

# Comparing Perturbation Models for Evaluating Stability of Neuroimaging Pipelines

Gregory Kiar, on behalf of

Pablo de Oliveira Castro, Pierre Rioux, Eric Petit, Shawn T. Brown, Alan C. Evans,  
Tristan Glatard



McGill



UNIVERSITÉ  
Concordia  
UNIVERSITY

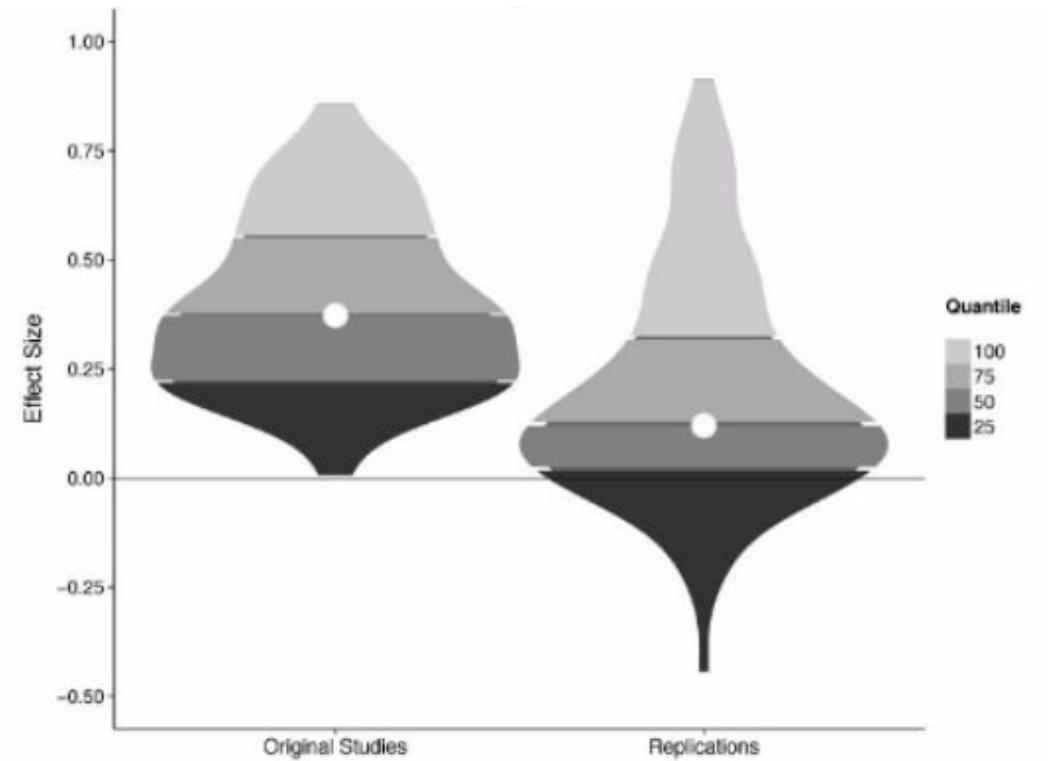
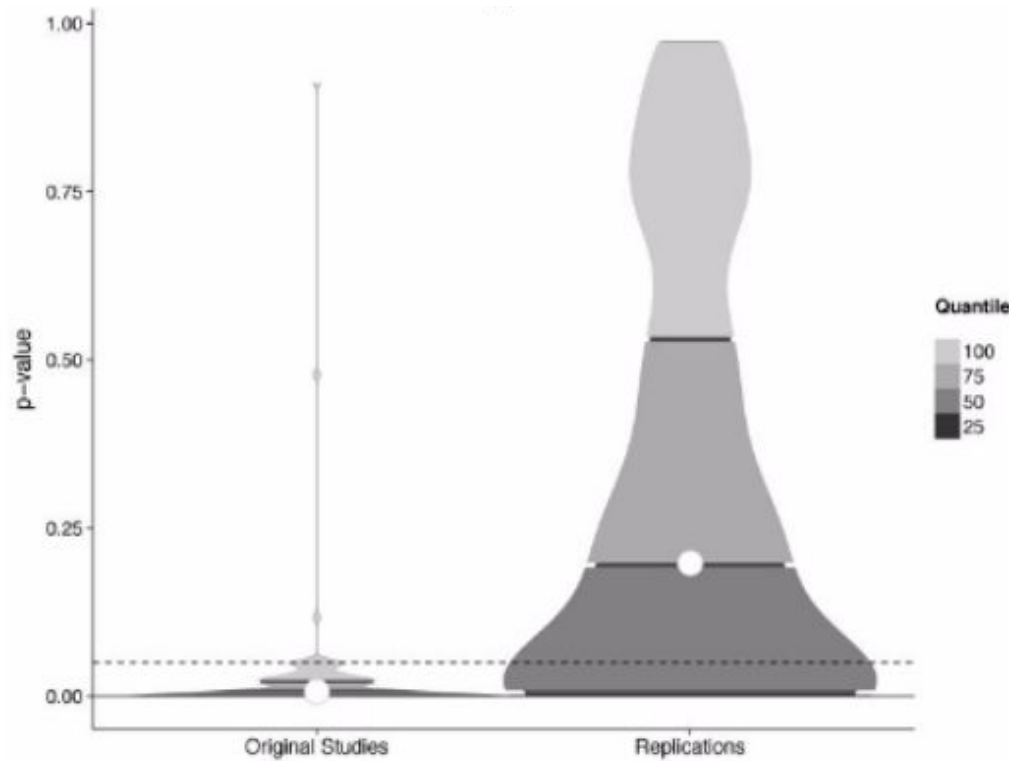
Exascale   
computing research

