

NeuroLang

ERC Starting Grant Application 2017

Demian Wassermann, Ph.D.
Inria, France



NeuroLang's PI: Demian Wassermann



UBA



Education

2006

Inria



- 2003-2005: **MsC. in Computer Sciences**, Universidad de Buenos Aires

- 2006-2010: PhD in **Signal Processing with Applications to Neuroimaging**
(with highest honours)

2010

Harvard



Research Positions

- 2010-2013: Post-Doc **Harvard Medical School** and Brigham and Women's Hospital
- Since 2014: Research Scientist, Inria

2014

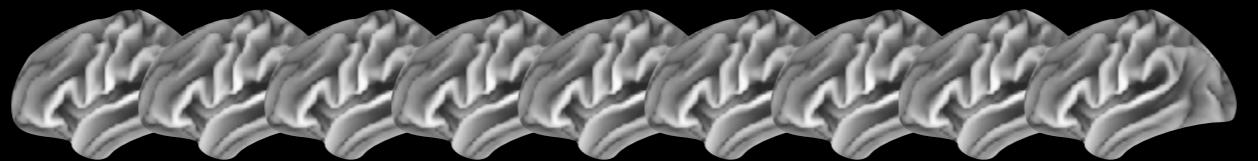
Inria



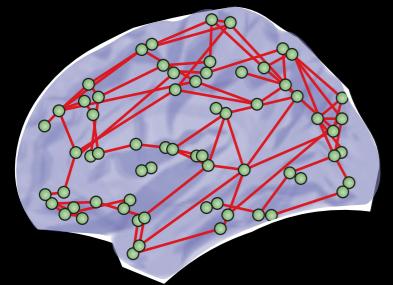
Early Achievements and Impact

- 15 journal publications; 30 conference publications; >550 citations
(25% citation increase in the last year)
- Mentoring 2 post-docs; 3 PhD students; 9 masters' students
- Raised more than **500k€ in research funding** as PI or co-PI since 2014

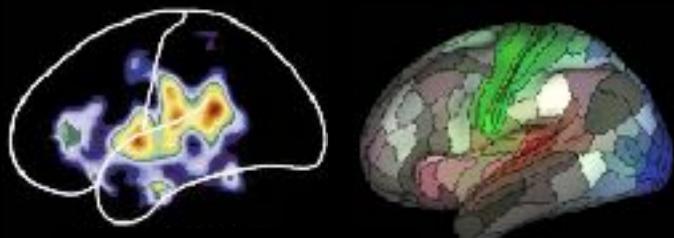
Large Scale / Big Data Studies



Network theory



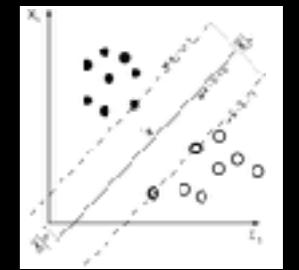
Neuroanatomy



Shape Spaces



Machine Learning

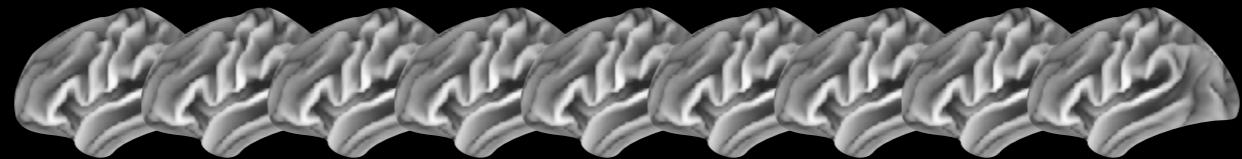


Genetics

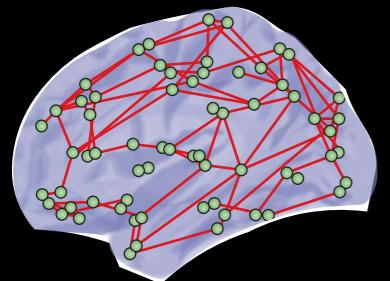


x

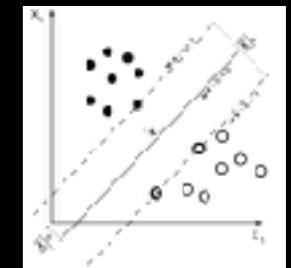
Large Scale / Big Data Studies



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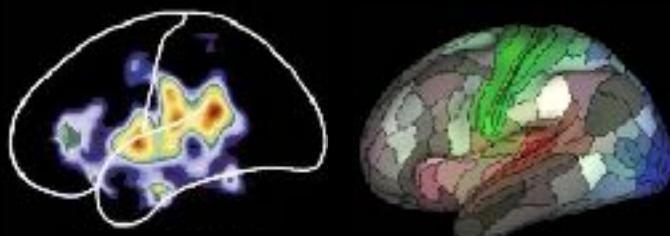


Machine Learning



There is a growing need for a **unifying formalism**
across all sub-areas of
computational neuroanatomy

