

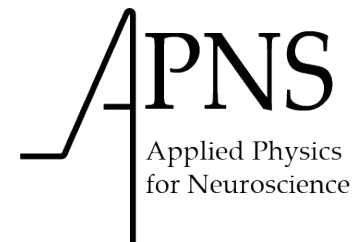
Investigating genetic factors driving brain morphology through Neuroimaging studies

Fabrizio Pizzagalli

fabrizio.pizzagalli@unito.it



Department of Neuroscience
University of Turin



Outline

Context

Why Imaging Genetics ?

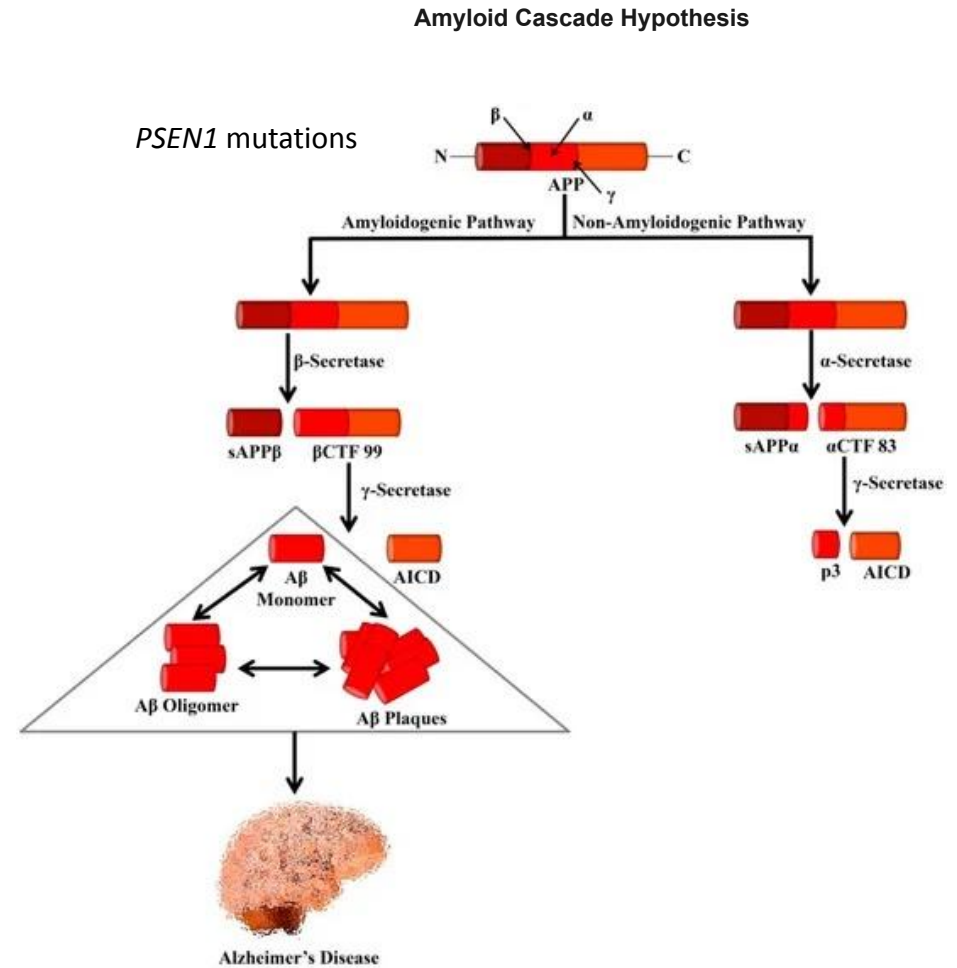
Genetics factors driving brain morphology

- The genetic architecture of the human cerebral cortex
- Sulcal-based morphometry
- Brain folding heritability
- The genetic architecture of brain folding

Why Imaging Genetics?

Genetics:

- Understand biological mechanisms



Why Imaging **Genetics**?

Genetics:

- Understand biological mechanisms
- Treatments

Treatments may not work in people with certain genetic variants in a given gene.

Why Imaging **Genetics**?

Genetics:

- Understand biological mechanisms
- Treatments
- Assess individual risk

Knowing risk for disease may change lifestyle habits and medical decisions.

Why Imaging Genetics?

Imaging

- Complex Diseases

Diseases or disorders most likely caused by a largely unknown combination of multiple genetic, environmental, and lifestyle factors.

E.g.:

- Alzheimer's disease
- Autism
- Schizophrenia
- Bipolar Disorder
- & more

Why **Imaging** Genetics?

