

## Towards reproducible neuroimaging across different analysis pipelines

Camille Maumet

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Camille Maumet. Towards reproducible neuroimaging across different analysis pipelines. CogBases 2023, Oct 2023, Paris, France. 2023. hal-04236791

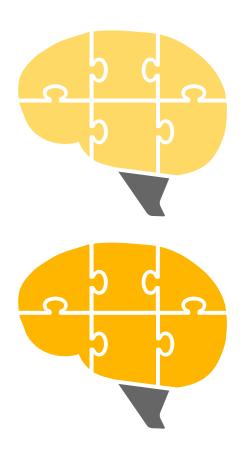
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Submitted on 11 Oct 2023

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across different analysis pipelines

CogBases workshop 2023

**Camille Maumet** 

Inria, Univ Rennes, Inserm, CNRS, IRISA



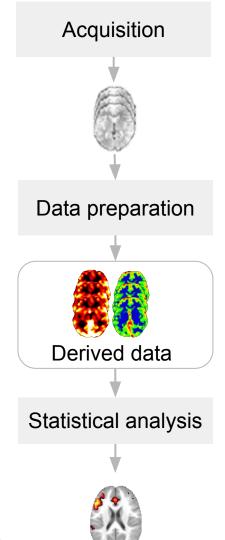




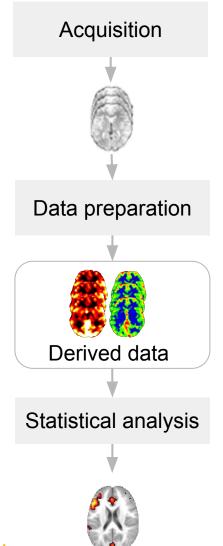




# Reproducible brain imaging



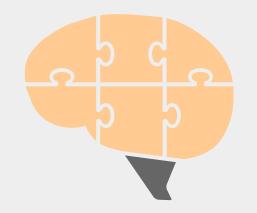
# Reproducible brain imaging



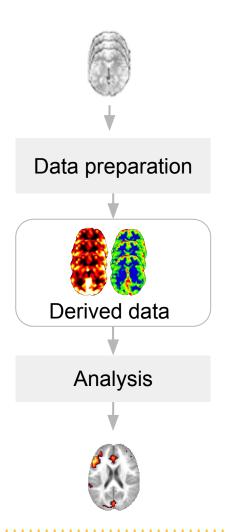
Different / Same Data

> Different / Same Method

> > Different / Same Experimenter

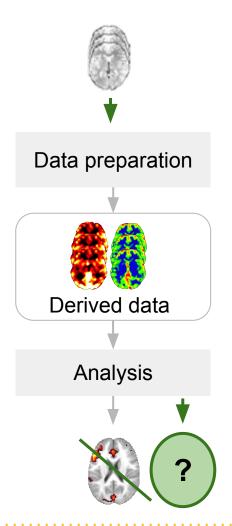


On our way to fix irreproducibility...



Irreproducible with...

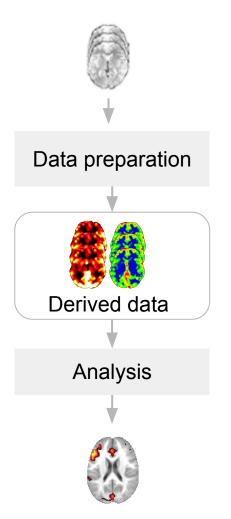
Same Data, Same Methods



- Sharing code
- Sharing environments:
   Docker, guix

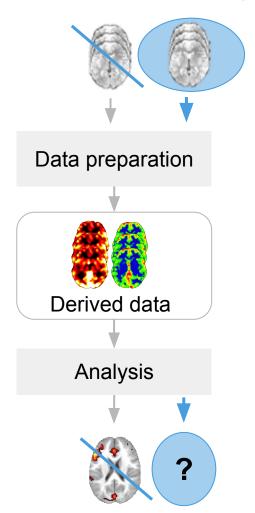
Irreproducible with...

**Different Data** 



Irreproducible with...

**Different Data** 



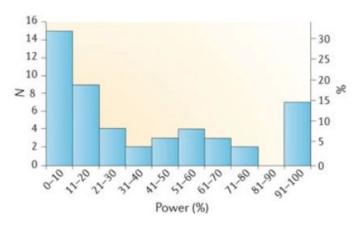
Irreproducible with...

**Different Data** 

#### Lack of power

#### Power of neuroscience studies

Power = Prob. to correctly find a significant effect when a the alternative hypothesis is true.



[Button et. al, Nat Rev Neurosci 2013]

We need bigger datasets

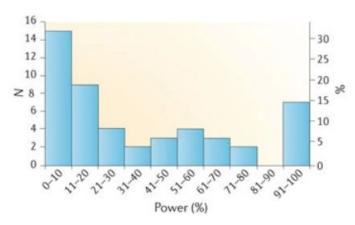
Irreproducible with...

**Different Data** 

#### Lack of power

#### Power of neuroscience studies

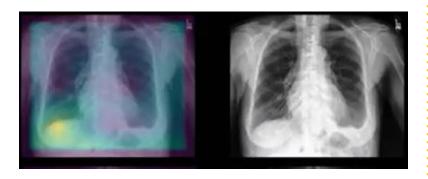
Power = Prob. to correctly find a significant effect when a the alternative hypothesis is true.