Neuroanatomy



Genetics



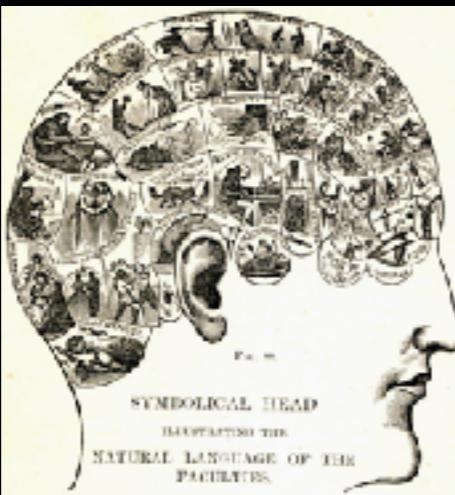
Shape Spaces



x

Current Neuroanatomy

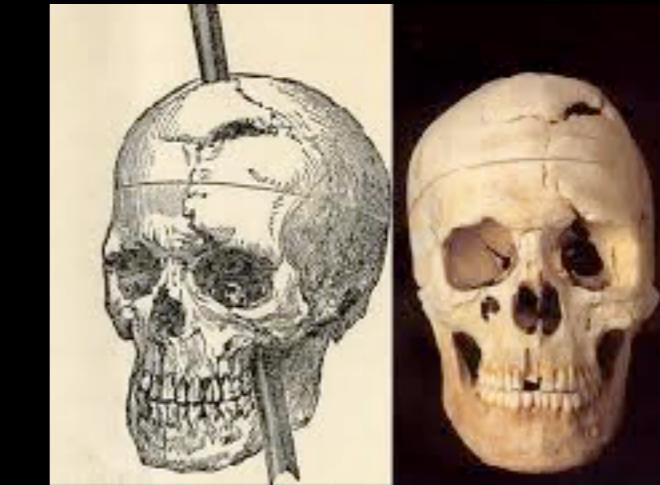
Bicentennial quest for the relationship between brain location and function



Francis Gall [1810]

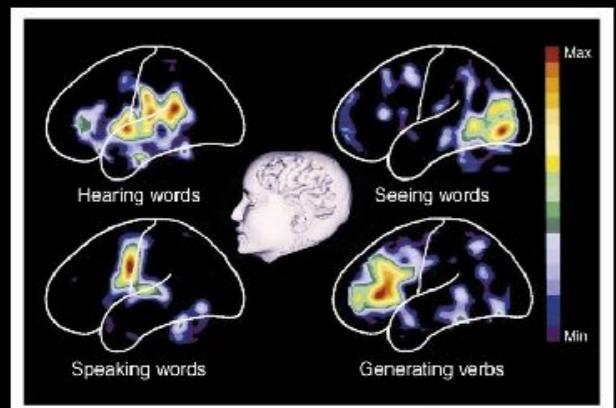


Paul Broca [1861]

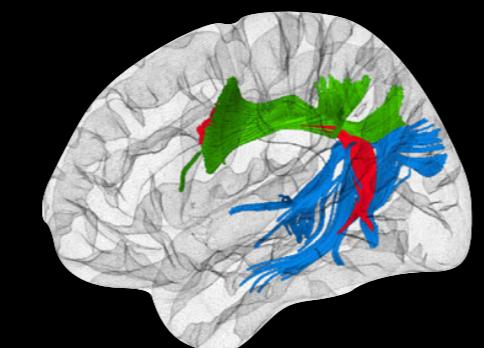


Phineas Gage case [1848]

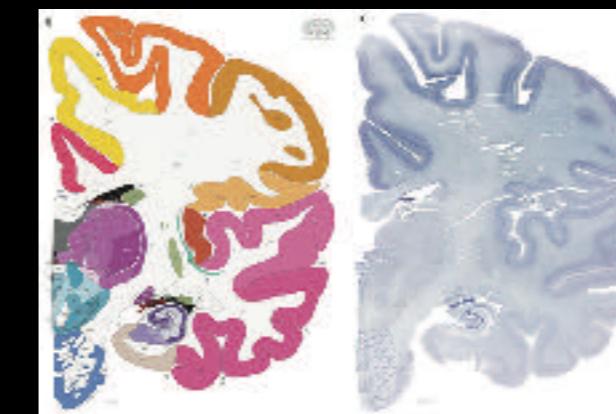
Advent of Brain Imaging: resurgence of the spatial specialization paradigm



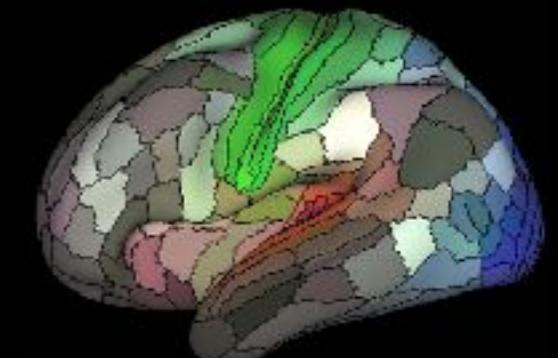
Activity Related to Function
Petersen et al [1988]



Structural Connectivity
Wassermann et al [2016]



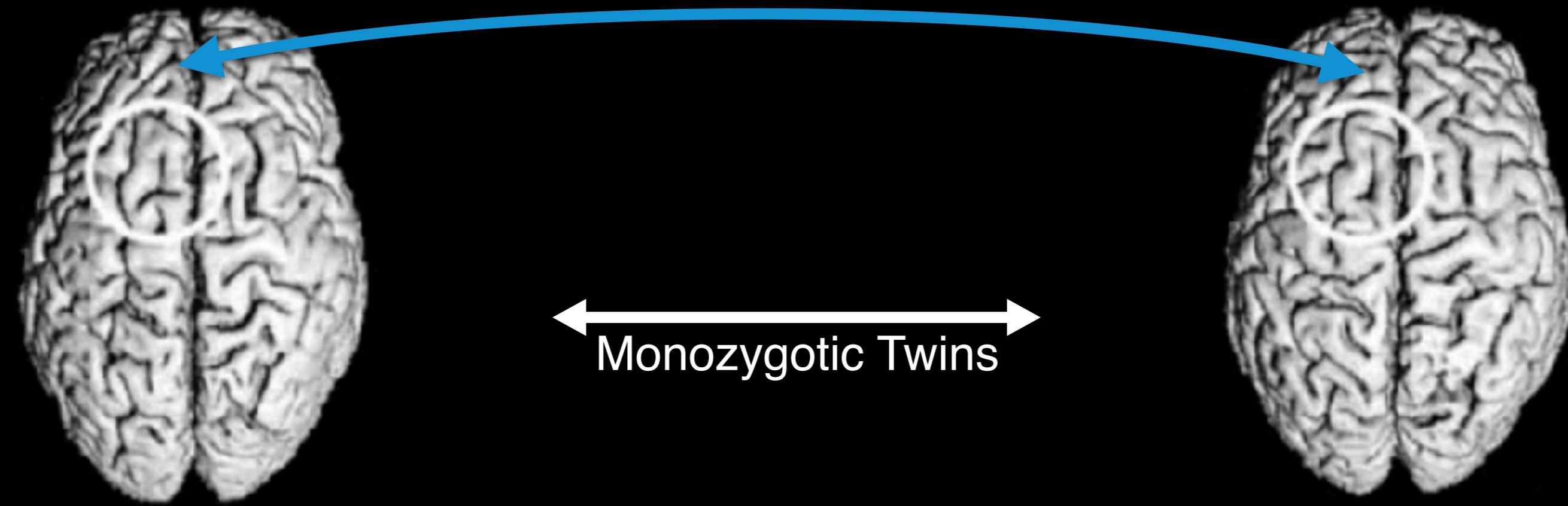
Cellular Structure
Ding et al [2016]