# Overview

- Topics in reproducibility
- Operationalizing Stability
- Evaluation of Neuroimaging Pipelines



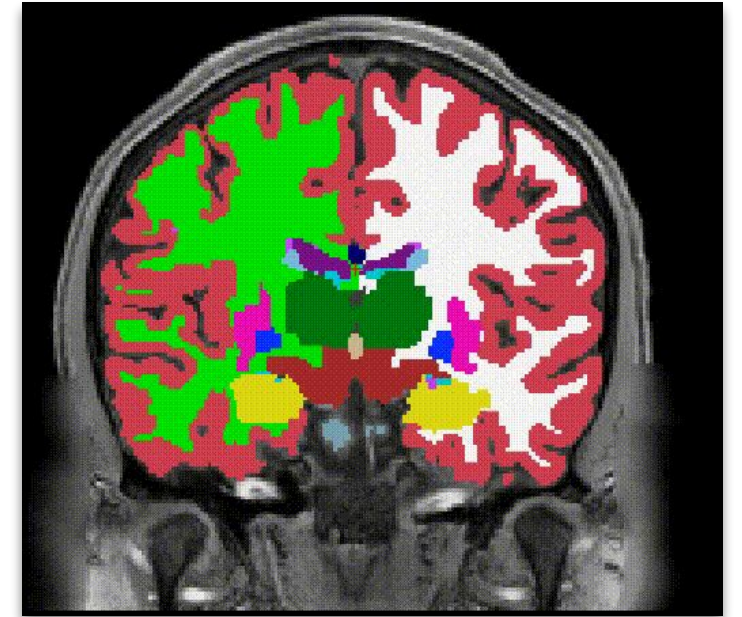
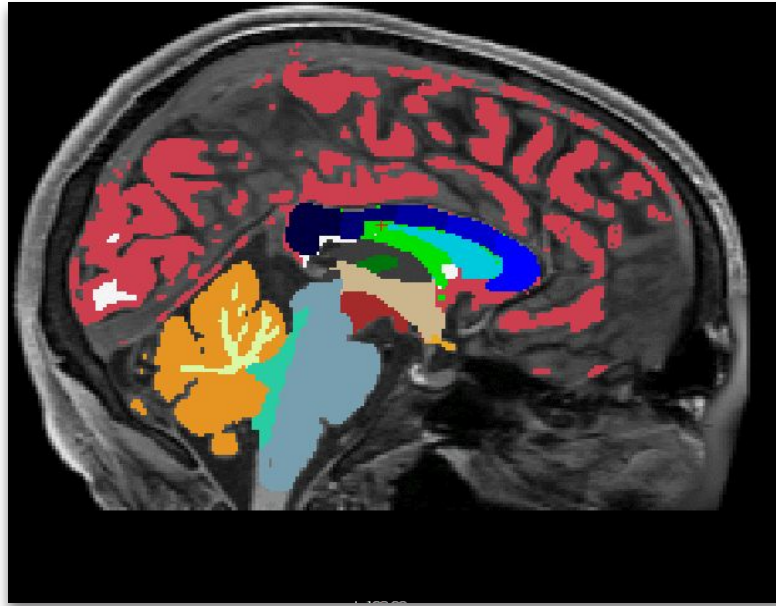
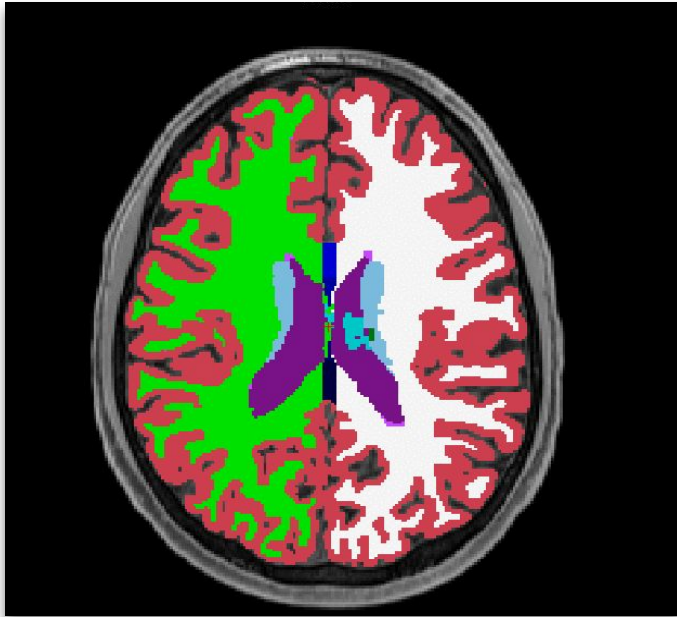
# Reproducibility is Measurable



(Open Science Collaboration, 2015)