[Button et. al, Nat Rev Neurosci 2013]

#### Generalizability and fairness

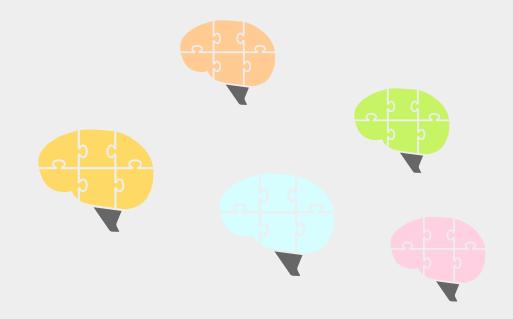
X-ray: Lung opacity detection Model trained on male images, tested on female images



[Larrazabal et. al, PNAS 2020]

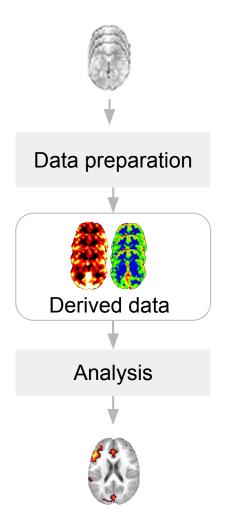
We need bigger datasets

We need representative and diverse datasets

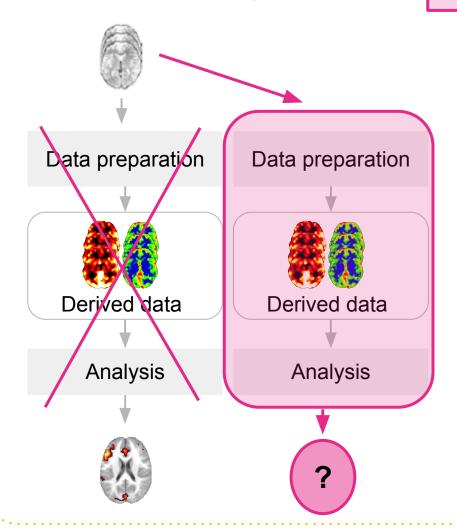


# Different pipelines... different results?

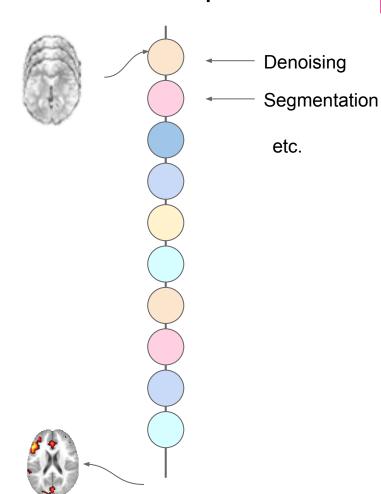
Irreproducible with...



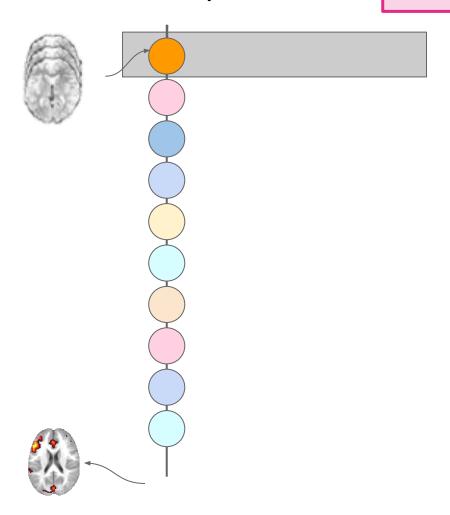
Irreproducible with...



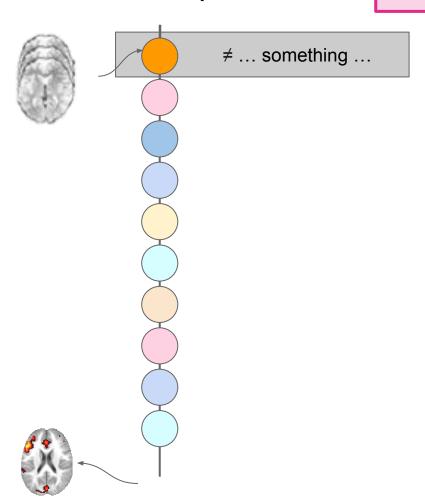
Irreproducible with...

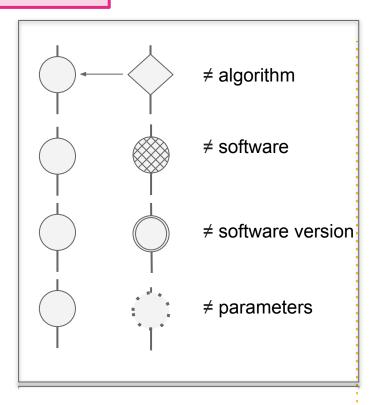


Irreproducible with...