Multi-Modal
Glasser et al [2016]

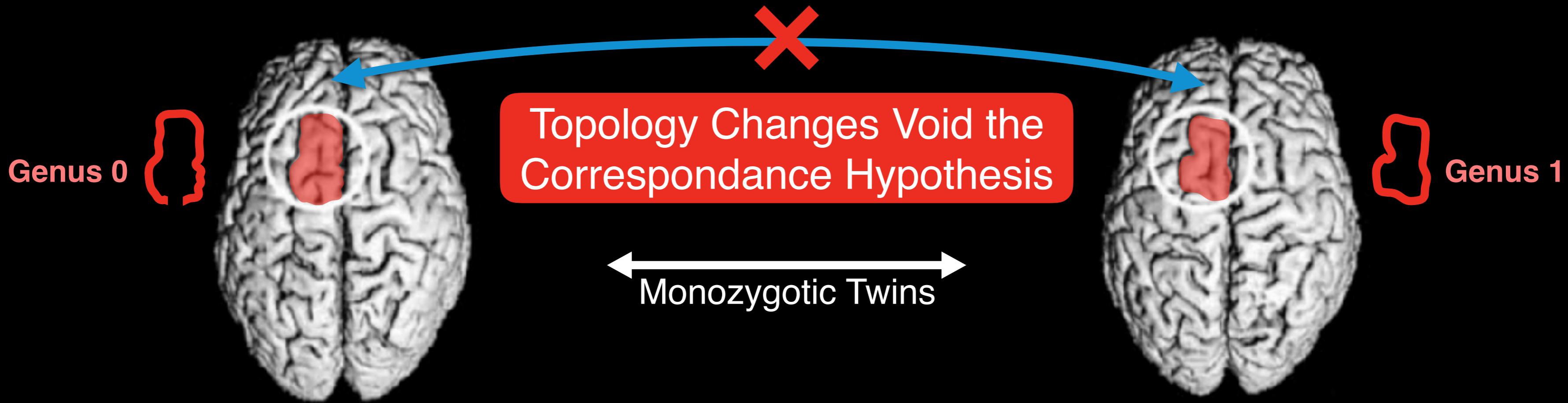
Spatial Specialization in Neuroanatomy

The current driving hypothesis is the existence of a
Spatial Correspondance across Individuals



Spatial Specialization in Neuroanatomy

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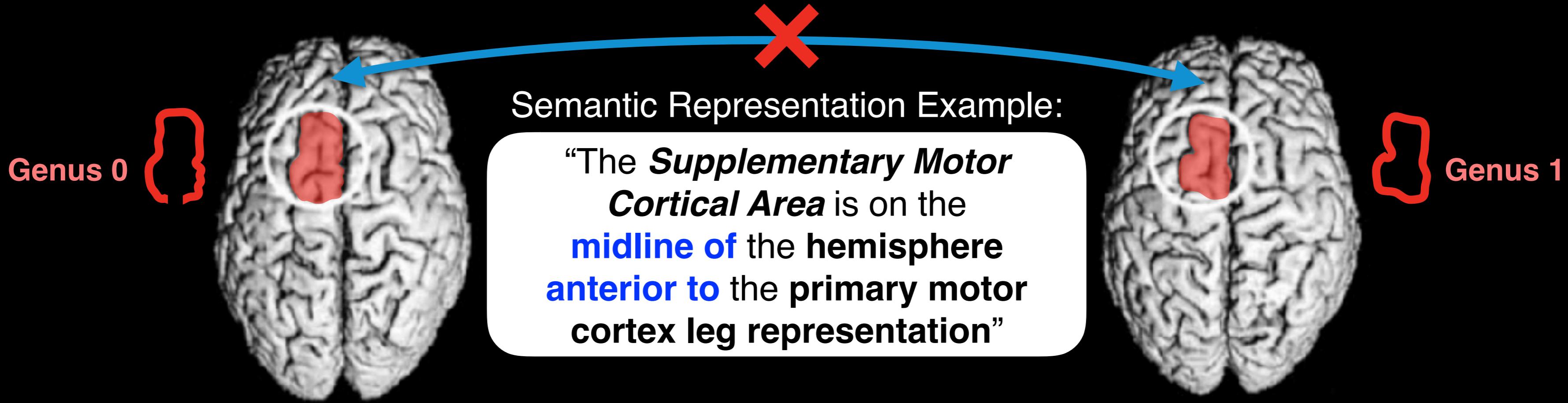


Spatial correspondance misrepresents individual characteristics **limiting 1000s of studies yearly**.

