



Slides: <https://doi.org/10.6084/m9.figshare.10299212.v1>



# Evaluating the Stability of Neuroimaging Pipelines

Gregory Kiar, with work contributed by

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

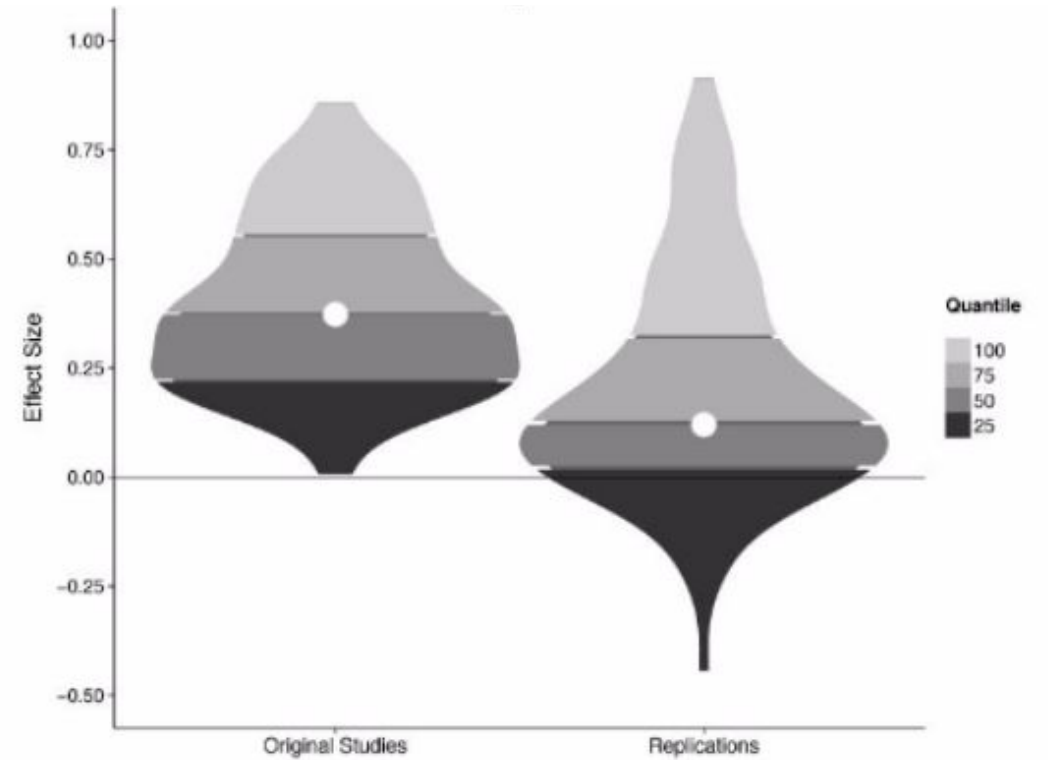
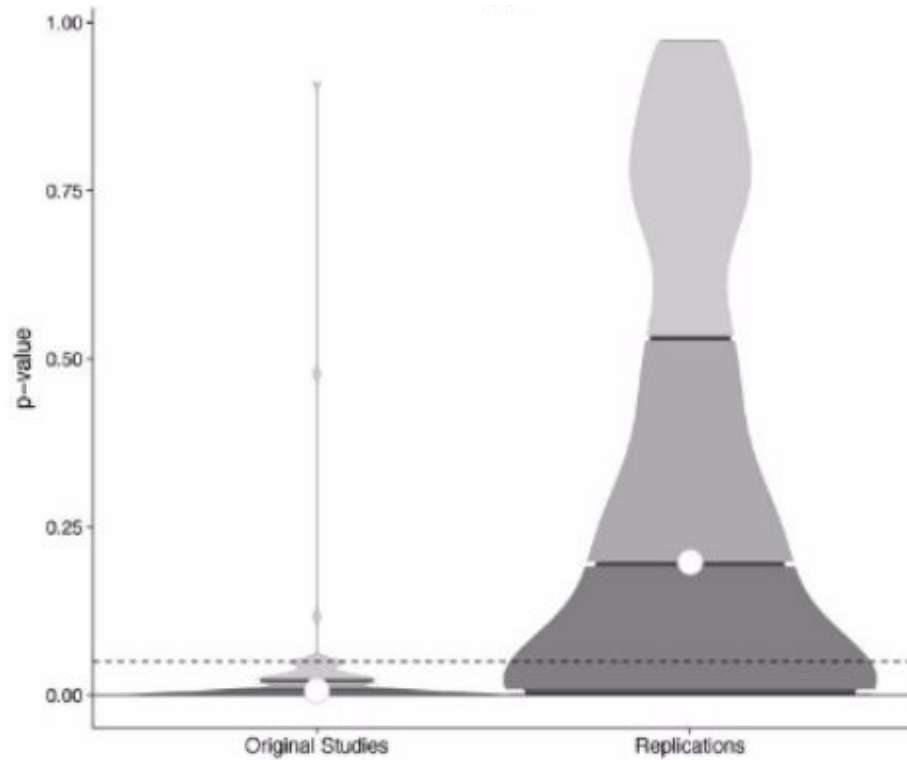


# Overview

- Topics in reproducibility
- Operationalizing Stability
- Evaluation of a Neuroimaging Pipeline



# Reproducibility is Measurable



(Open Science Collaboration, 2015)

# Replication can be difficult...



# But this talk is NOT about replicability

Let's say we go so far as to get that right, then...



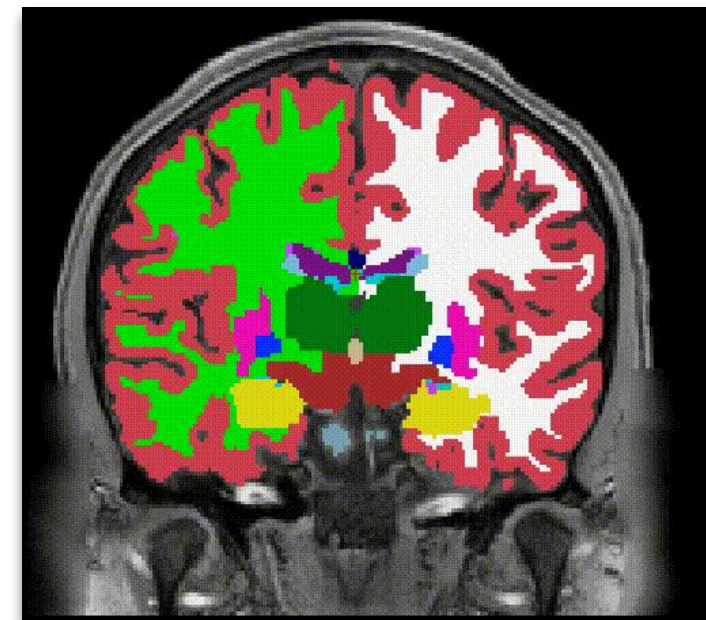
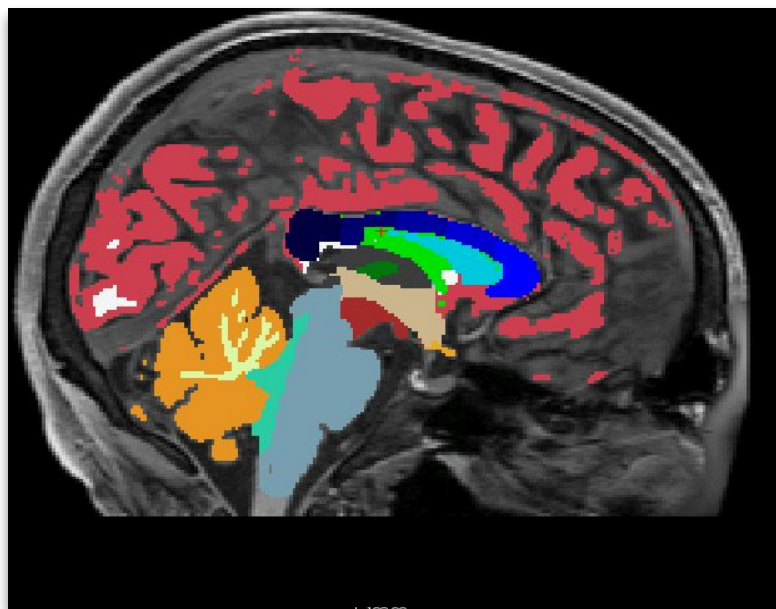
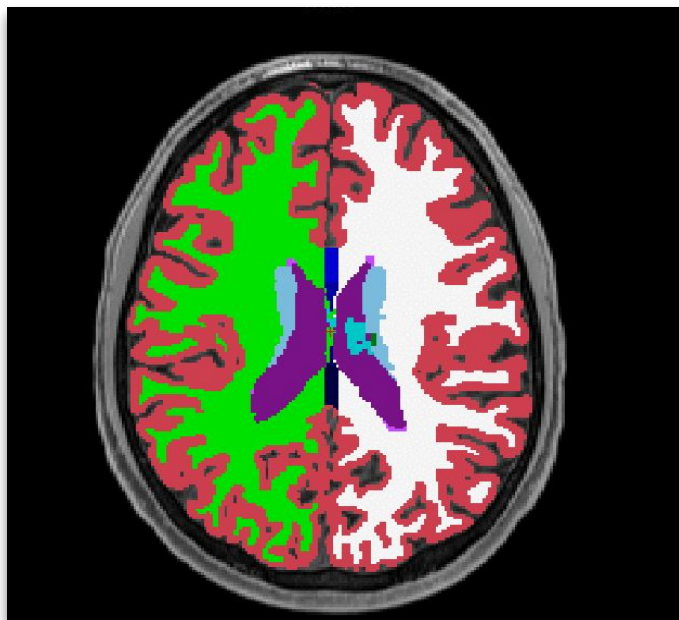
Stability is playing spot the difference