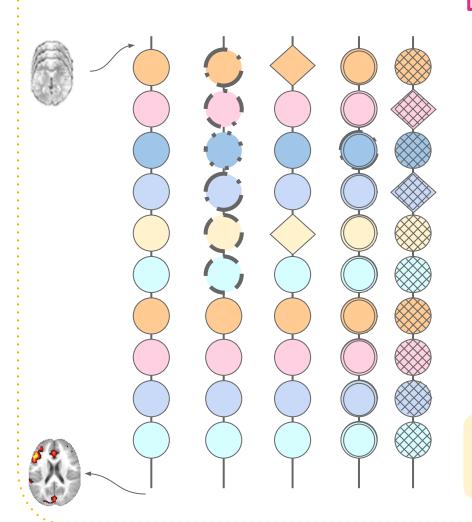
Irreproducible with...

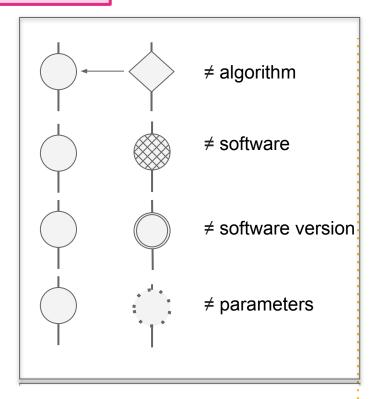




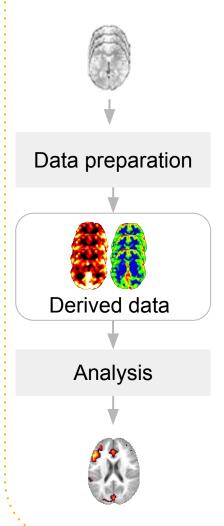
Irreproducible with...

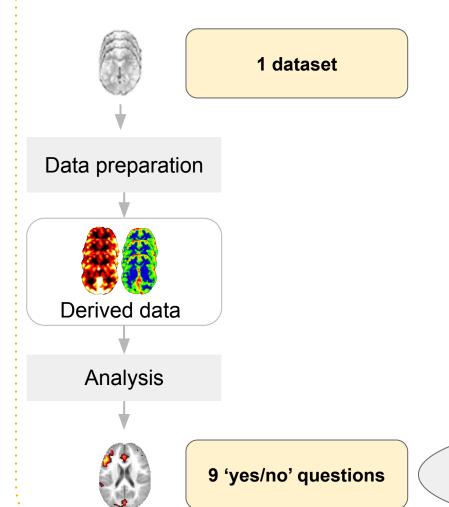
**Different Analysis** 



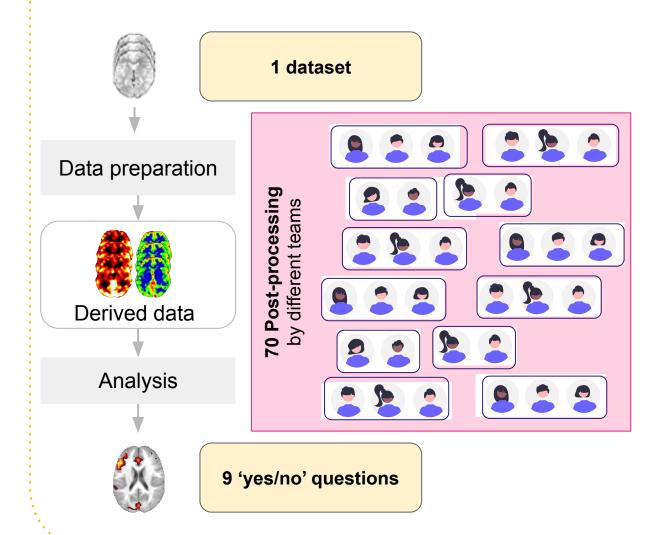


A family of acceptable pipelines over 10^30 combinations...

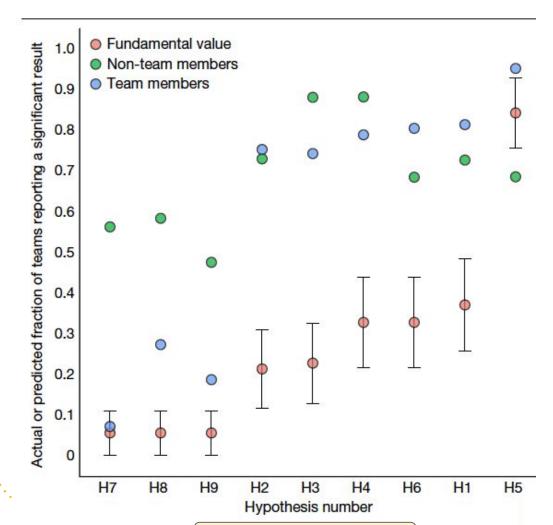




e.g. Is there a parametric effect of gain: Positive effect in ventromedial PFC for the equal indifference group?