We need a new paradigm:
Semantic Representations of Neuroanatomy

Spatial Specialization in Neuroanatomy

The current driving hypothesis is the existence of a
Spatial Correspondance across Individuals



Semantic Representation Example:

“The **Supplementary Motor Cortical Area** is on the **midline of the hemisphere anterior to** the primary motor cortex leg representation”

Spatial correspondance misrepresents individual characteristics **limiting 1000s of studies yearly**.

We need a new paradigm:
Semantic Representations of Neuroanatomy

NeuroLang: Semantic Neuroanatomy

The **main challenge** is to develop a **specific language** to unify neuroanatomy and apply it to **solve cognitive and neurological problems**

We propose to develop a Neuroanatomy Domain Specific Language (DSL).

DSLs are computer languages that:

- **Express solutions** in the idiom of the problem domain.
- **Produce programs** that are reusable & self documenting.
- **Embody domain knowledge**, enabling its conservation.

Semantic Representation of White Matter

“The **arcuate fascicle connects** the frontal and temporal lobes **coursing through** the parietal lobe”

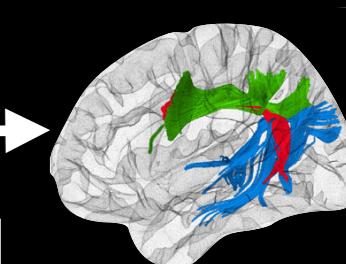


White Matter DSL Query

arcuate_fascile is
white matter tract **and**
connects temporal lobe **with** frontal lobe
and courses **through** parietal lobe


$$\{x : x \in \mathcal{W} \leftarrow p_1(x, C_1, C_2) \wedge p_2(x, C_3)\}$$


White Matter DSL Engine



[Wassermann et al 2013, 2016]

The NeuroLang DSL

“What cortical regions are involved in language processing?”



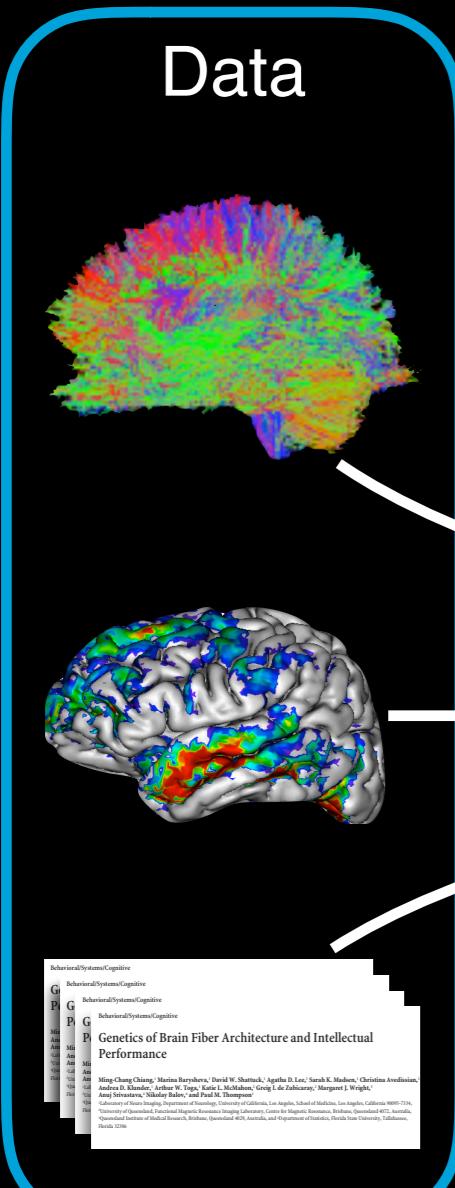
“What were the most probably active contiguous regions under the simple object naming task?”



NeuroLang Query

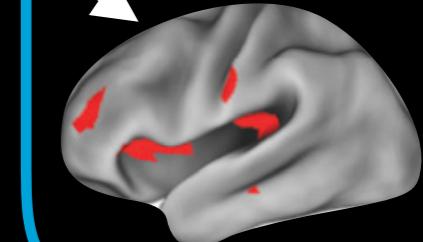
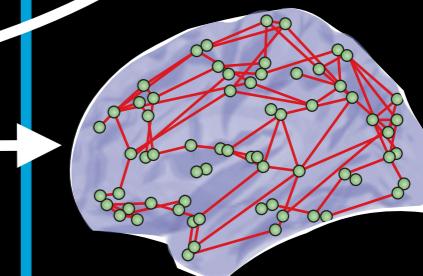
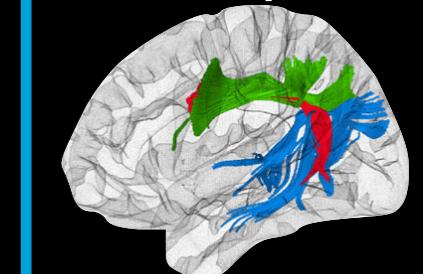
**Probability(x) > 0.9 where x are contiguous regions and
probability_is_active(x , object_naming_task) > 0.95**

$$\{x : \mathbb{P}(x \subset \mathcal{B} | p_1(x, C)) > 0.9\}, C \in \mathcal{B} \cup \mathcal{C}$$



NeuroLang Engine

Calculation Output



X

NeuroLang: Challenges

The **main challenge** is to develop a **specific language** to unify neuroanatomy and apply it to **solve cognitive and neurological problems**

For this, NeuroLang will tackle the **specific challenges**:

1. Design a neuroanatomical Domain-Specific Query Language
2. Represent neuroanatomical organisation and data.
3. Enable large-scale statistical Inference in a neuroanatomical DSL.
4. Accelerate reproducible research and applications in clinical and cognitive research.