# Difference Spotted! OS

CentOS 6 vs CentOS 7



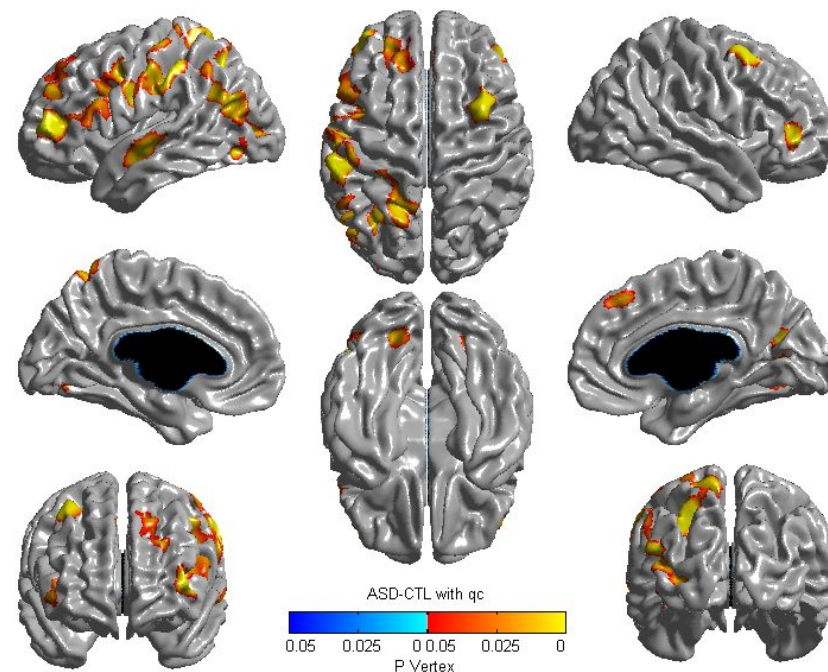
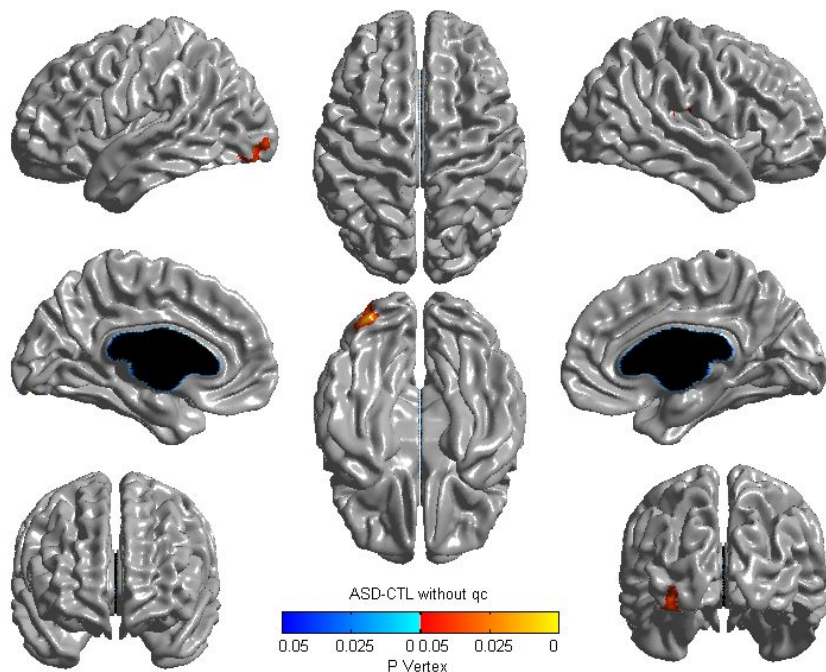
(Scaria, 2017)

(Scaria et al., 2017)

# Difference Spotted! Data Quality

Full Dataset (N ~ 1100)

vs QC'd Dataset (N ~ 400)



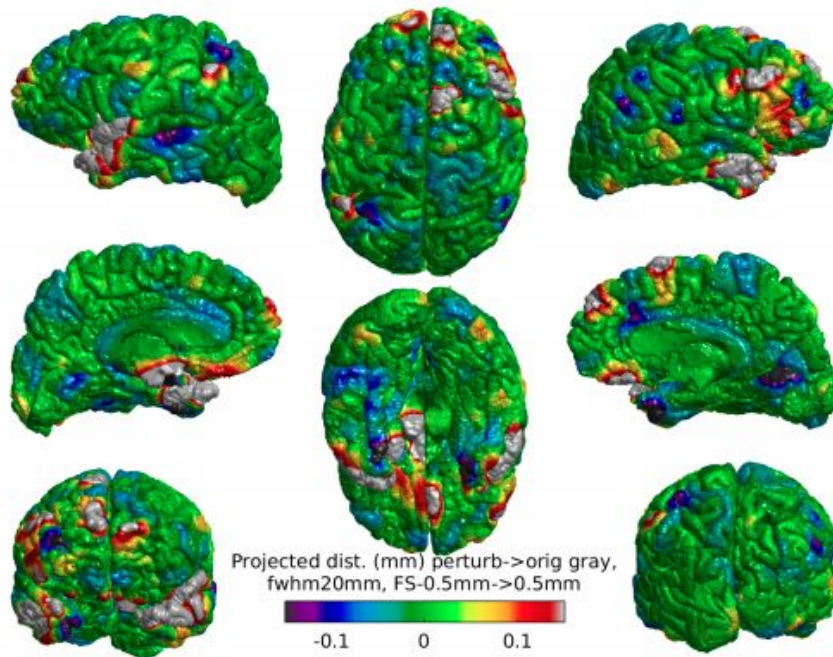
(Scaria, 2017)