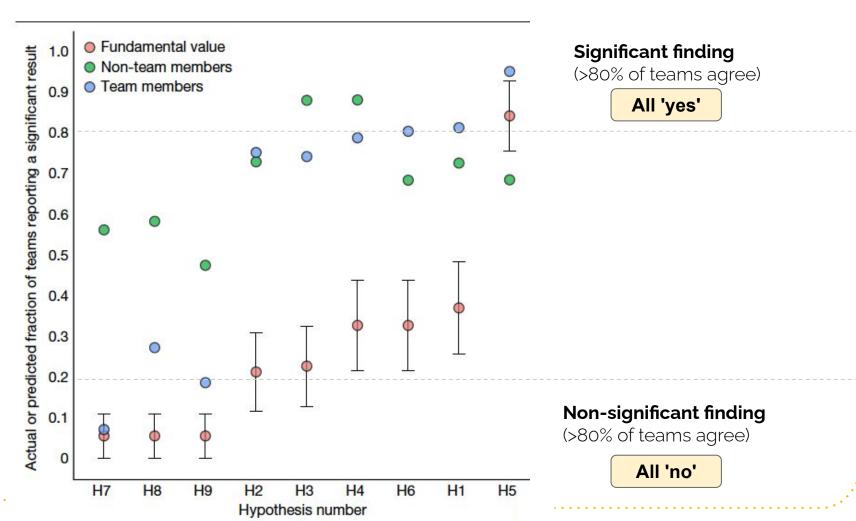


Results over the 70 post-processing pipelines:



Results over the 70 post-processing pipelines:

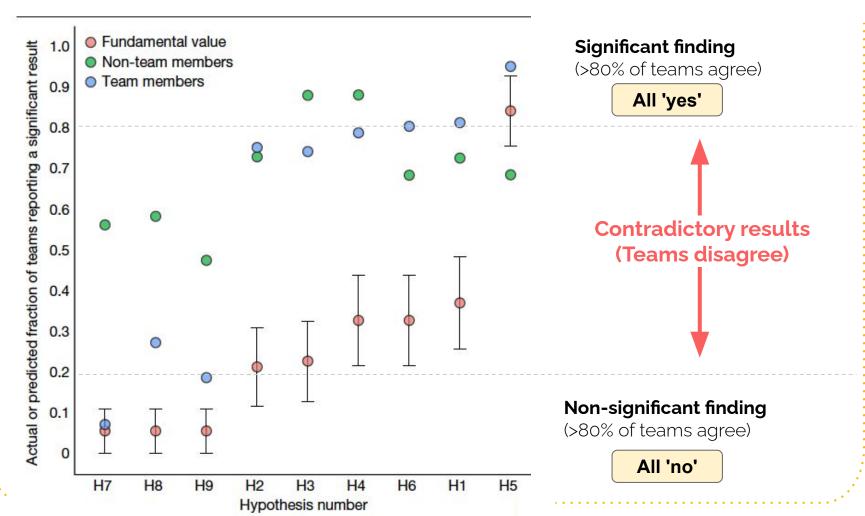
9 'yes/no' questions



[Botvinik-Nezer et. al, Nature 2020]

Results over the 70 post-processing pipelines:

9 'yes/no' questions



[Botvinik-Nezer et. al, Nature 2020]

# Different levels of variability in post-processing pipelines

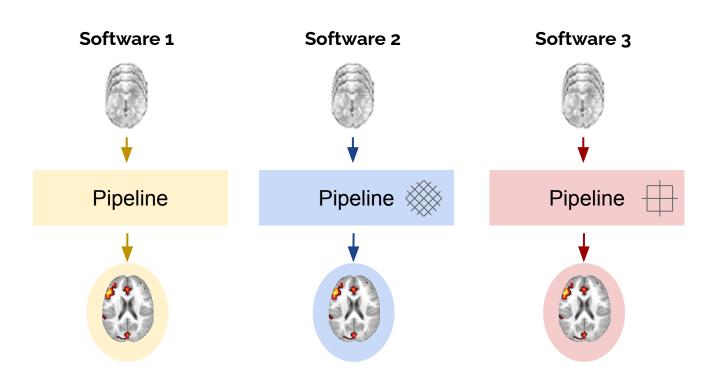
- Different end-to-end pipelines [Carp 2012]
- Different software version [Gronenschild 2012]
- Different operating system [Glatard 2015]
- Different post-processing software packages
   [Bowring 2019, 2021]

# Variability across software

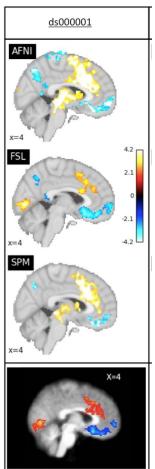
- Reproduced 3 published functional MRI studies
- Using 3 different software







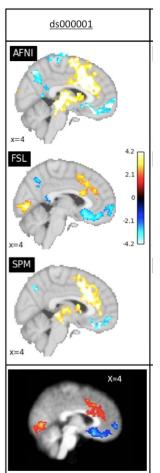
# Variability across software

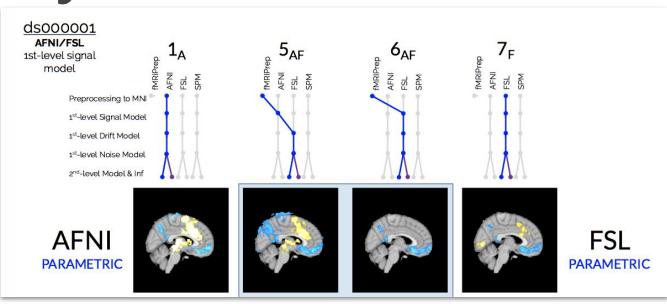


Thresholded maps in AFNI, FSL, SPM + original

[Bowring et. al, HBM 2019]