NeuroLang: Challenges

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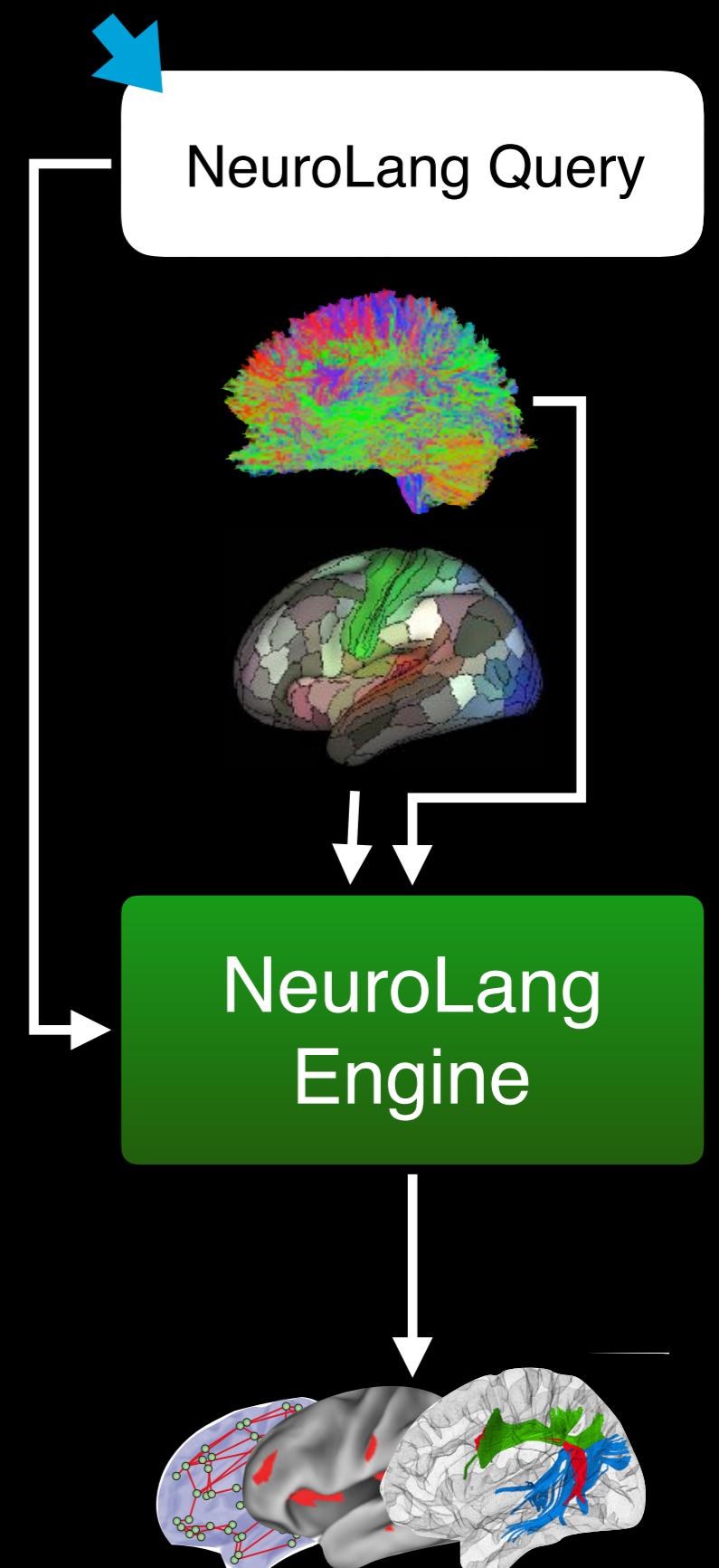
Challenge 1: Design a Neuroanatomical DSL

Approach

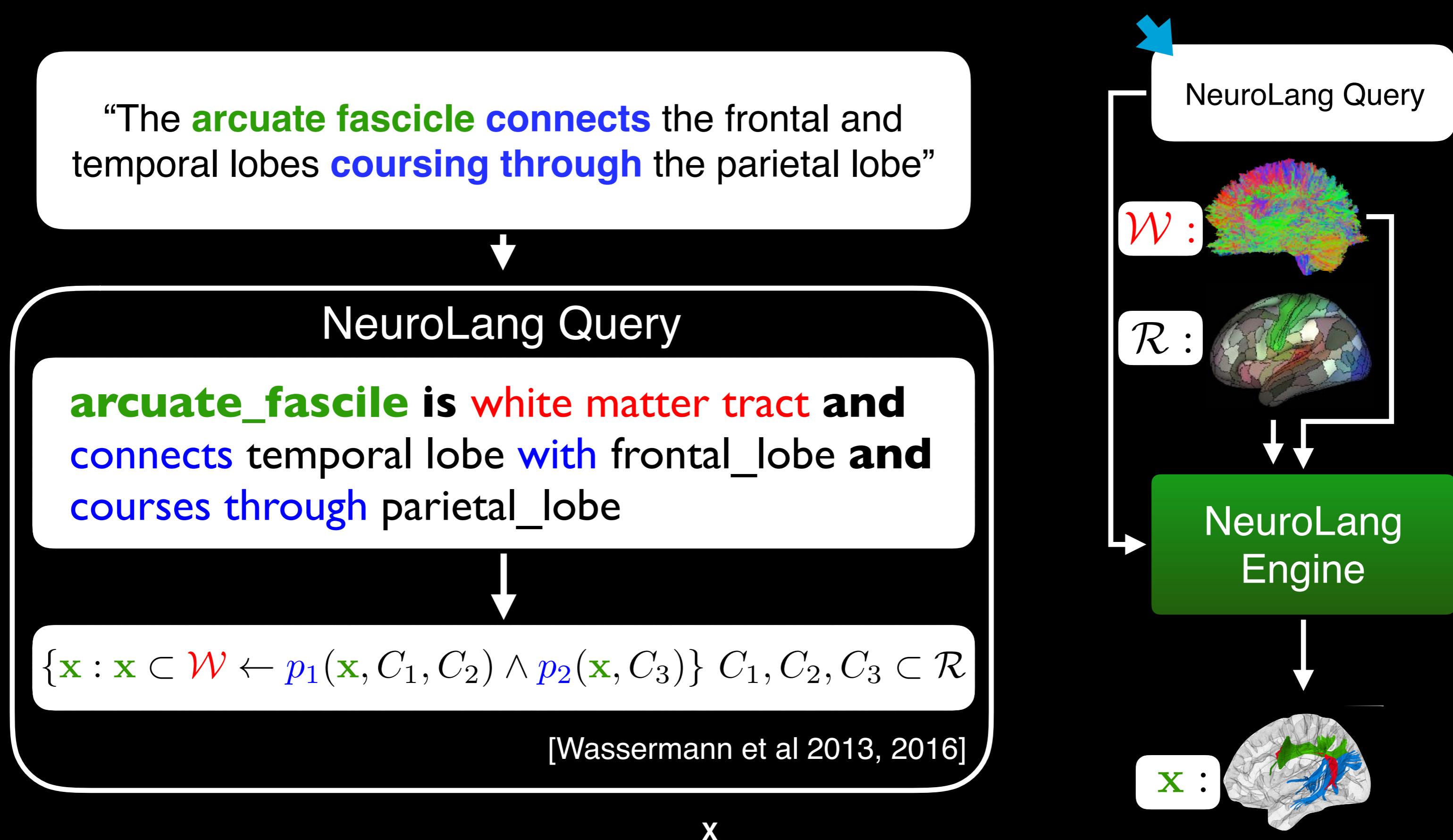
1. **Catalogue** the neuroanatomist's lexical phrases for characterising brain structures, connectivity, and function.
2. **Formalise** the neuroanatomist's lexicon as queries and design the corresponding query DSL.
3. **Enable** this DSL to use these queries to perform neuroimaging data analysis within and across databases.

