stroke patients + 14 controls



Fatemeh Geranmayeh

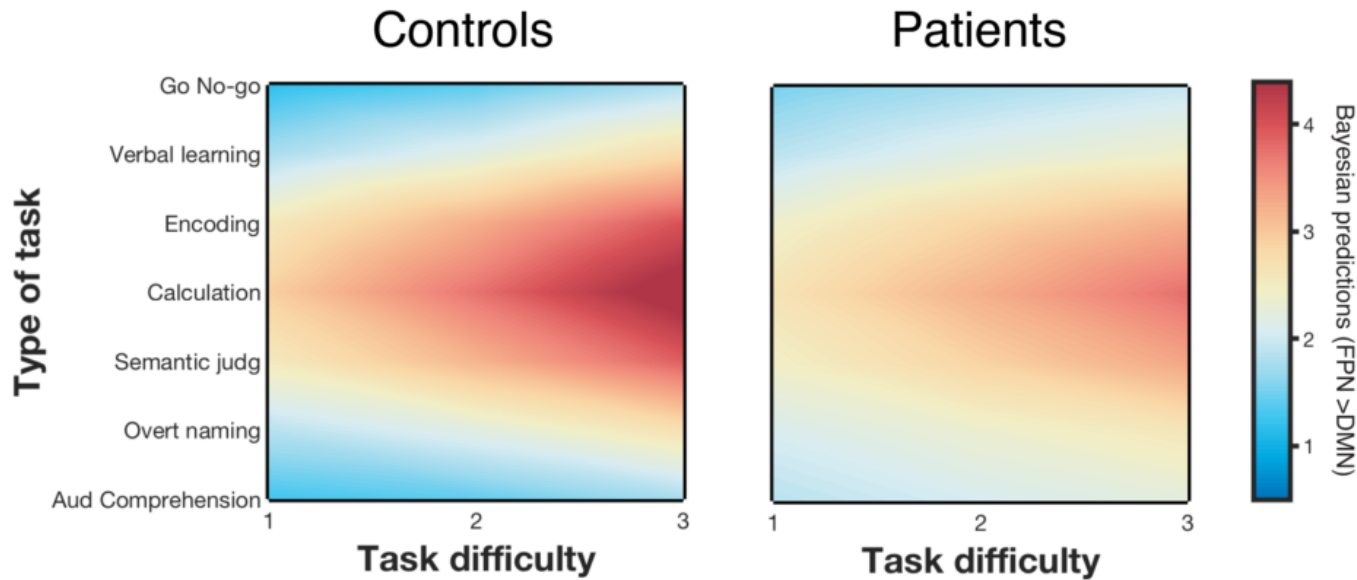
task space based on pilot data



Lorenz et al. *in preparation*

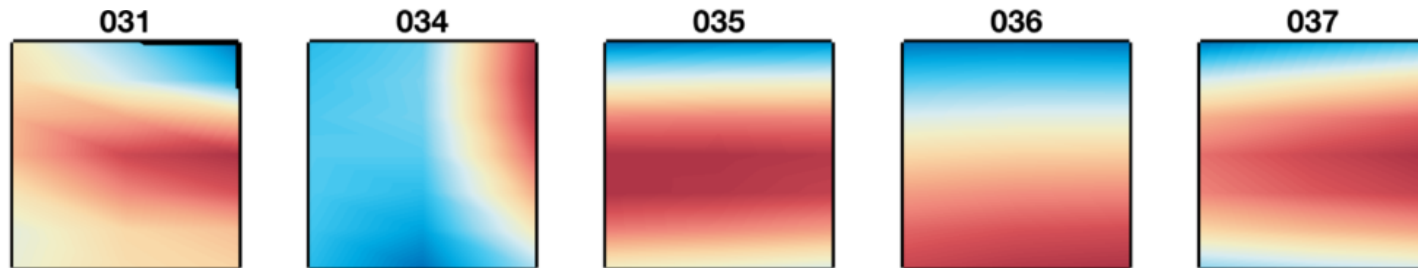
Biomarker discovery

Group results



Fatemeh
Geranmayeh

Subject results

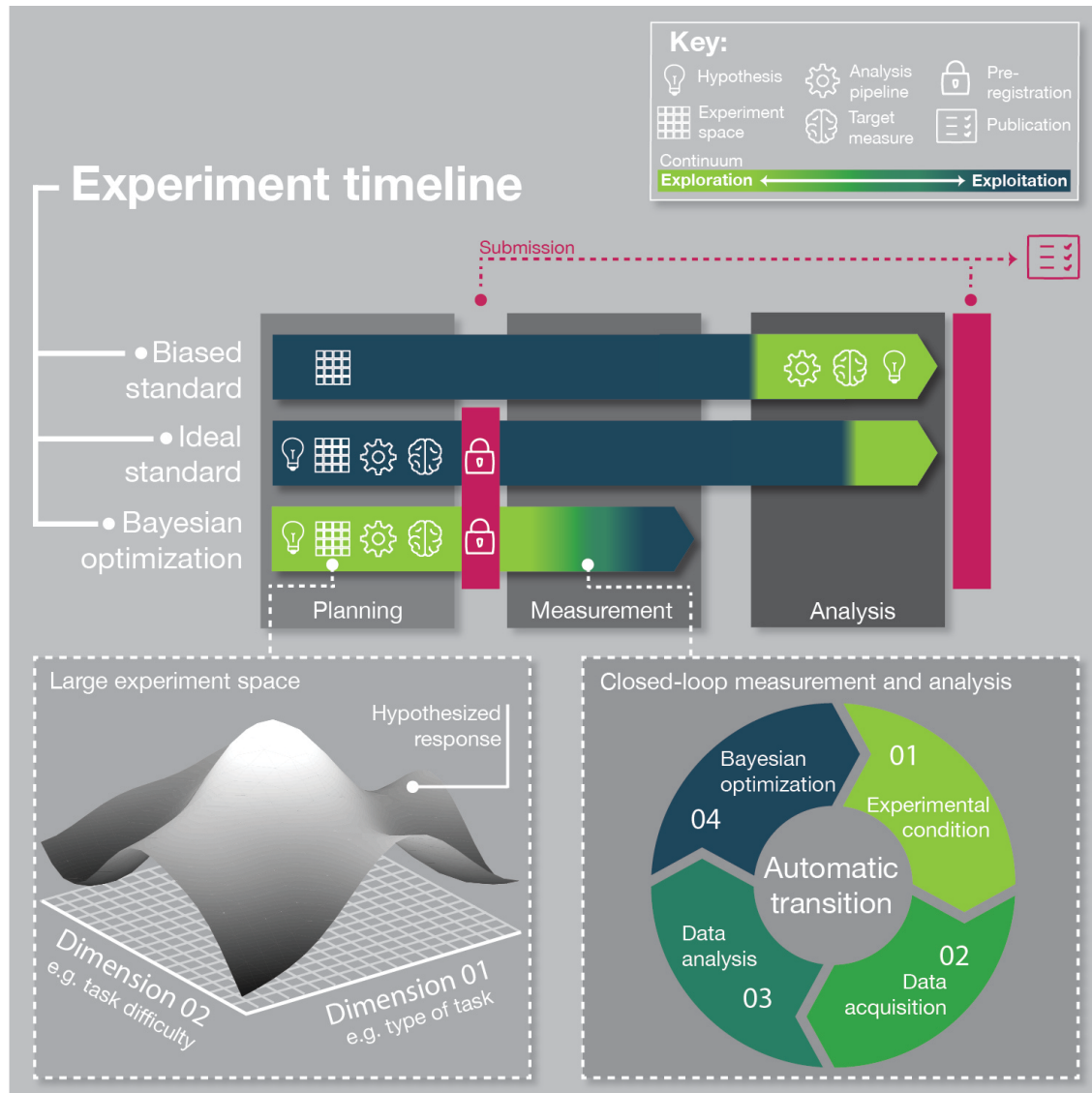


Lorenz et al. *in preparation*

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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

Registered Reports Special Initiative at NeuroImage

Guidelines for Authors:

“In place of conventional hypothetico-deductive methods, authors are welcome to propose more innovative analytic approaches such as [integration of Bayesian optimisation and hypothesis testing](#).”

Deadline: 19th March 2019

Acknowledgement

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wellcome

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London**

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Neuroimaging Laboratory **C³NL**

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



Gatsby Computational Neuroscience Unit

Ricardo P. Monti



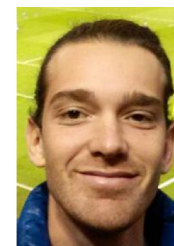
Rob



Adam



Ines



Ricardo



- Code

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

- Publications



Lorenz R, Violante IR, Monti RP, Montana G, Hampshire A, Leech R (2018). **Dissociating frontoparietal networks with neuroadaptive Bayesian optimization.** *Nature Communications*, 9:1227.

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*, 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.** *In submission* (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*

Lorenz R, Johal M, Dick F, Hampshire A, Leech R, Geranmayeh F. **Identifying individual functional profiles for a frontoparietal network in aphasic stroke patients.** *In preparation*

Questions



Questions/Feedback?



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