## Variability across software

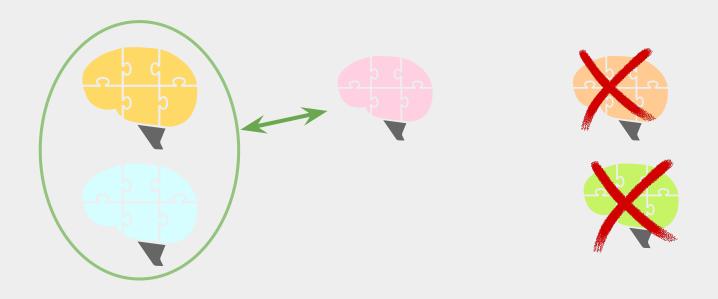




[Bowring et. al, HBM 2021]

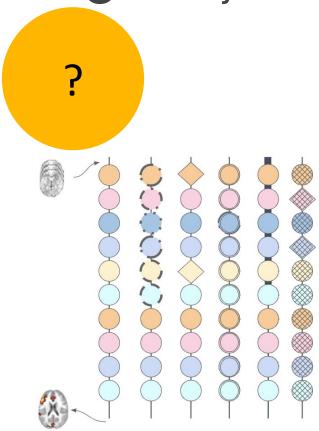
Thresholded maps in AFNI, FSL, SPM + original

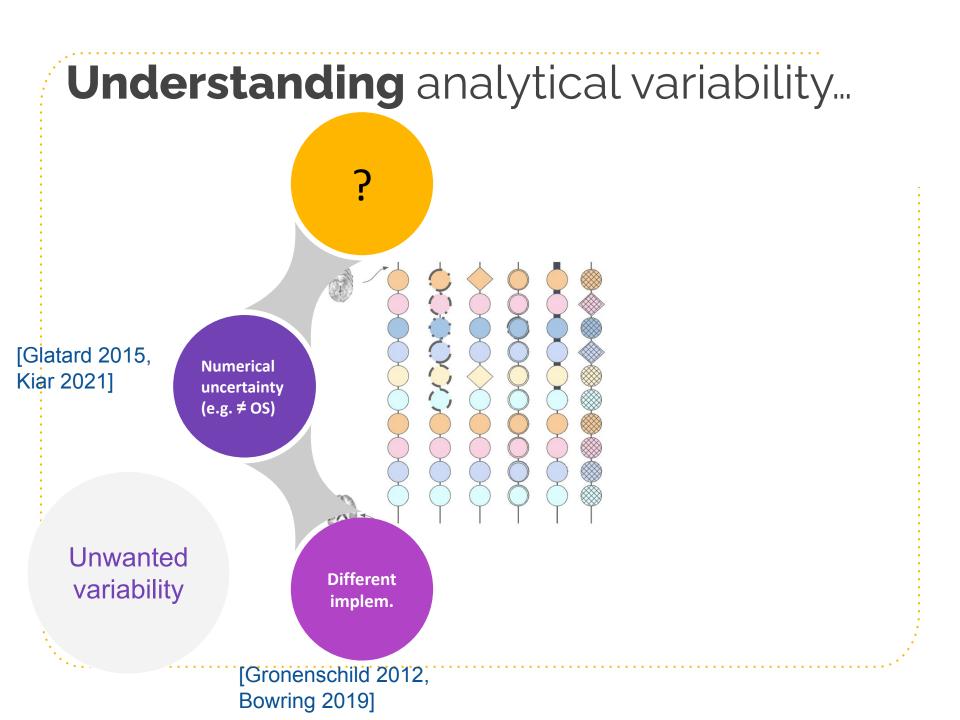
[Bowring et. al, HBM 2019]