Outcome

- DSL-based **formalisation of brain anatomy as a query language**.
- **Identification of neuroanatomical regions** beyond spatial matching.

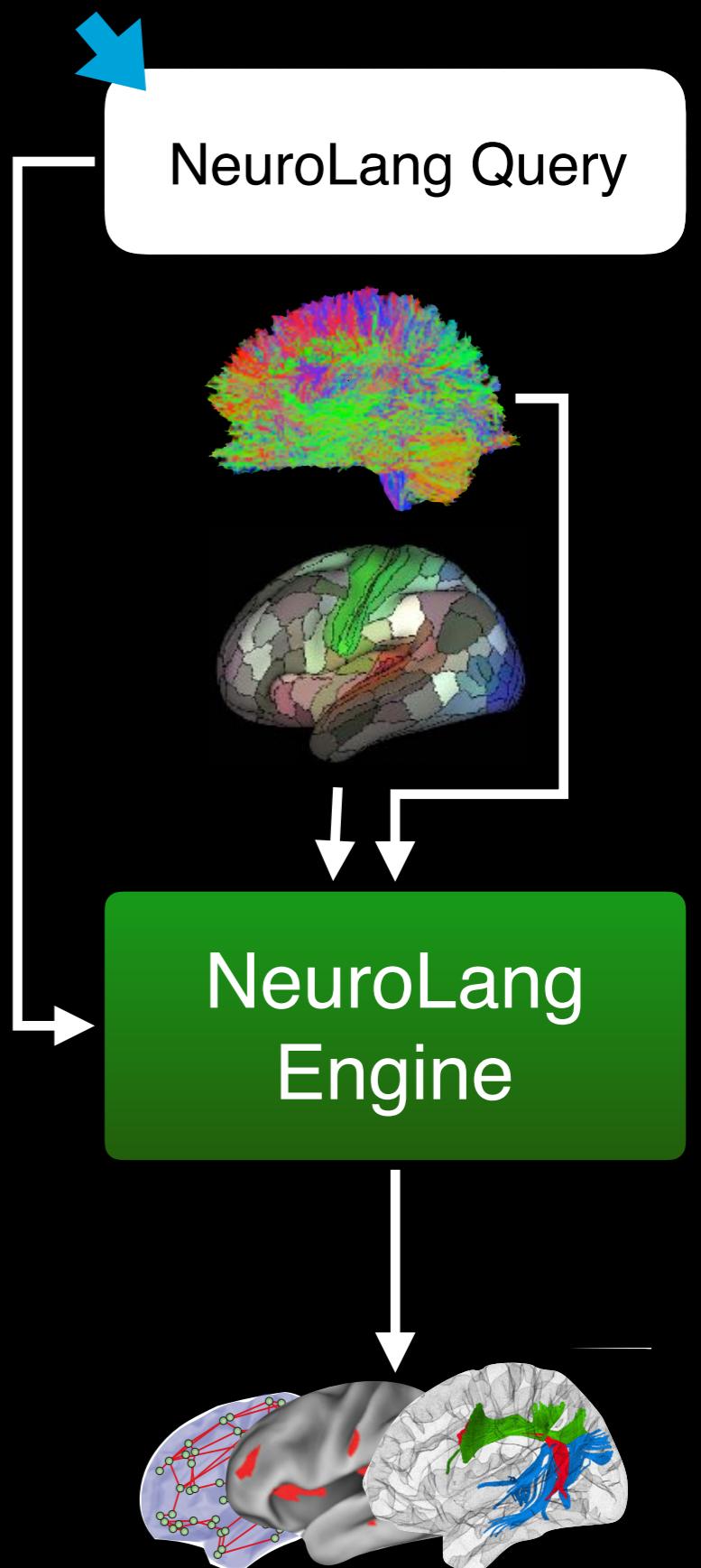


Challenge 1: Design a Neuroanatomical DSL, an Example



Challenge 1: Design a Neuroanatomical DSL: Language Properties

- **Modular and Compositional**: avoid unnecessary code duplication.
- **Well Behaved Formal Semantics**: to develop type systems and improve code quality.
- **Tractable**: solving queries should be feasible with the computing power of current architectures.
- **Programmable**: expressing neuroanatomical definitions should require little effort.