(Khundrakpam et al., 2017)

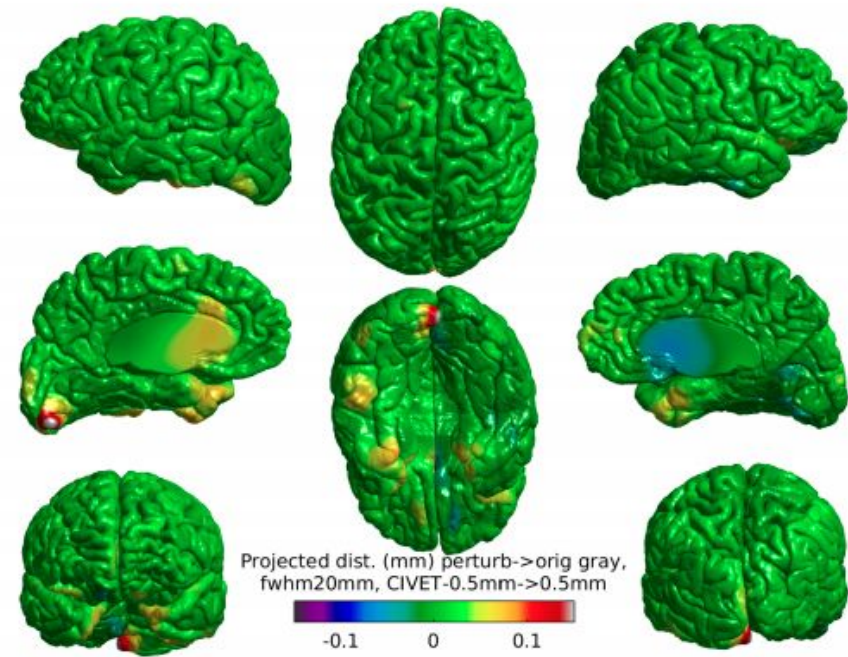


# Difference Spotted! Instability

1-voxel noise injections at 1% intensity



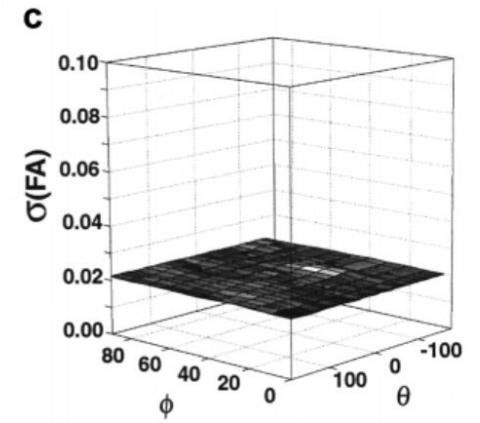
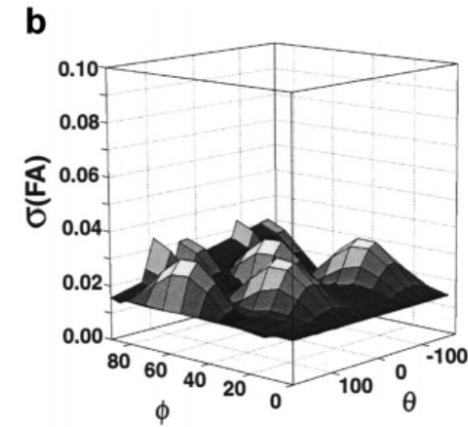
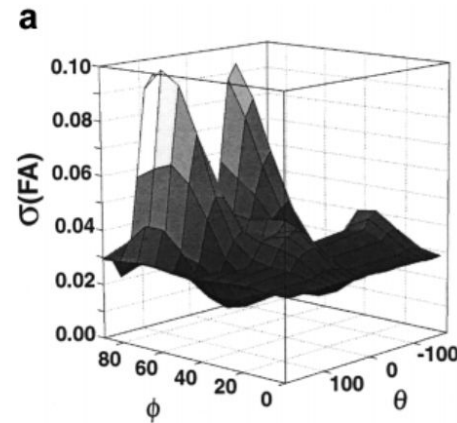
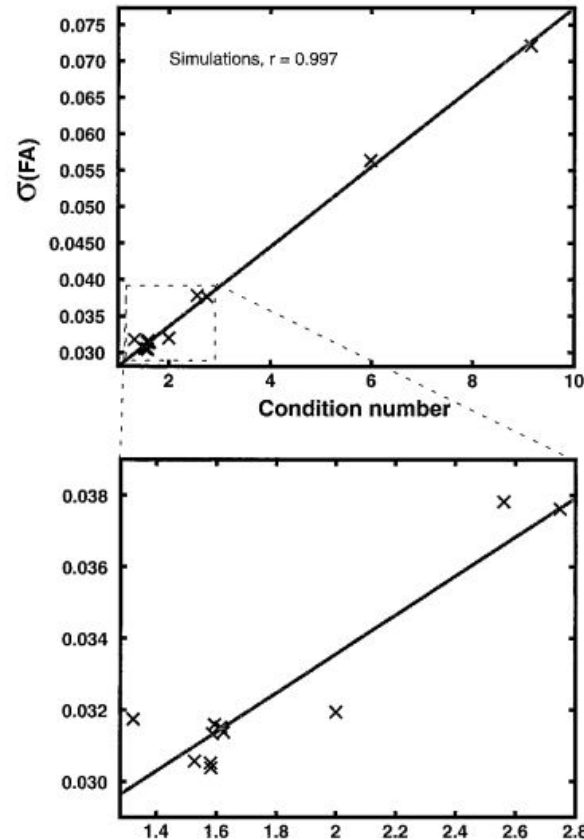
Freesurfer



CIVET

(Lewis et al., 2017)

# Instabilities can\* be Anticipated



\*sometimes

(Skare et al., 2000)

What about the even simpler errors 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

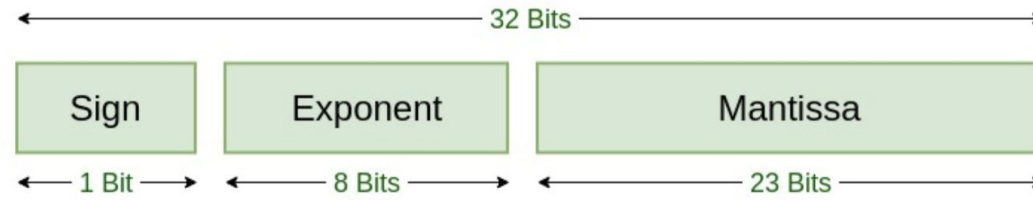
```
In [2]: N = 5  
        print(count_to_one(N))  
  
(0.2, 1.0, 1.0)
```

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

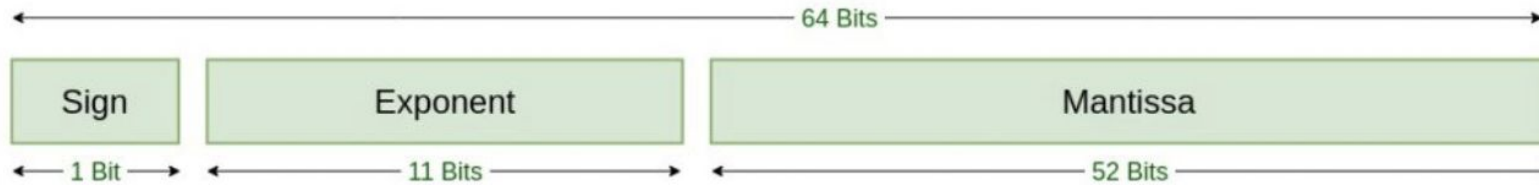
```
In [3]: N = 1e3  
        print(count_to_one(N))  
  
(0.001, 1.0, 1.00000000000000007)
```



# 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

Let's say we have 4 decimal digits of precision:

$$(2000. \oplus -1998.) \oplus 1.333 = 2.000 \oplus 1.333 = 3.333$$

$$2000. \oplus (-1998. \oplus 1.333) = 2000. \oplus -1997. = 3.000$$

What is the right answer?