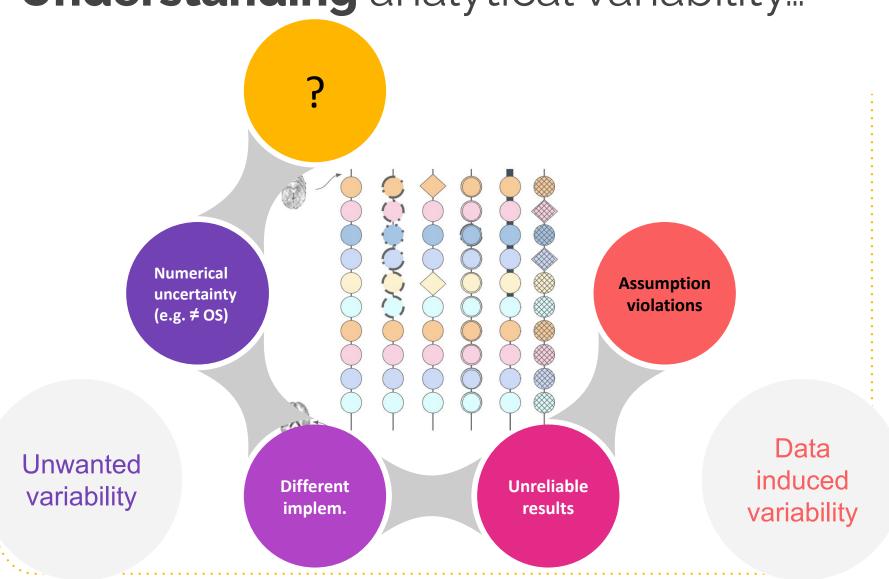
# Understanding analytical variability

#### Understanding analytical variability...

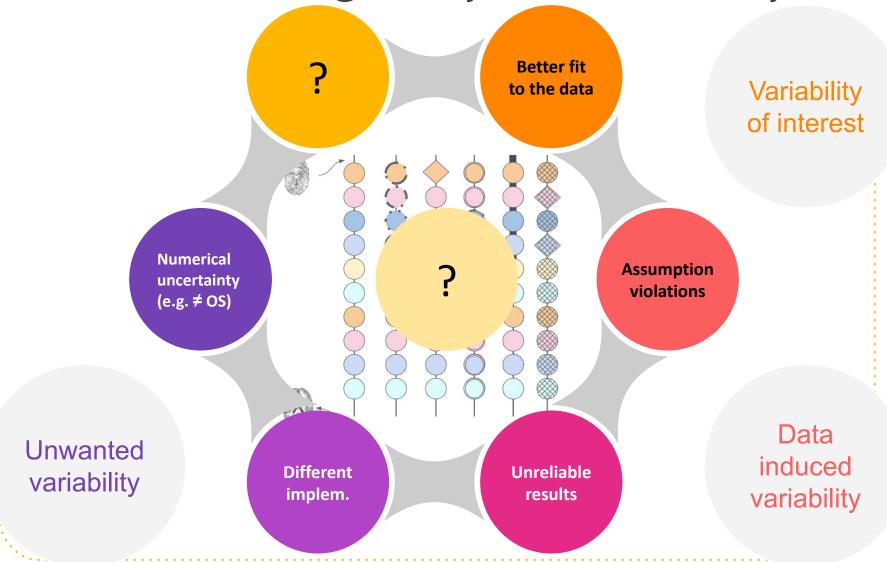


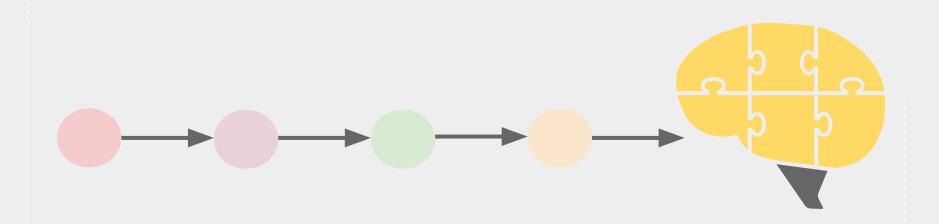


#### Understanding analytical variability...



#### Understanding analytical variability...





# Describing the neuroimaging pipeline space

## NARPS open pipelines

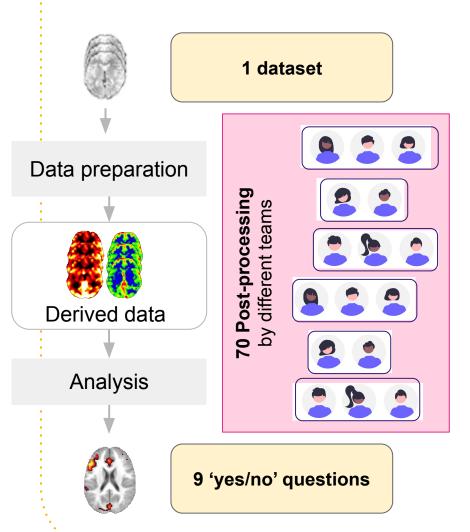






Boris Clenet

Elodie Jérémy Germani Lefort-Besnard



## NARPS open pipelines







Boris Clenet

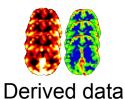
Elodie Germani

e Jérémy ıni Lefort-Besnard



#### 1 dataset

Data preparation



Denveu uala

Analysis



9 'yes/no' questions

70 Post-processing by different teams



- (Re)-coding the 70 pipelines
- Collaborative effort



#### Project NARPS open pipelines

Join us at the Brainhack Global!

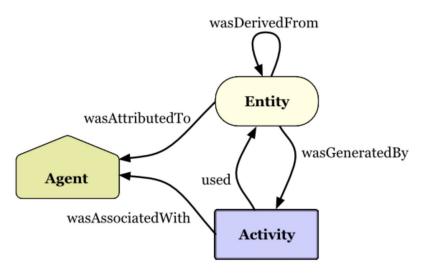


https://github.com/Inria-Empenn/narps\_open\_pipelines/

#### **Provenance**

"Provenance describe how an artifact or set of artifacts was produced" Wikipedia

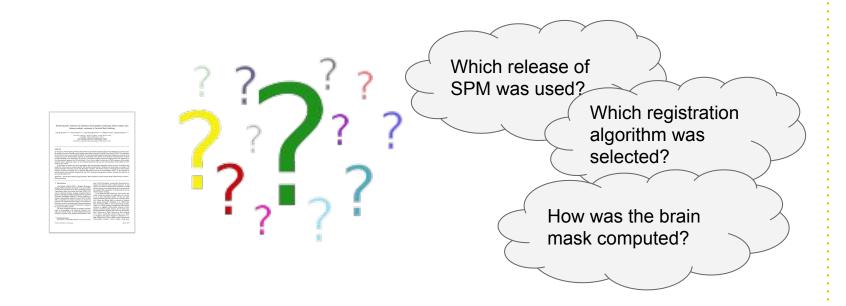
"Provenance is information about entities, activities, and people involved in producing a piece of data [...]" W3C PROV.