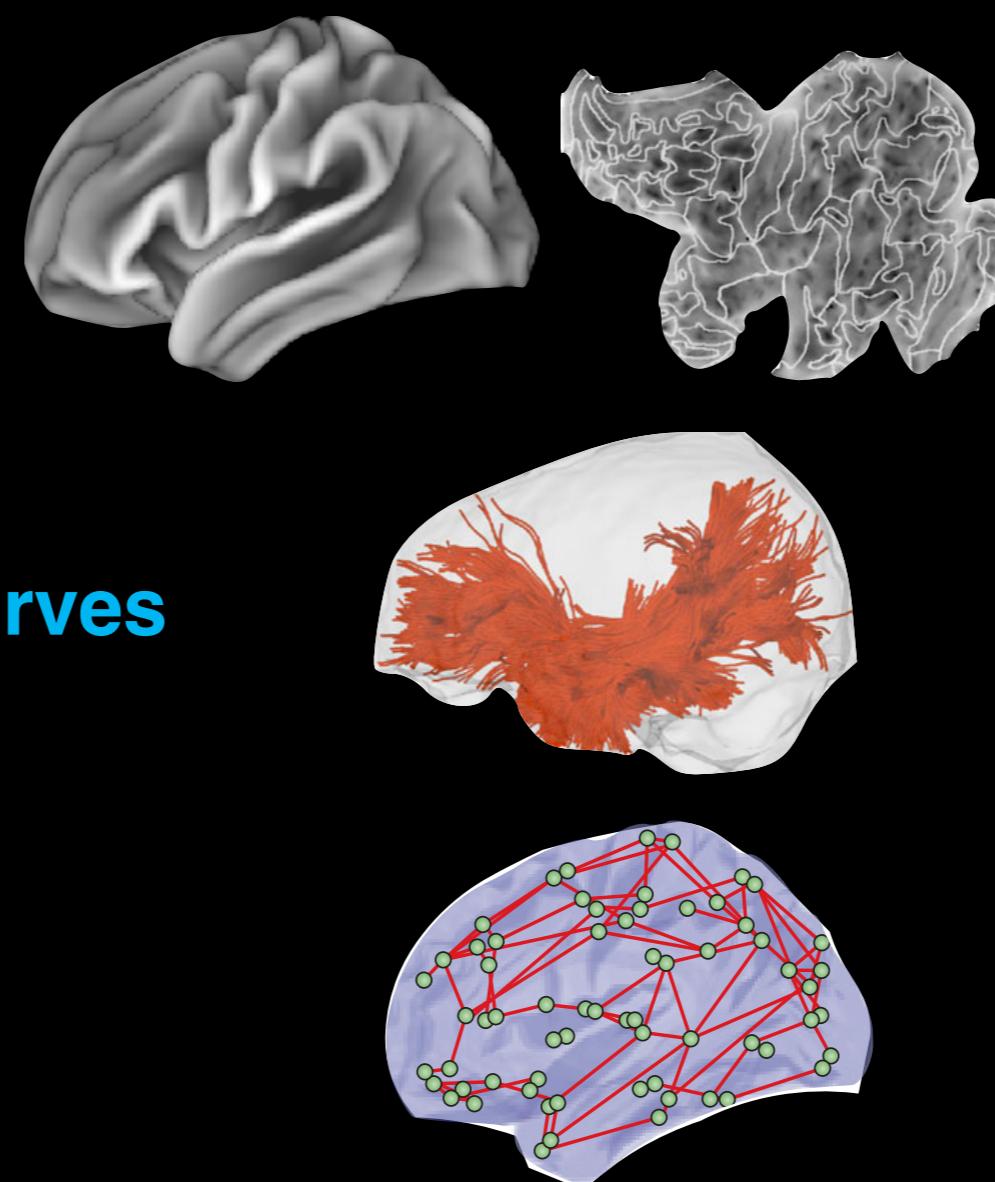


Challenge 2: Represent Neuroanatomical Organisation

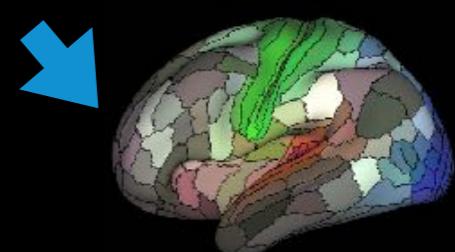
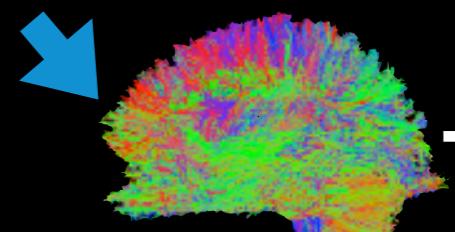
Main Problem: Querying across Topologies

The brain is conceived as a **union of different entities with distinct topologies**:

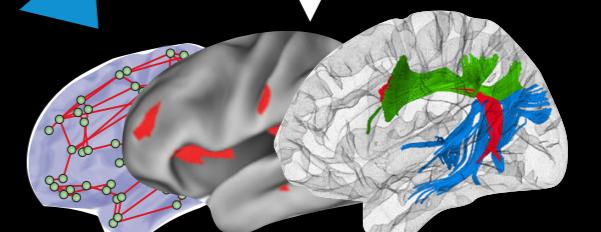
- The cortex is usually represented as a **3D-surface** or a **flat map**
- The white matter as a set of **3D-curves**
- Brain Connectivity as a **graph**