# Contributing Factor: OS

CentOS 6 vs CentOS 7



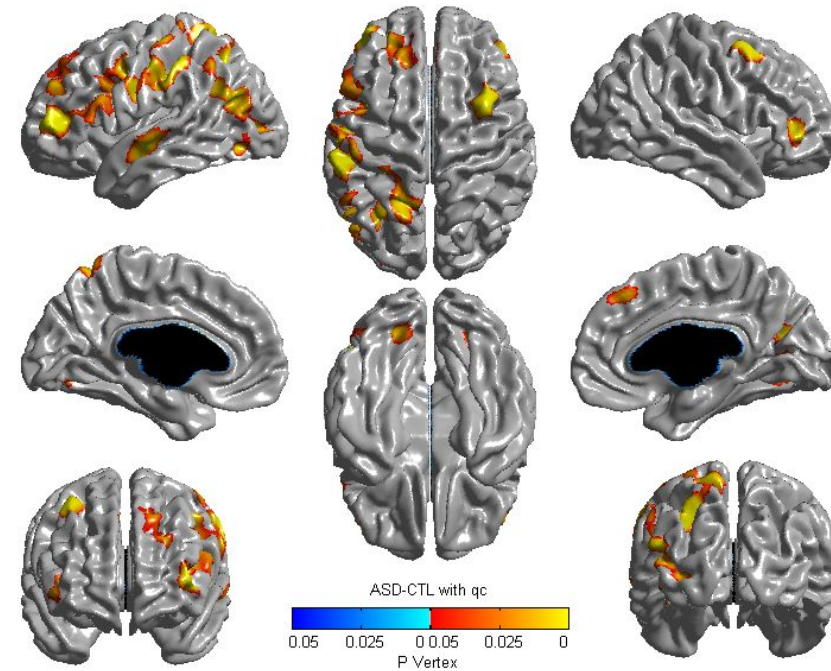
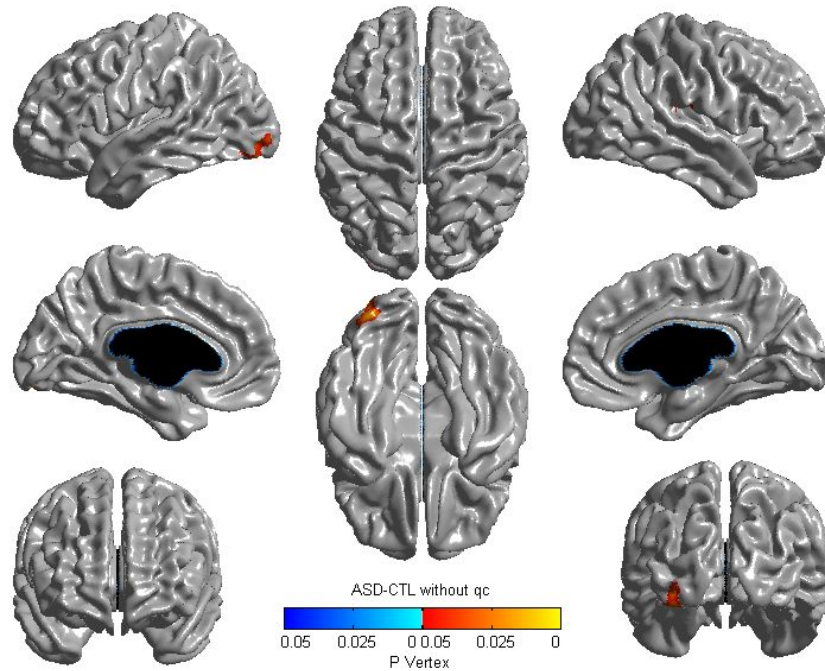
(Scaria, 2017)

(Scaria et al., 2017)  
(Lewis, 2017)

# Contributing Factor: Data Quality

Full Dataset (N ~ 1100)

vs QC'd Dataset (N ~ 400)



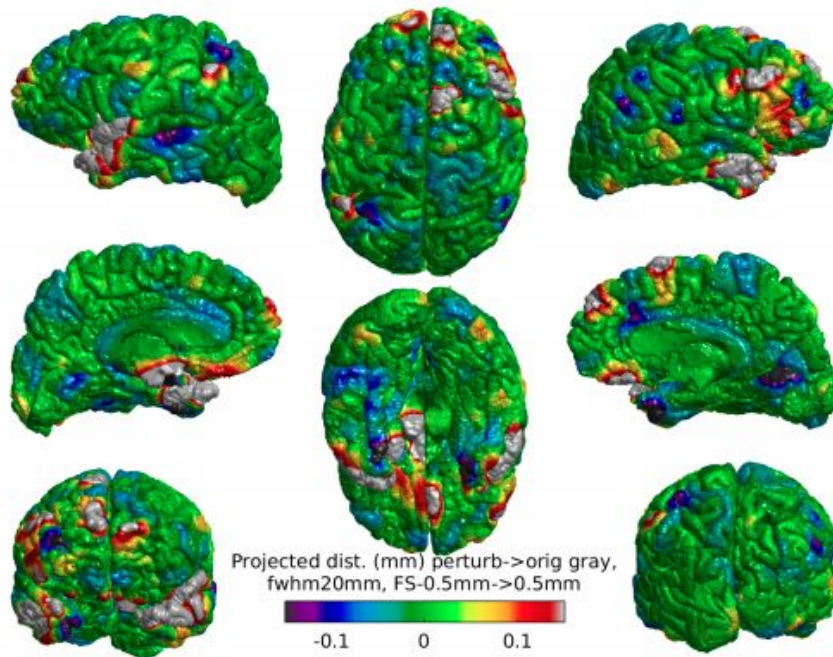
(Scaria, 2017)

(Khundrakpam et al., 2017)  
(Lewis, 2017)