Imaging

- Complex Diseases

Issues:

- The disease may have a latent period.
- They ignore effects that are stage-specific.
- The disease may be difficult to diagnose.

Why **Imaging** Genetics?

Imaging

- Complex Diseases

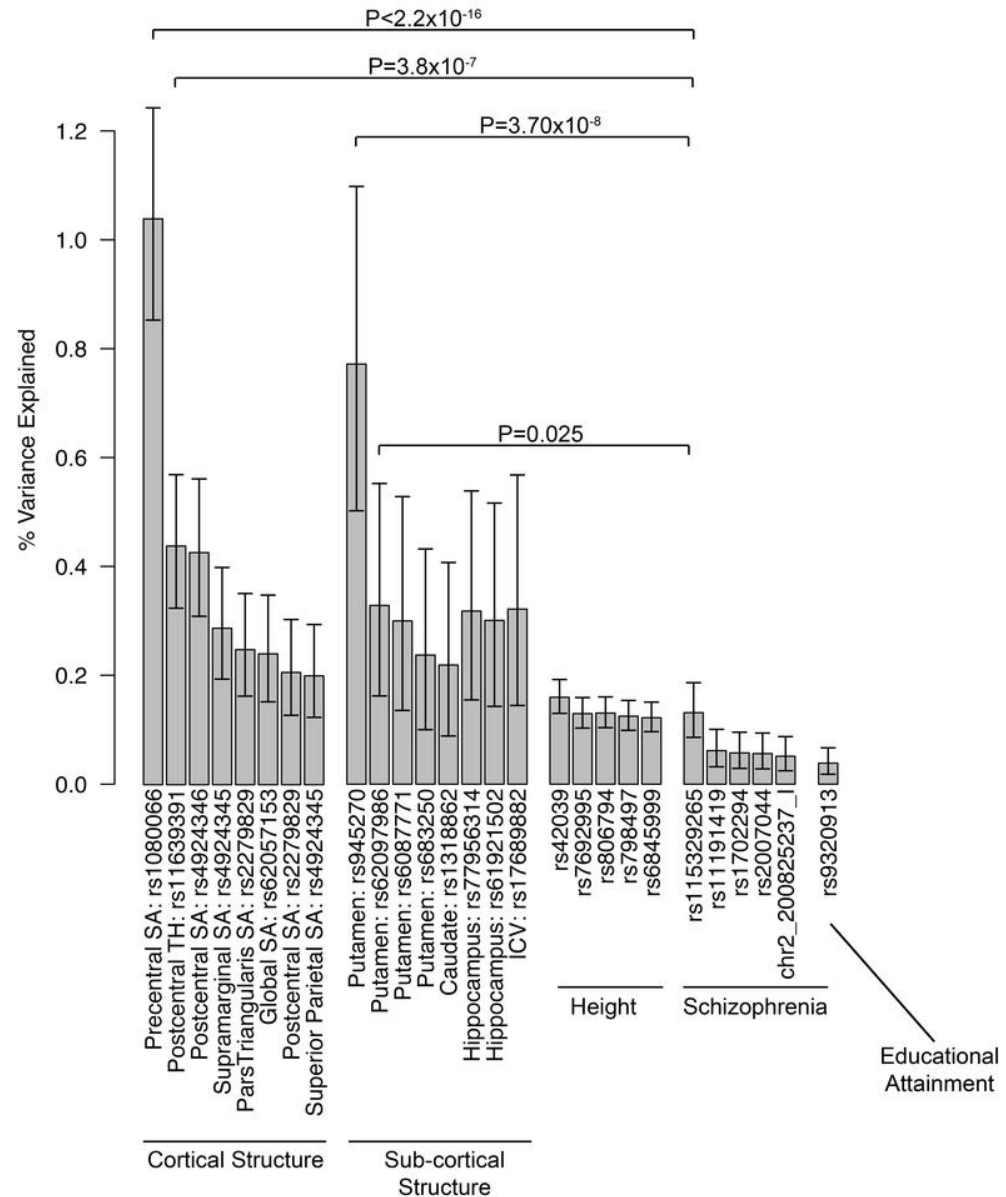
Imaging Genetics

- We can investigate vulnerabilities to disease and stage-specific effects.
- Genetics may be more closely related to brain measures than to disease diagnosis.