NeuroLang Query



NeuroLang Engine



Challenge 2: Represent Neuroanatomical Organisation

Approach

1. Elaborate a **type system** and representation for **spatial structures**

$$\{\mathbf{x}_1 : \mathbf{x}_1 \in \mathcal{B} \leftarrow p_1(\mathbf{x}_1, C_1) \dots p_N(\mathbf{x}_1, C_N)\}, C_1 \dots C_N \subset \mathcal{B} = \{\mathcal{W}, \mathcal{R}, \dots\}$$



$$\{(\mathbf{x}_1, \mathbf{x}_2) : \mathbf{x}_1, \mathbf{x}_2 \in \mathcal{B} \leftarrow p_1(\mathbf{x}_1, \mathbf{x}_2, C_1) \dots p_N(\mathbf{x}_1, \mathbf{x}_2, C_N)\}$$

2. Expand the **type system** to functions on **collections of structures**

$$\{f(\mathbf{x}, C) : \mathbf{x} \subset \mathcal{B} \leftarrow p_1(\mathbf{x}, C_1) \dots p_N(\mathbf{x}, C_N)\}$$

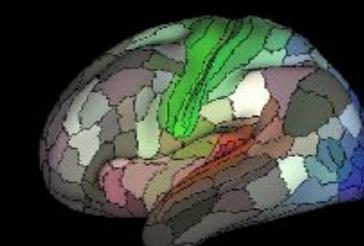
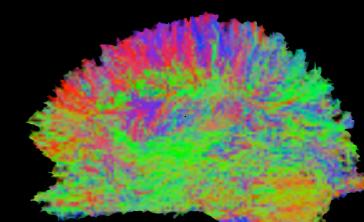
3. Develop **efficient techniques** to evaluate $p_i(\mathbf{x}, C_i)$
This will require **indexing across different topologies**.

Outcome

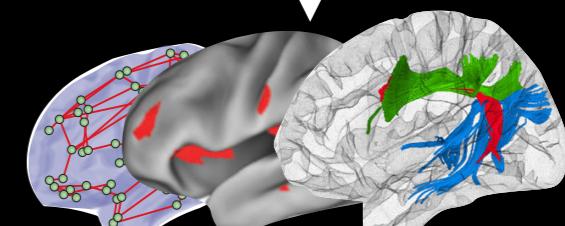
A DSL capable of formalising and inferring **relationships between objects of different topologies**.

X

NeuroLang Query



NeuroLang
Engine



Challenge 4: Accelerate Reproducible Research and Applications in Clinical and Cognitive Research

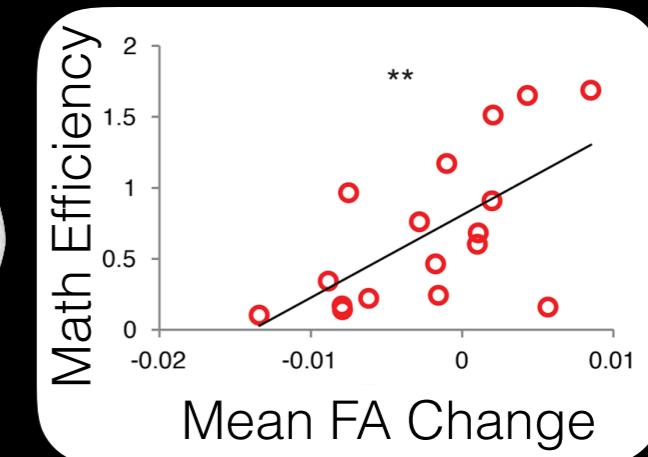
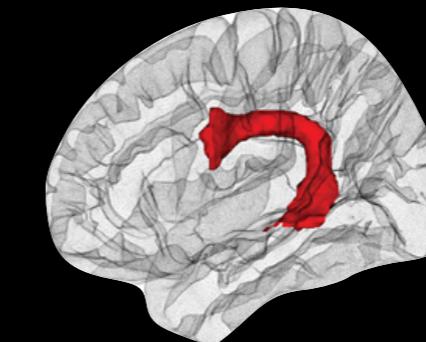
Brain Sciences Experience a Reproducibility Crisis

- Neuroanatomical protocols are **study-specific** and **hard to reproduce**.
- This makes **results harder to replicate**.



Approach

- Early application of NeuroLang to **cognitive and clinical problems**.
- Development of **high-quality software** for NeuroLang.



[Jolles, Wassermann et al 2015]

Outcome

A DSL and high quality software to **facilitate reproducibility** in neuroanatomical, cognitive and clinical research.

NeuroLang At a Glance

The main challenge is to develop a specific language to unify neuroanatomy and apply it to solve cognitive and neurological problems

Risks

- Efficient handling of neuroanatomical equations and structures
- Stochastic modelling of neuroanatomical data for large scale statistical inference

Impact

- A new generation of computational tools that will unify neuroanatomy
- New approaches for querying databases of topologically and dimensionally varied large objects
- Novel tools for understanding neurological diseases

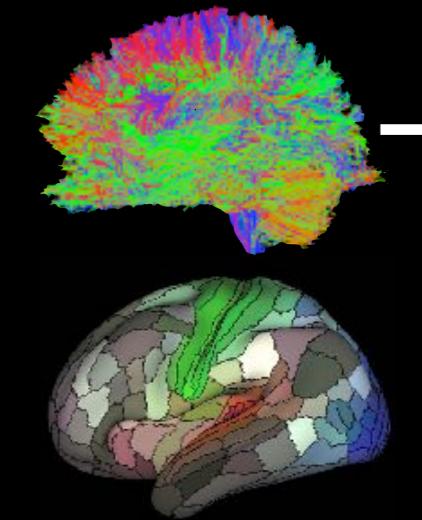
The NeuroLang Team

- PI + 1 Junior Researcher + 4 PhD Students + 1 Engineer

Selected Collaborations

- Center for Brain Morphometric Analysis, Harvard Medical School, USA
- Stanford Cognitive & Systems Neuroscience Lab, Stanford, USA
- Core Facility for Clinical Research, Brain and Spine Institute, France
- Research Institute on Fundamental Informatics, CNRS / Paris 7, France

NeuroLang Query



NeuroLang Engine