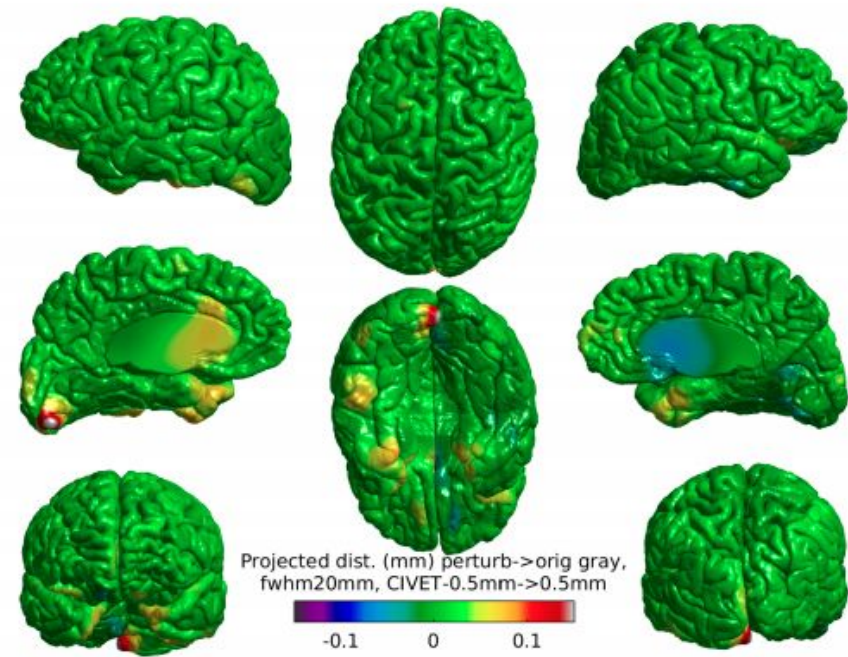


# Contributing Factor: Instability

1-voxel noise injections at 1% intensity



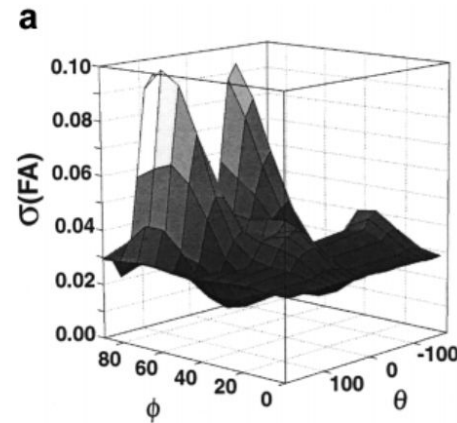
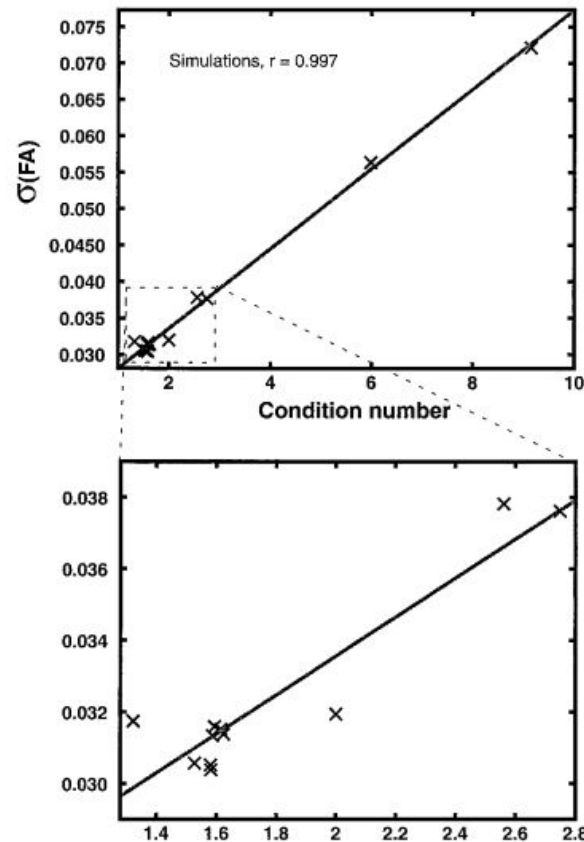
Freesurfer



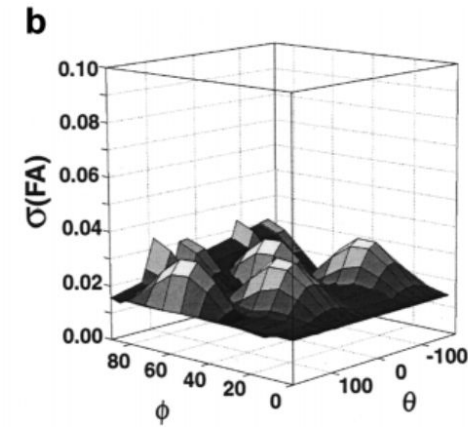
CIVET

(Lewis et al., 2017)  
(Lewis, 2017)

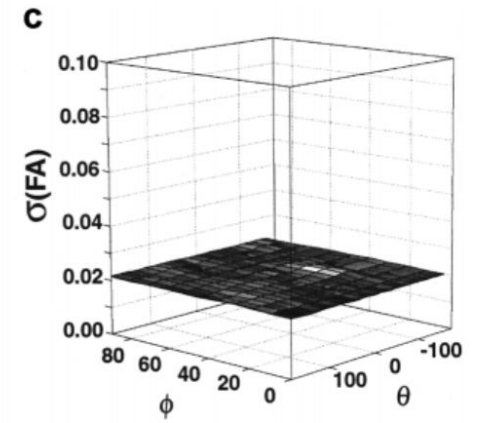
# Instabilities can\* be Anticipated



Scheme 1, Tetrahedral



Scheme 10, Jones (N=6)



Scheme 8, Jones (N=30)

\*sometimes

(Skare et al., 2000)

What about the instabilities  
that we see *in practice*?