(inspired by Parker et al., 1997)



# Floating Point Arithmetic is Inexact

E.g. addition is non associative

Catastrophic Cancellation

$$(2000. \oplus -1998.) \oplus 1.333 = 2.000 \oplus 1.333 = 3.333$$

$$2000. \oplus (-1998. \oplus 1.333) = 2000. \oplus -1997. = 3.000$$

(inspired by Parker et al., 1997)

# Floating Point Arithmetic is Inexact

E.g. addition is non associative

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

Round-off Error

(inspired by Parker et al., 1997)

# Monte Carlo Arithmetic (MCA)

Inexact FP 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

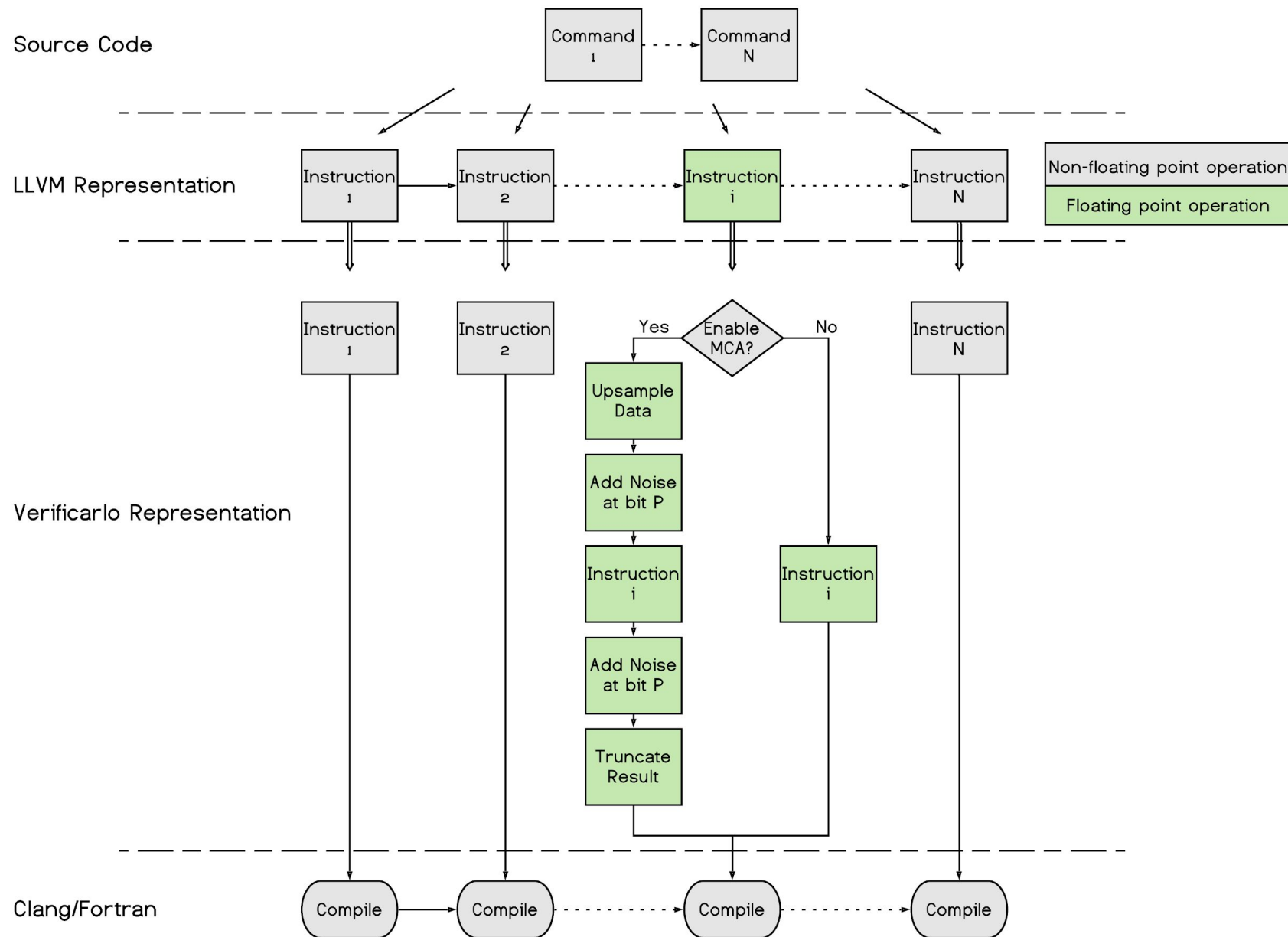
1. if floating point operation:
  2. {float, double} -> {double, quad}
  3. (PB) simulate unrounding
  4. perform operation
  5. (RR) simulate rounding
  6. {double, quad} -> {float, double}
  7. endif
- } MCA



Verificarlo v0.2.3

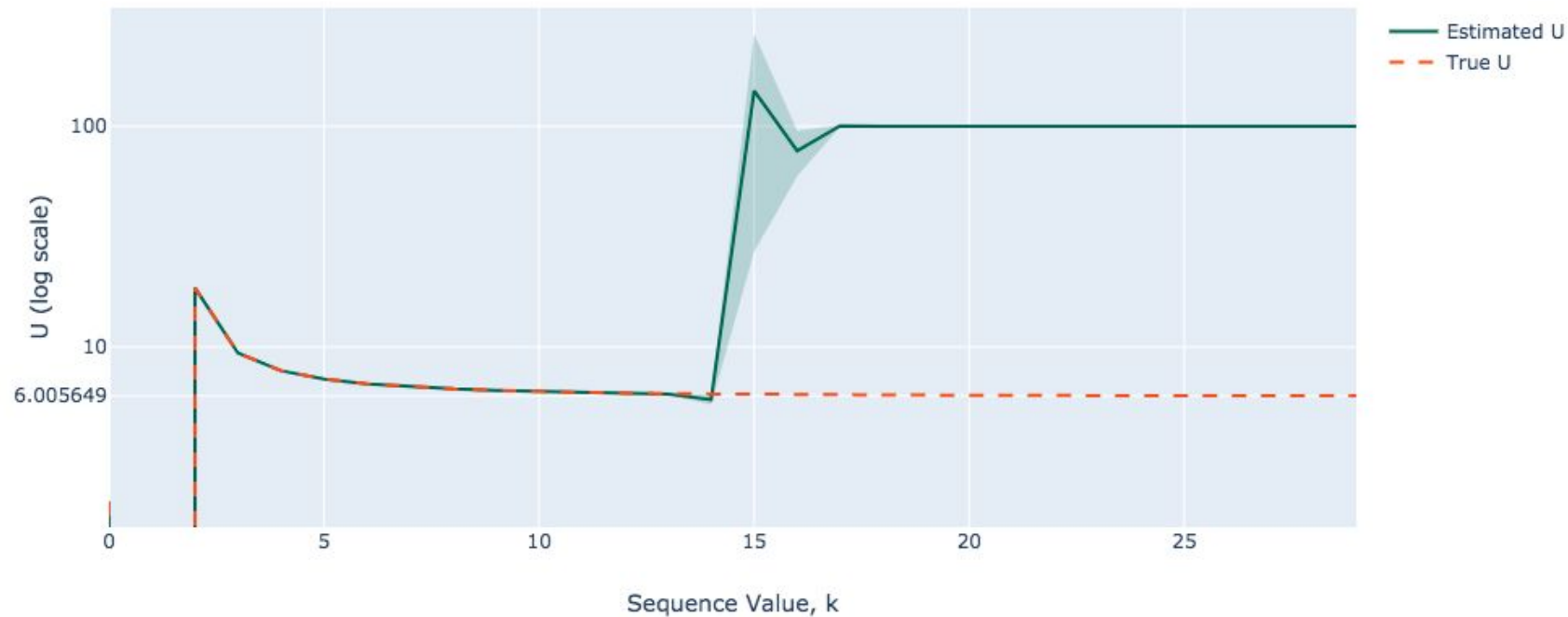
build passing DOI 10.5281/zenodo.3370928

A tool for automatic Montecarlo Arithmetic analysis.