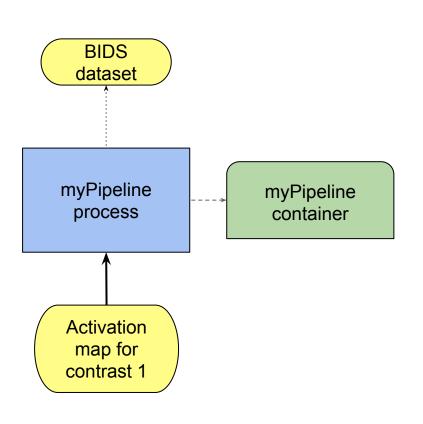
W3C PROV model

https://www.w3.org/TR/2013/NOTE-prov-primer-20130430/





reproduce
exactly the same
results
(same data, same
methods)?



Can we
reproduce
exactly the same
results
(same data, same
methods)?

### Provenance Use Case 2 Dealing with unwanted variability

A previously widely-used parameter is no longer considered as best practice

### Provenance Use Case 2 Dealing with

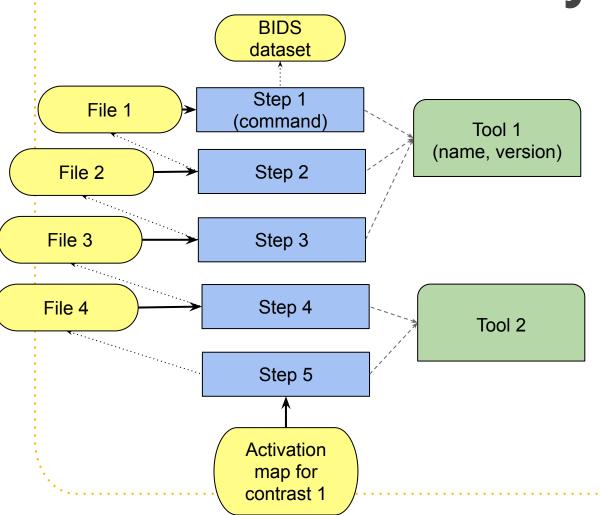
unwanted variability

A previously
widely-used
parameter is no
longer considered
as best practice

A bug was found in a given tool version, can we retrieve all corresponding results?

#### Provenance Use Case 2 Dealing with

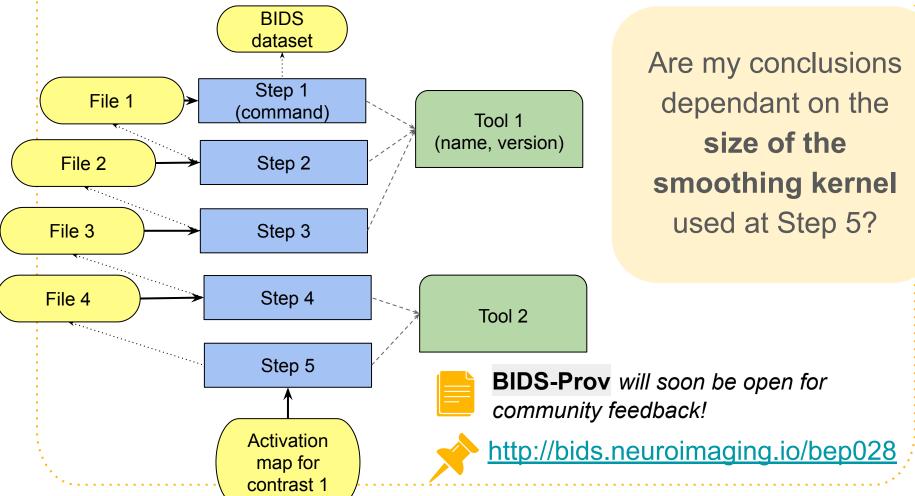
unwanted variability

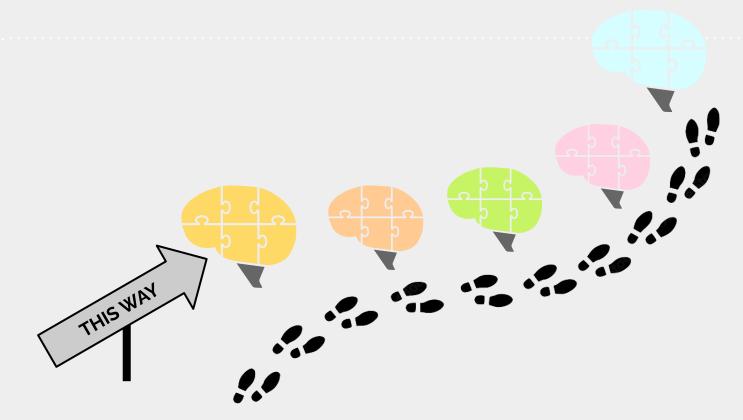


A previously
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A bug was found in a given tool version, can we retrieve all corresponding results?