# Such as addition in Python...

```
In [1]: def count_to_one(N):  
        step = 1.0 / N  
        return step, step*N, sum([step for _ in range(int(N))])
```

- The function above should return the number 1: we're adding N steps of size 1/N together

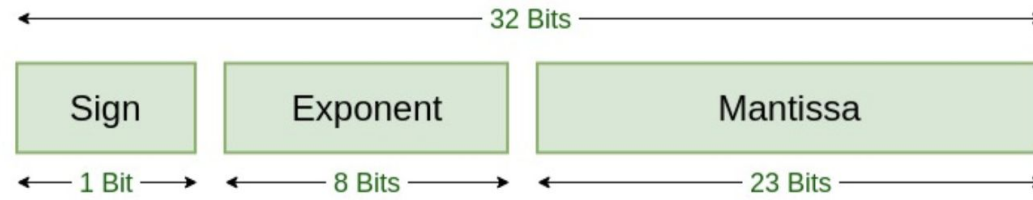
```
In [2]: N = 5  
        print(count_to_one(N))  
  
(0.2, 1.0, 1.0)
```

- great, it works, now let's try a slightly bigger number

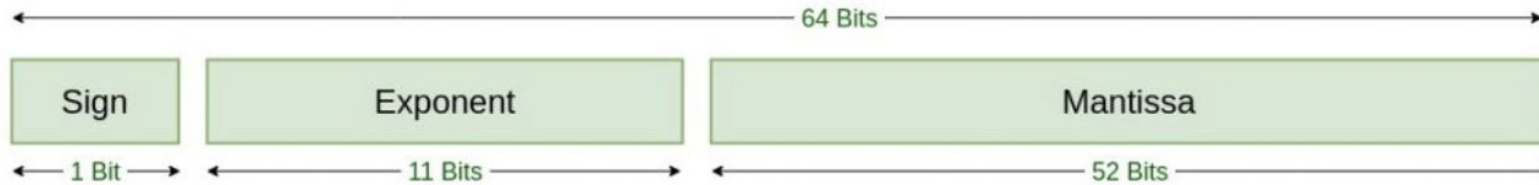
```
In [3]: N = 1e3  
        print(count_to_one(N))  
  
(0.001, 1.0, 1.000000000000000007)
```



# Floating Point Data are Finite



Single Precision  
IEEE 754 Floating-Point Standard



Double Precision  
IEEE 754 Floating-Point Standard

<https://www.geeksforgeeks.org/ieee-standard-754-floating-point-numbers/>

# Floating Point Arithmetic is Inexact

E.g. addition is non associative

$$\begin{aligned}(2000. \oplus -1998.) \oplus 1.333 &= 2.000 \oplus 1.333 = 3.333 \\ 2000. \oplus (-1998. \oplus 1.333) &= 2000. \oplus -1997. = 3.000\end{aligned}$$

(inspired by Parker et al., 1997)

# Monte Carlo Arithmetic (MCA)

Inexact quantities become random variables

$$\tilde{x} = \text{inexact}(x, s, \xi) = x + 2^{e-s} \xi \quad \text{where } e \text{ is the order of magnitude of } x$$

$$t\_digit\_precision(x) = \begin{cases} x & \text{if } x \text{ can be expressed exactly with } t \text{ digits} \\ \text{inexact}(x, t, \xi) & \text{otherwise.} \end{cases}$$

(Parker et al., 1997)

# Setup

Compile C/C++/Fortran lib with Verificarlo

## Instrumentation (MCA)

1. if floating point operation:
2.   {float, double} -> {double, quad}
3.   (PB) simulate unrounding
4.   perform operation
5.   (RR) simulate rounding
6.   {double, quad} -> {float, double}
7. endif



Verificarlo v0.2.3

build passing DOI 10.5281/zenodo.3370928

A tool for automatic Montecarlo Arithmetic analysis.