# MCA can help identify failure points

Evaluation of the sequence:  $u(k+1) = 111 - 1130/u(k) + 3000/(u(k) u(k-1))$





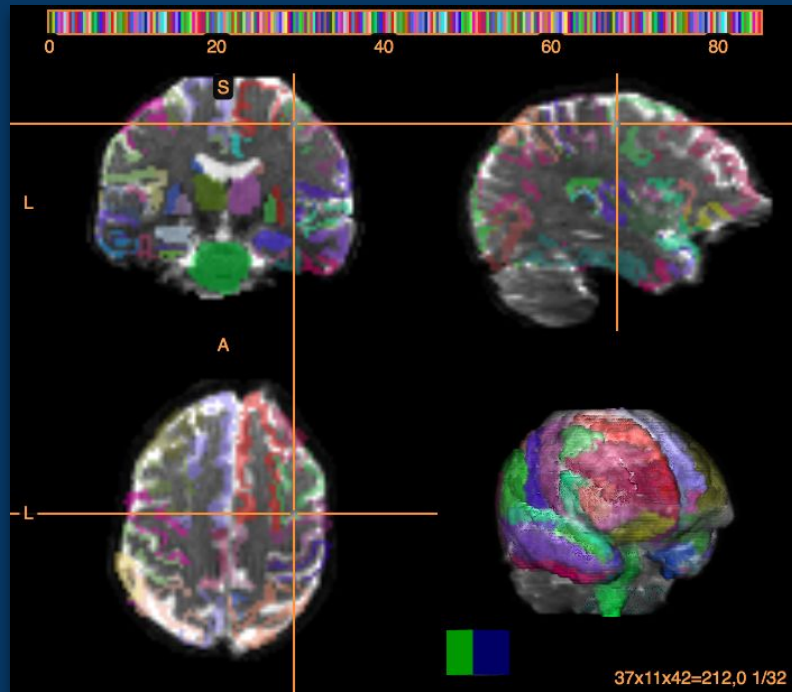
Objective:

Perturb a neuroimaging pipeline and  
observe induced instabilities



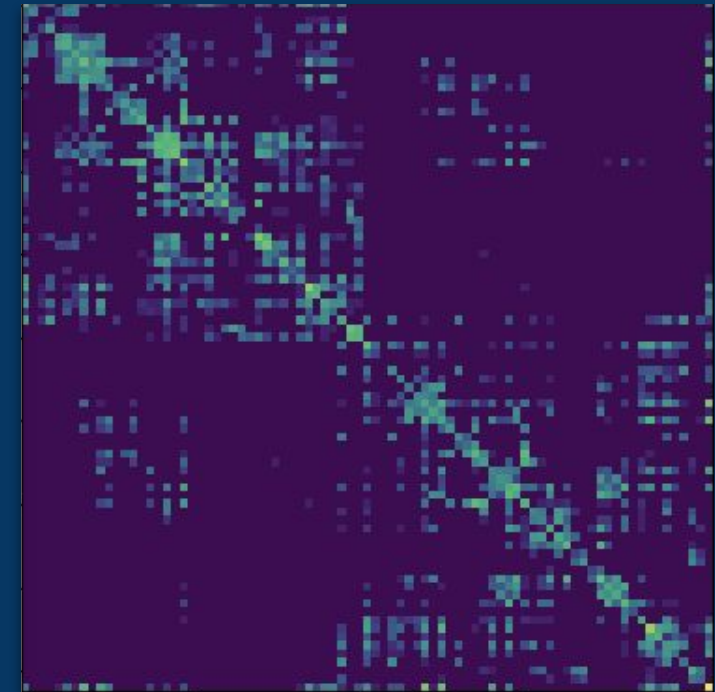
# Data in Structural Connectomics

Diffusion MRI Volumes + Labels



$(X, Y, Z, D) \sim O(10M \text{ voxels})$

Connectivity Matrix



$(83, 83) = 3403 \text{ edges}$

# Perturbation Models

## MCA

Instrument all FLOPs with...

- Full MCA
- RR only (less aggressive)

## 1-voxel noise

Double image intensity in...

- Single (1 vox / (X, Y, Z, D))
- Independent (1 vox / (X, Y, Z))



# We Instrumented...

- Python
- Cython libs
- Numpy
- BLAS
- Lapack
- Libmath

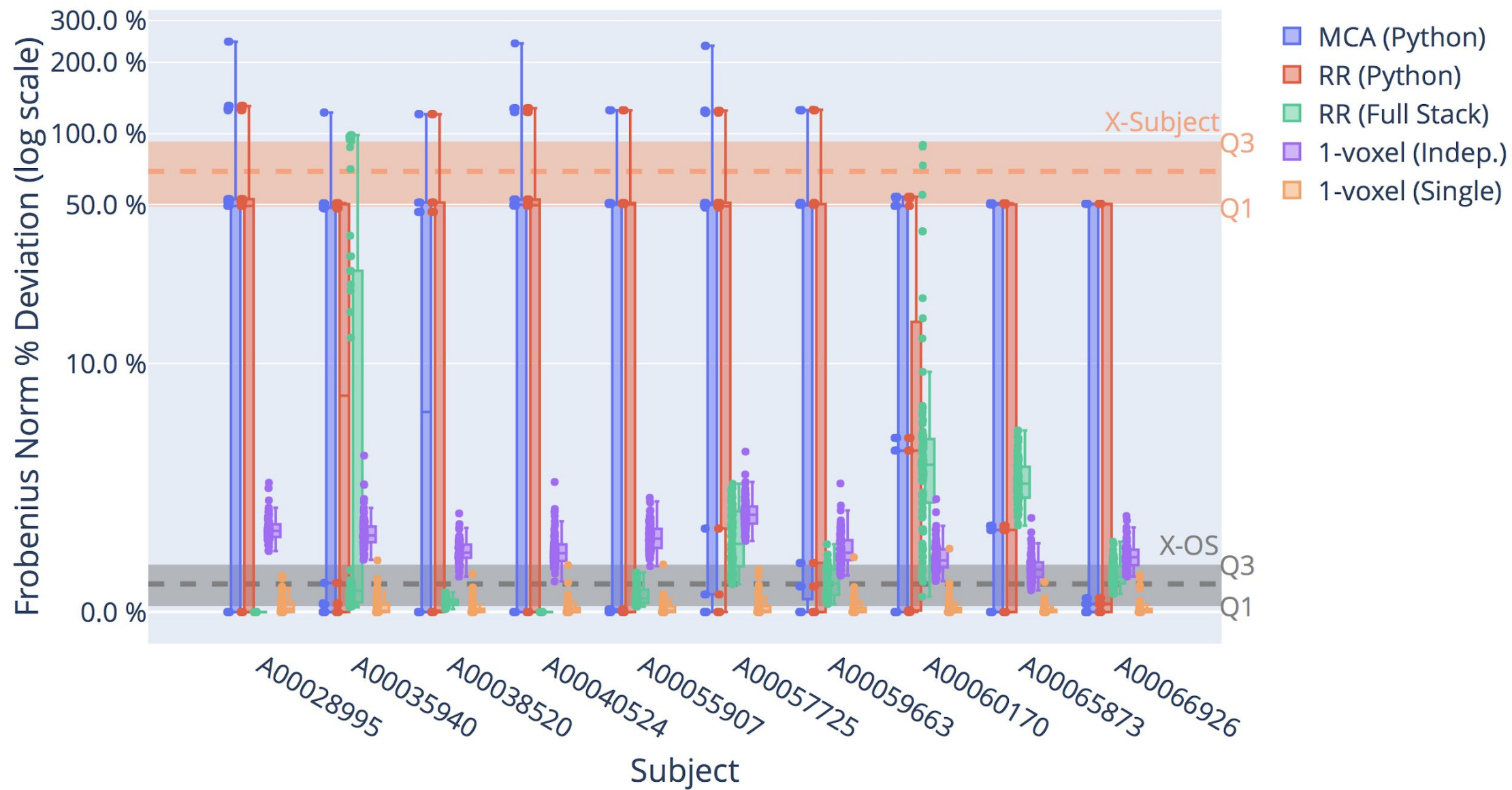
# We ran our Pipeline with...