# Provenance Use Case 3 Exploring factors impacting analytical variability





# Which solutions for practitioners?

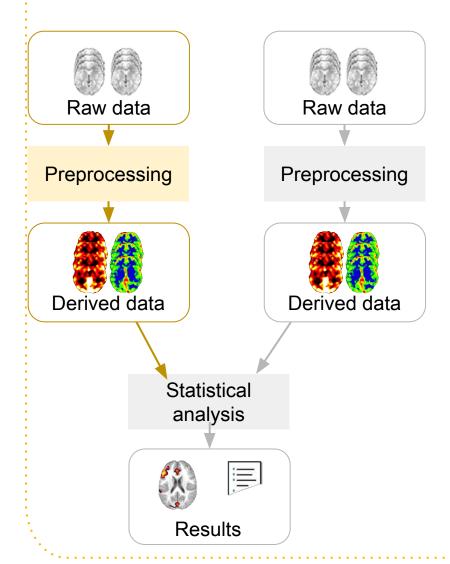
#### Data re-use







Pierre Maurel



#### Data re-use

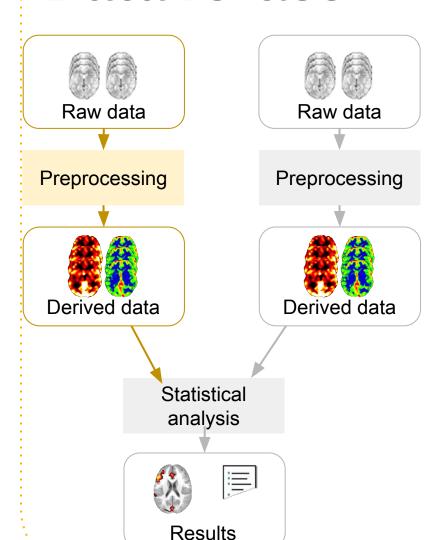


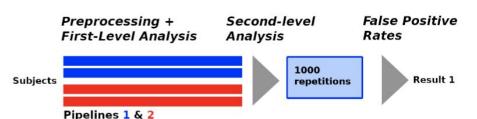


Xavier Rolland

Pierre Maurel

- Using HCP data under the null
- Comparisons of pipelines 2-by-2
- 3 variations
  - Smoothing: 5 or 8
  - HRF derivative: yes or no
  - Motion regressors: 0, 6, 24
- "Default": 5mm, 24 regressors, no temp. deriv.





#### Data re-use

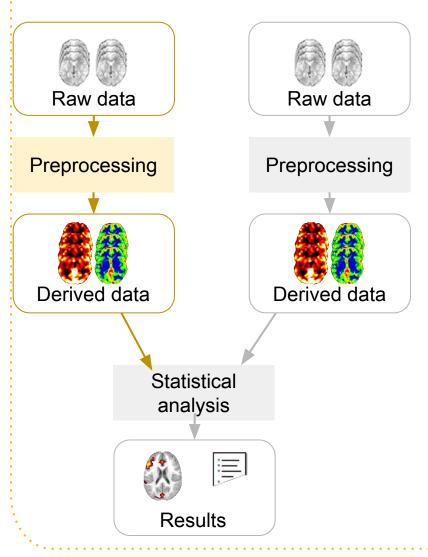


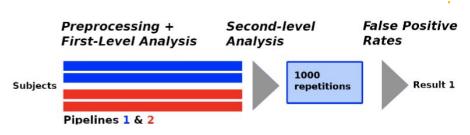


Xavier Rolland

Pierre Maurel

- Using HCP data under the null
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Default
Presence of derivatives
8mm for smoothing
no motion regressors

0.5

-0.5

-0.5

expected -log10(p-values)

[Rolland et. al, ISBI 2022]

### Building on large and more

Elodie Germani

Elisa Fromont

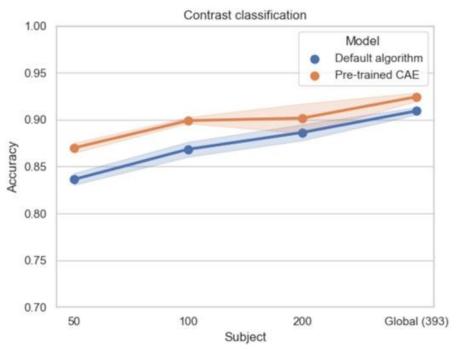
diverse samples

Latent representation of fMRI statistic maps based on

NeuroVault

Apply to a decoding task





[Germani et. al, Gigascience 2023]

https://inria.hal.science/hal-03769993

#### Same data meta-analysis



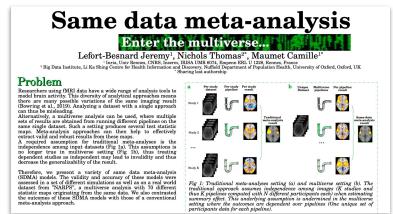


Jérémy Tom Lefort-Besnard Nichols

 A statistical approach to combine the results of different pipelines and obtain a single (consensus) summary

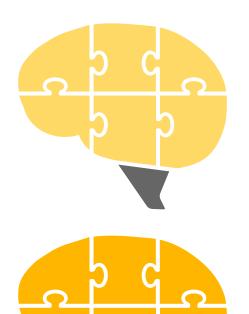


**Poster** Systematic review and evaluation of meta-analysis methods for same data meta-analyses **Jérémy Lefort-Besnard** 





https://inserm.hal.science/inserm-04222730



Towards reproducible neuroimaging across different analysis pipelines

**Camille Maumet** 

### Thank you!



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**Credit**: Presentation template and icons by <u>SlidesCarnival</u>, adapted



@cmaumet@fediscience.org