# We Instrumented...

- Python
- Cython libs
- Numpy
- BLAS
- Lapack

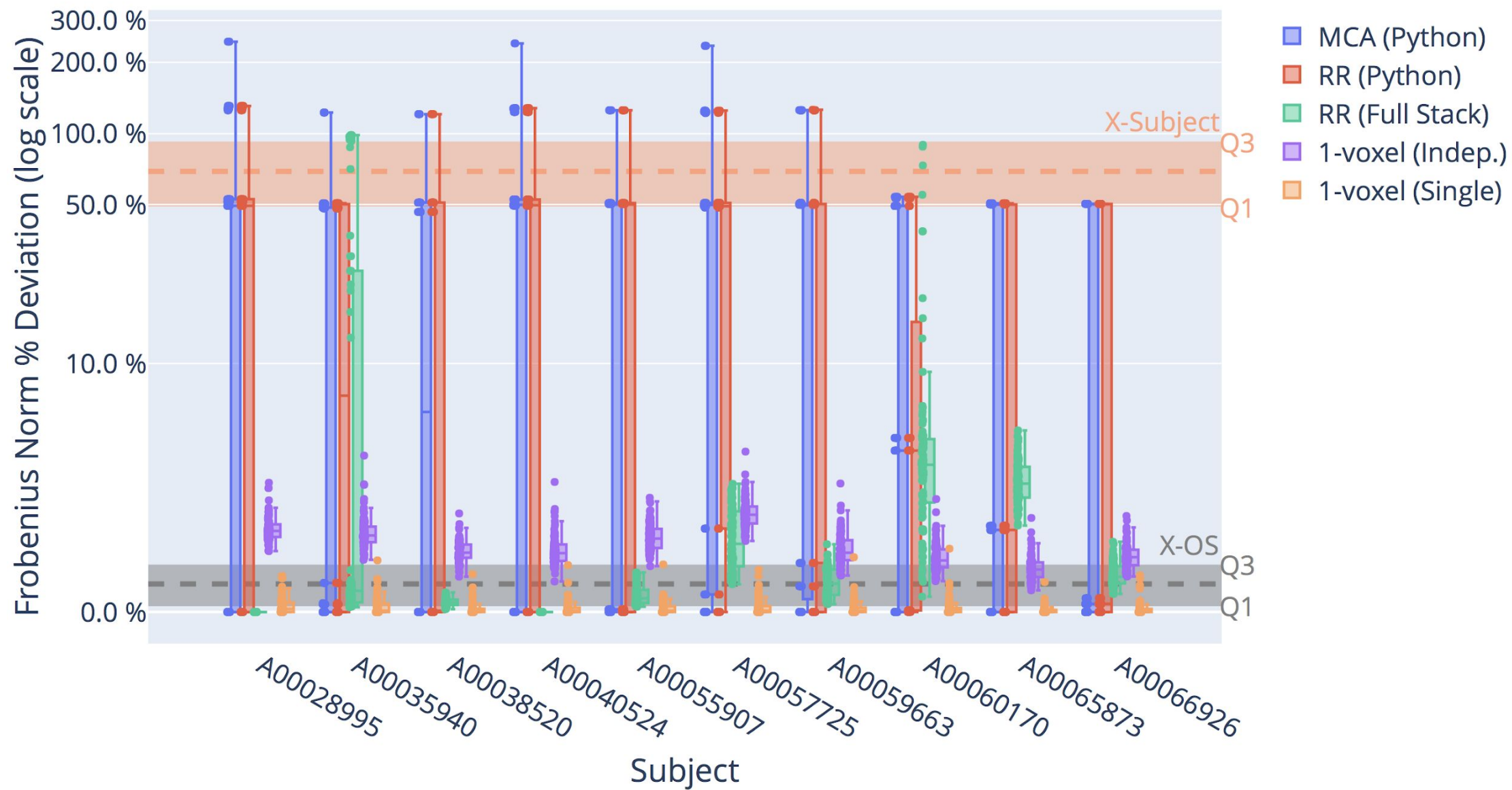
<https://hub.docker.com/gkiar/fuzzy/>

# We ran a Pipeline with...

- MCA (Python/Cython only)
- RR (Python/Cython only)
- RR (Full Stack)
- 1-Voxel noise (per 3D volume)
- 1-Voxel noise (per 4D volume)

... 100 x each, for 10 subjects

## Differences in Perturbed Structural Connectomes



Takeaway #1:

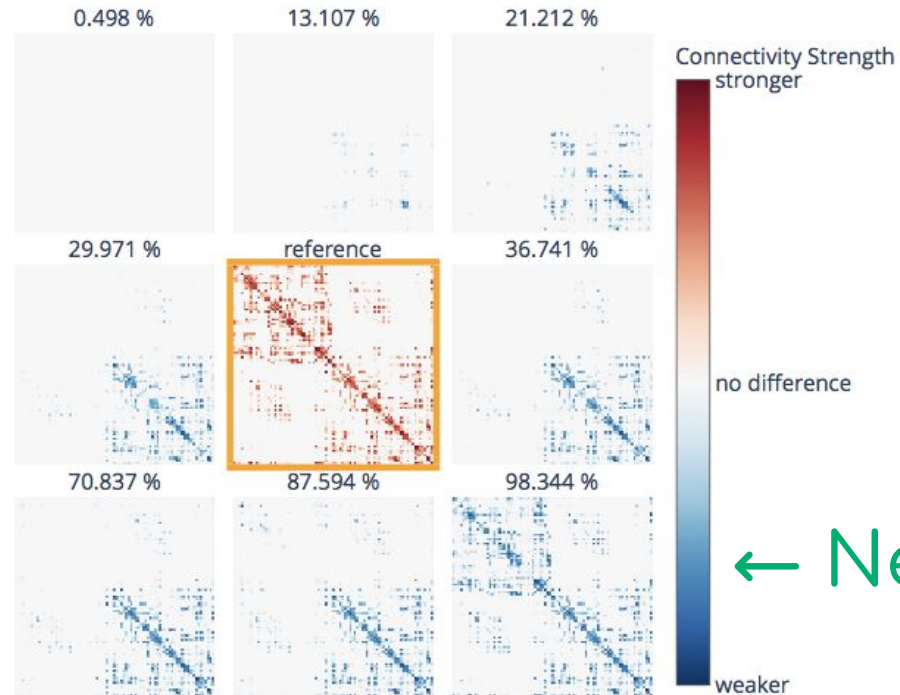
A deterministic pipeline shows instabilities comparable to individual-level variation