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

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

... 100 x each, for 10 subjects

## Differences in Perturbed Structural Connectomes



Takeaway #1:

A deterministic pipeline shows instabilities comparable to individual-level variation

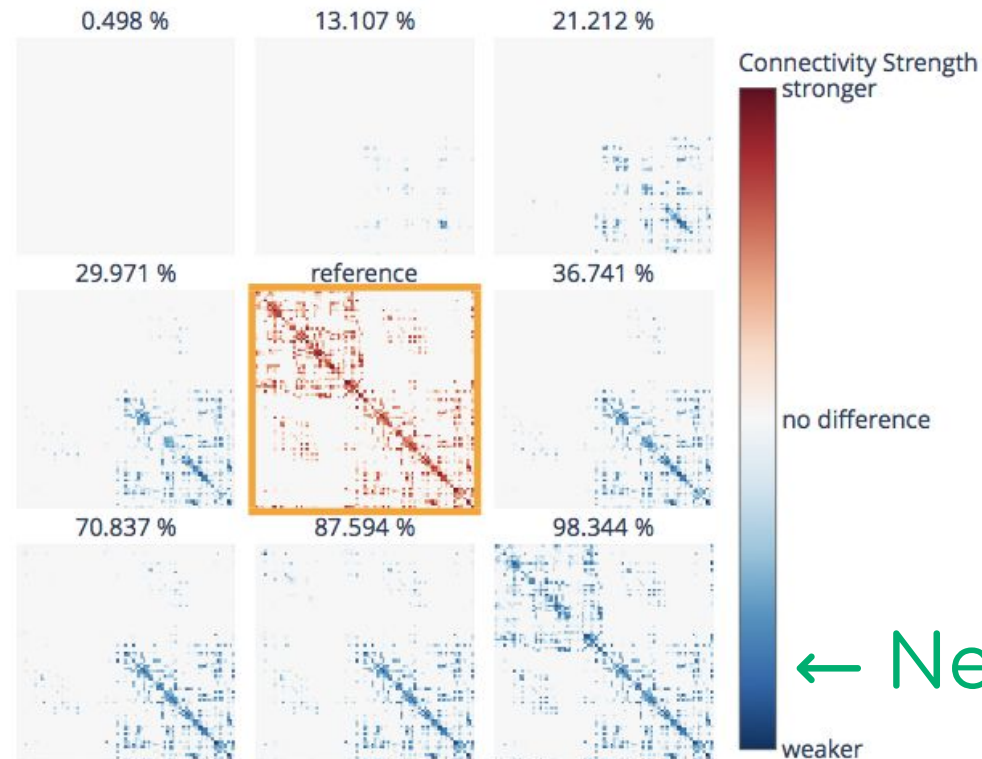




# What do these changes look like?

No deviation →

Error-Induced Deviations from Reference Connectome



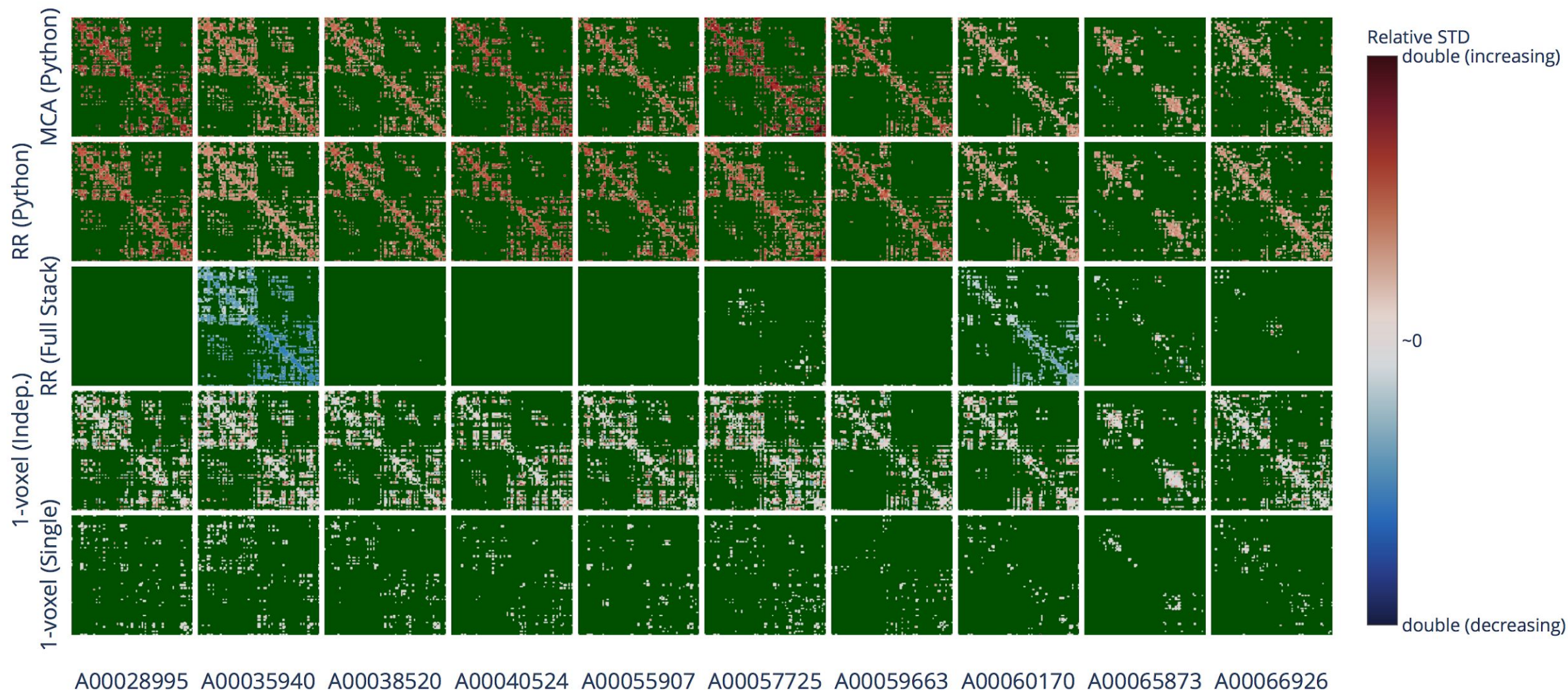
← Near full deviation

Takeaway #2:

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



## Structural Differences Across Perturbation Modes and Subjects



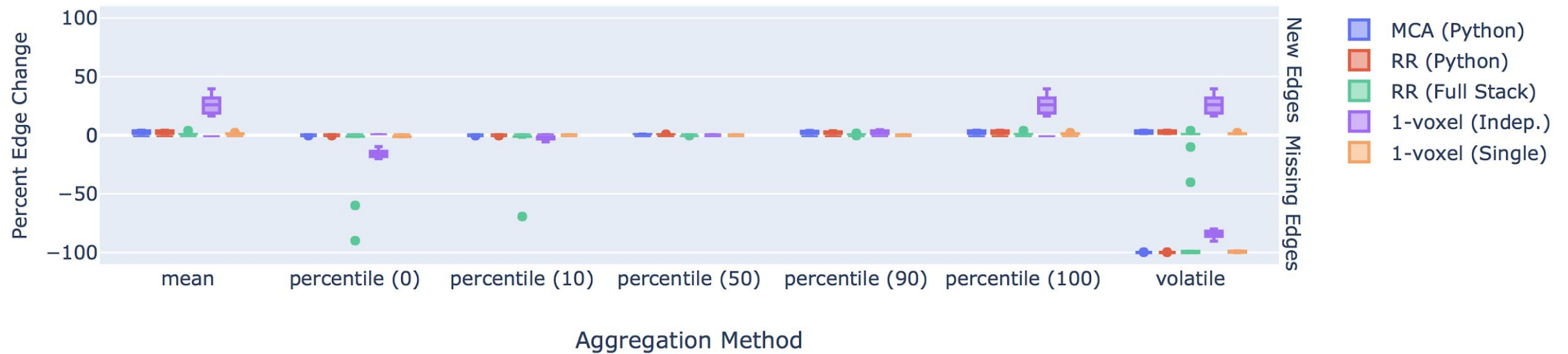
Takeaway #3:

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



# Can aggregation help?

Deviations in Aggregated Edge Count from Reference



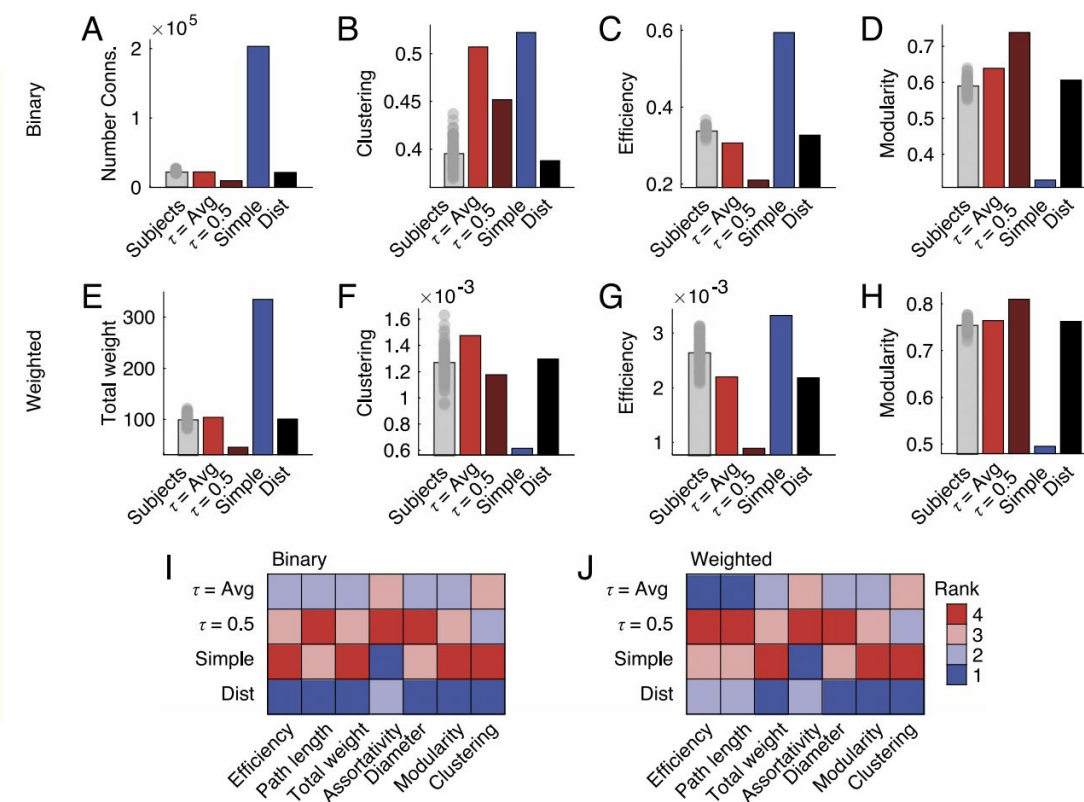
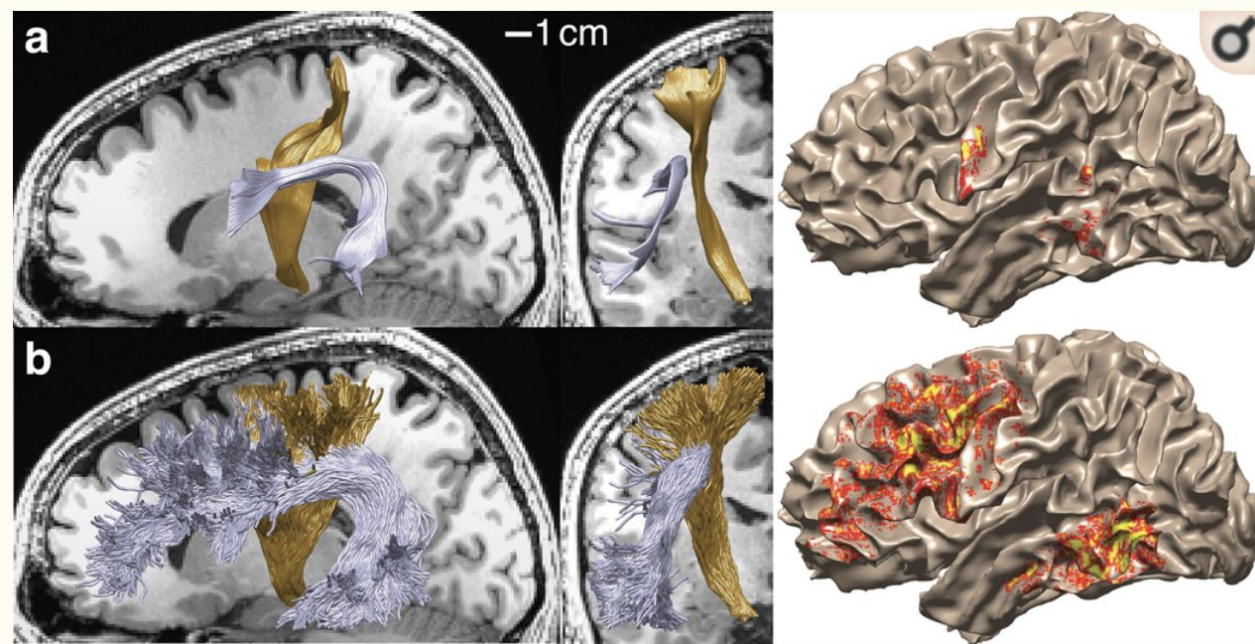
Takeaway #4:

Aggregation may create stable derivatives  
alongside estimates of their variance





# Evaluation of "Validity" Still Needed



(Pestilli, 2015)

(Betzal, 2019)

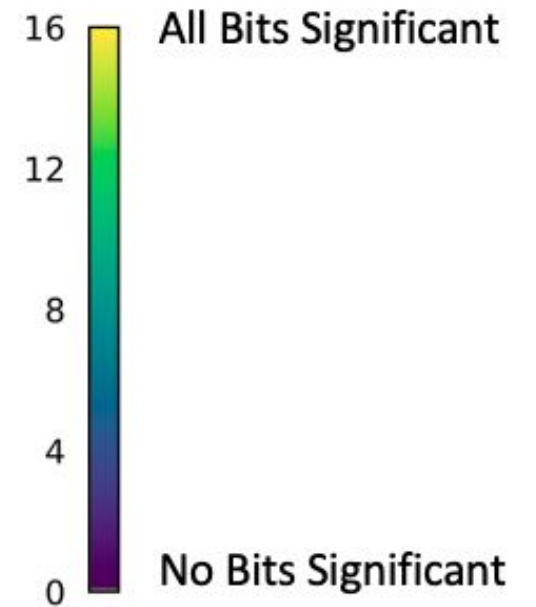
# And one more example...