# What do these changes look like?

No deviation →

Error-Induced Deviations from Reference Connectome



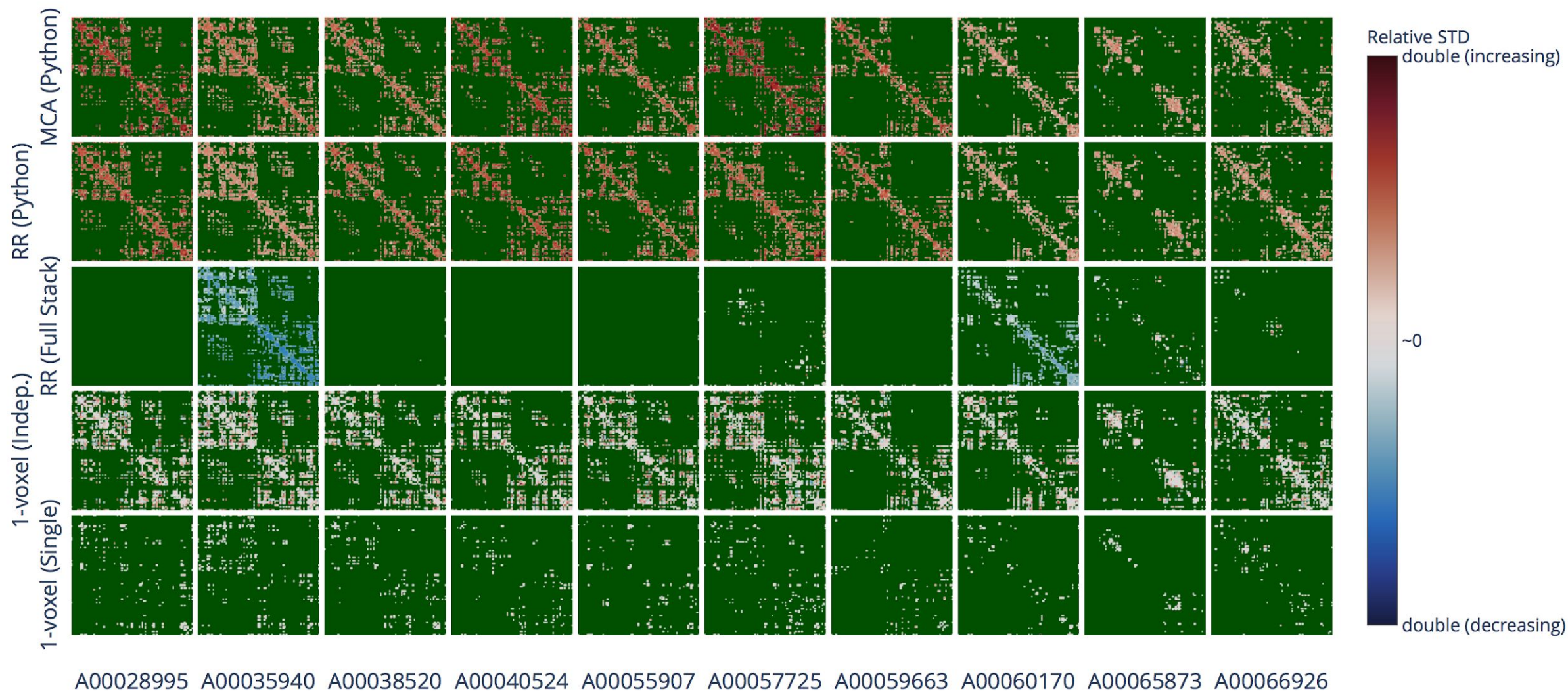
Takeaway #2:

The perturbation-induced instabilities are,  
at least in some cases, structured





## Structural Differences Across Perturbation Modes and Subjects



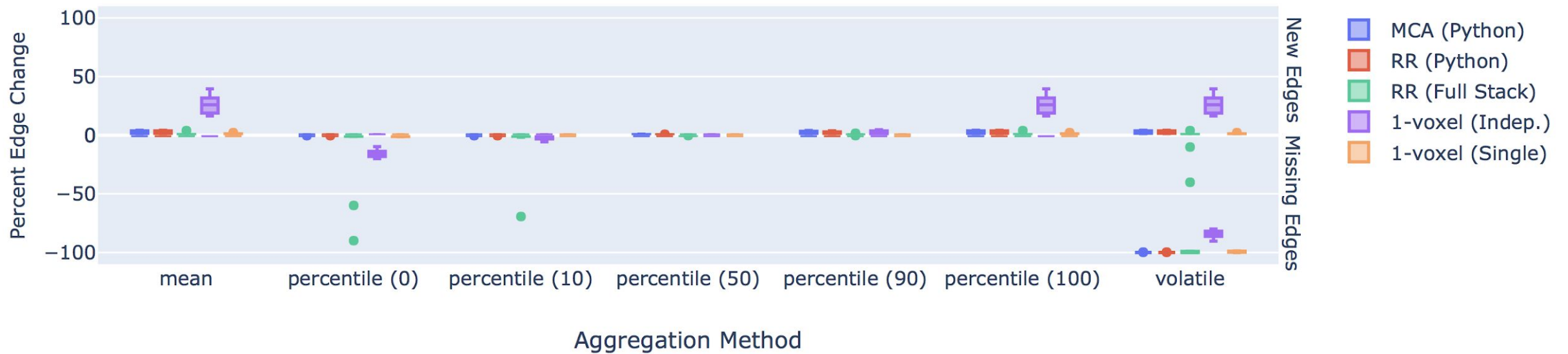
Takeaway #3:

Instabilities have a bi-directional effect  
and are highly data-dependent



# Can aggregation help?

Deviations in Aggregated Edge Count from Reference



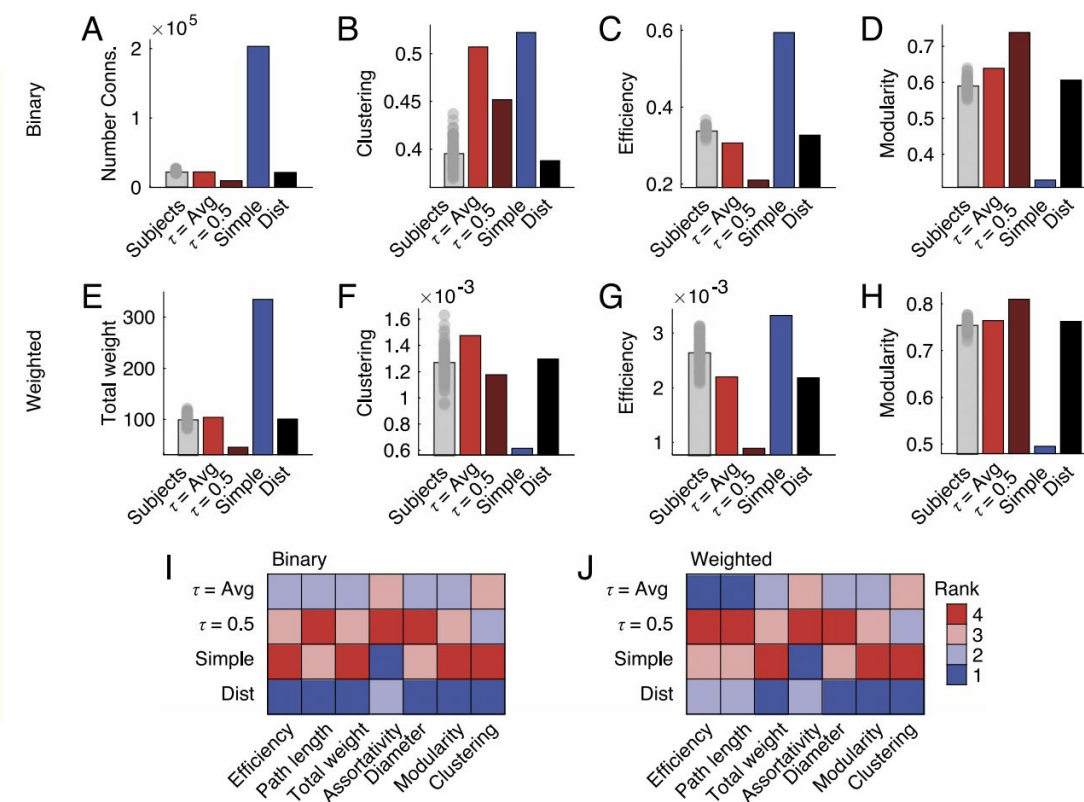
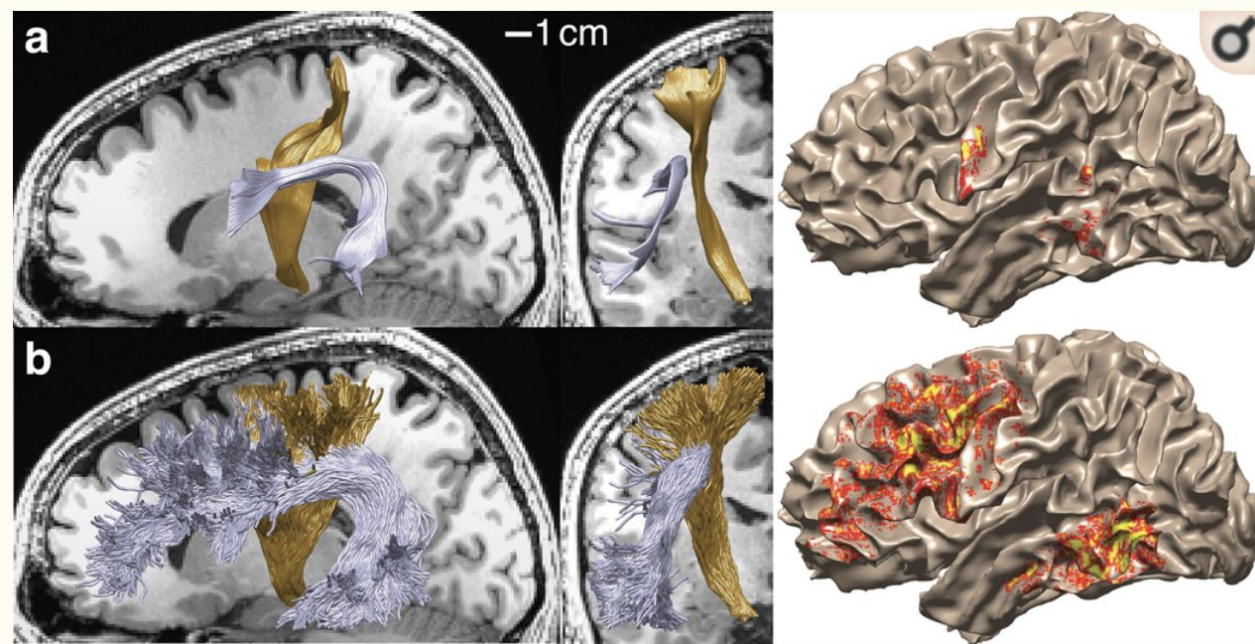
Takeaway #4:

Aggregation may create stable derivatives  
alongside estimates of their variance





# Evaluation of "Validity" Still Needed



(Pestilli, 2015)

(Betzel, 2019)



# Summary

- Minor perturbations can induce severe instabilities in neuroimaging pipelines
- Stability analyses inseparably evaluate tool-dataset pairs, rather than either in isolation
- Analytical impact of instabilities is in progress



All code mentioned in this presentation is publicly available on GitHub.

Thanks!

Find me @



gkiar



g\_kiar



greg.kiar@mail.mcgill.ca

# Acknowledgements



...



Fondation  
Brain Canada  
Foundation



HEALTHY BRAINS  
FOR **HEALTHY LIVES**



**NSERC**  
**CRSNG**



**CANADA**  
**FIRST**  
RESEARCH  
EXCELLENCE  
FUND

**APOGÉE**  
**CANADA**  
FONDS  
D'EXCELLENCE  
EN RECHERCHE

**Mitacs**  
Globalink

**Inria**  
inventeurs du monde numérique



# Questions?

---