(that does impact your work regardless of modality)



# Linear Registration with FLIRT

Significant Digits in Aligned Image



# Linear Registration with FLIRT

perfect world v1

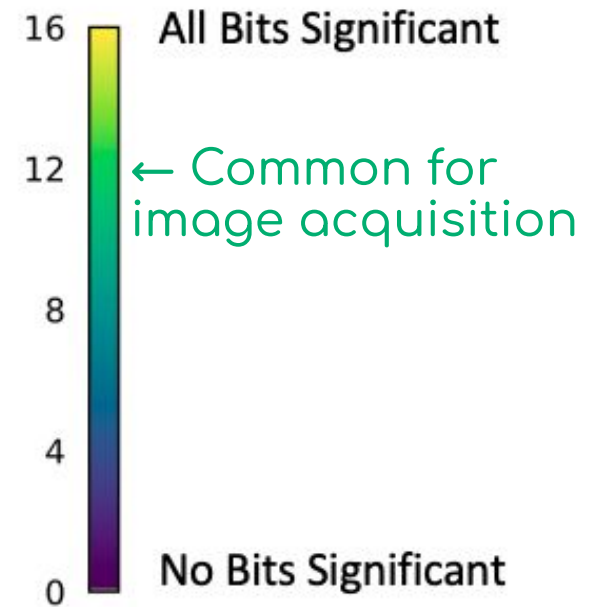
Significant Digits in Aligned Image



# Linear Registration with FLIRT

perfect world v2

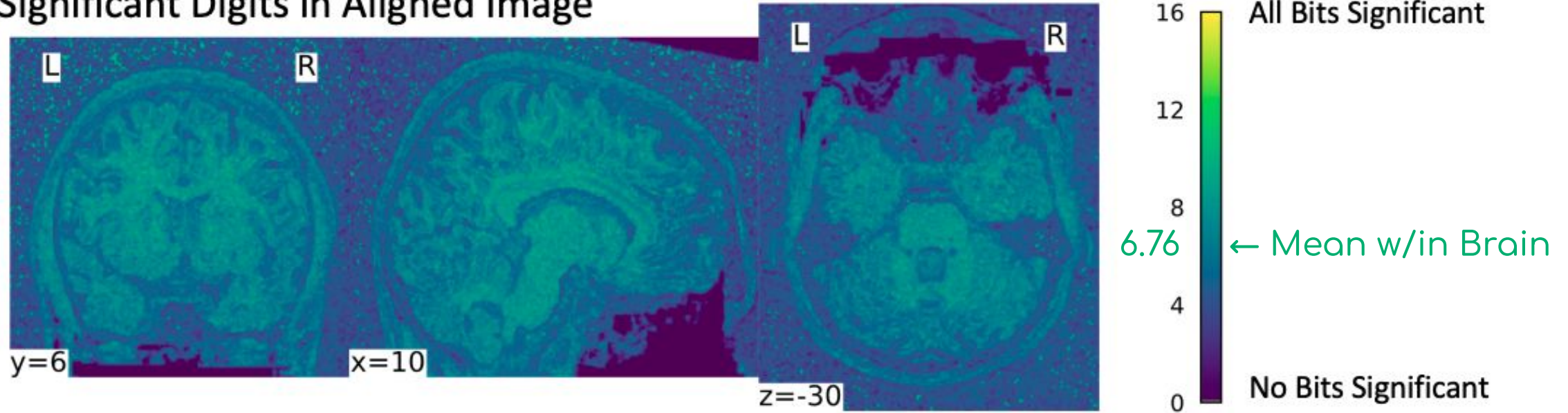
Significant Digits in Aligned Image



# Linear Registration with FLIRT

real world

Significant Digits in Aligned Image

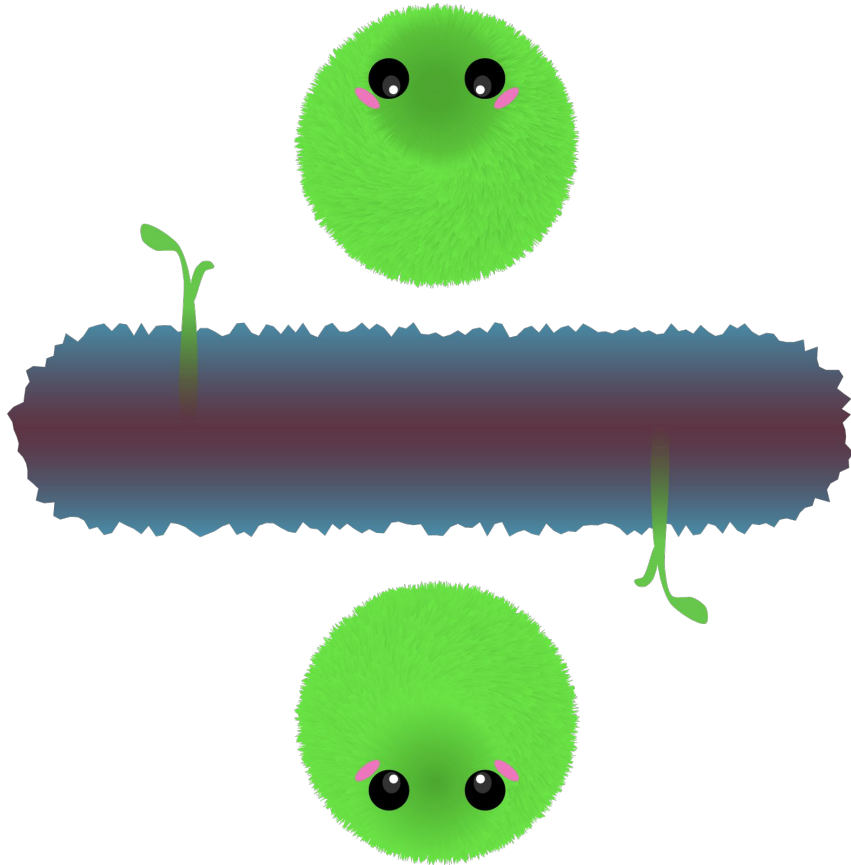


# Before you ask...

- Yes, non-linear registration is coming soon  
(it just takes 90 hrs per run with MCA...)
- Yes, we can do this for "your" tool



# <https://github.com/gkiar/fuzzy>



## ✓ Base

🕒 2 min 8 sec

✓ # 68.1

AMD64



Verificarlo

🕒 2 min 8 sec

## ✓ Environments (level 1)

🕒 21 min 42 sec

✓ # 68.2

AMD64



Blas & Lapack

🕒 7 min

✓ # 68.3

AMD64



Python

🕒 21 min 42 sec

✓ # 68.4

AMD64



Libmath

🕒 1 min 29 sec

## ✓ Environments (level 2)

🕒 21 min 51 sec

✓ # 68.5

AMD64



Python & Numpy

🕒 21 min 51 sec

## ✓ Applications

🕒 8 min 17 sec

✓ # 68.6

AMD64



FSL

🕒 5 min 20 sec

✓ # 68.7

AMD64



AFNI

🕒 8 min 17 sec

✓ # 68.8

AMD64



AFQ (Dipy)

🕒 3 min 43 sec

# Summary

- Minor perturbations can induce severe instabilities in neuroimaging pipelines
- Stability analyses inseparably evaluate tool-dataset pairs, rather than either in isolation
- Jointly exploring analytical impact 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?

---