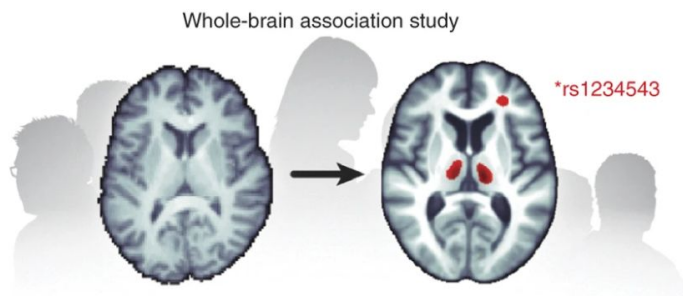
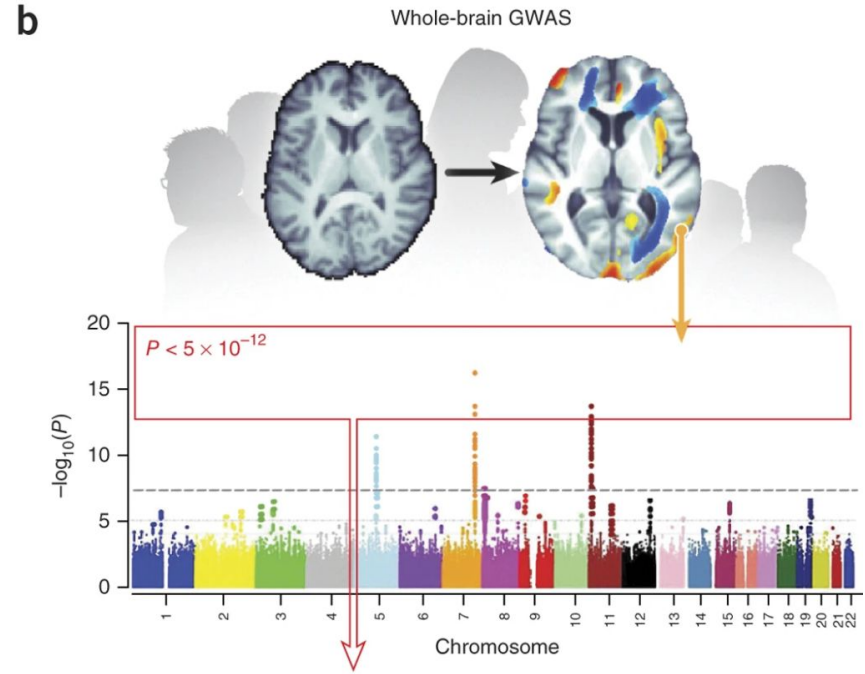
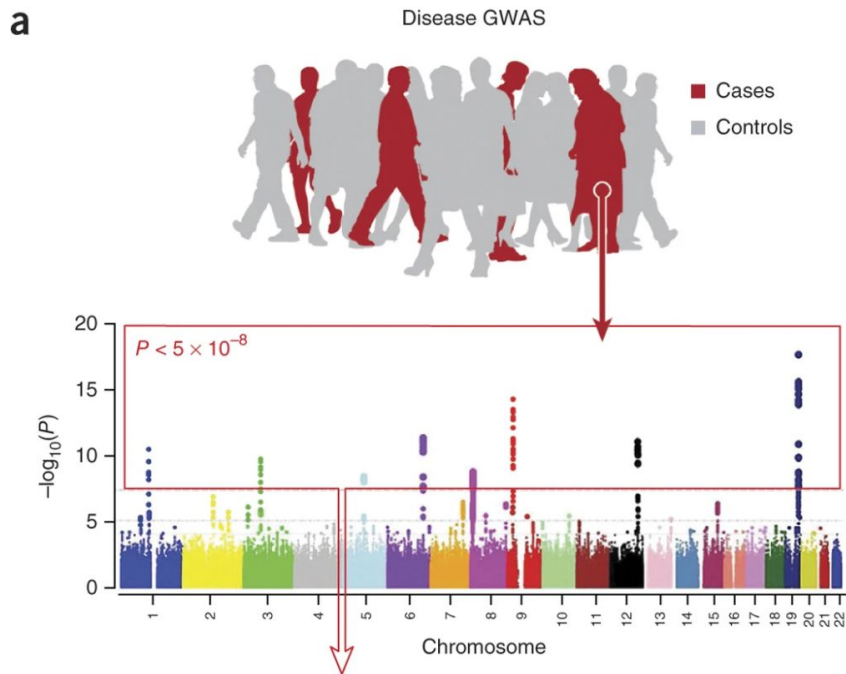
Imaging Genetics

The idea behind imaging genetics is that “genetic variation has a stronger impact on brain traits than on heterogeneous and behaviorally defined disease categories, and that genetic variation associated with brain changes will allow a greater understanding of the mechanism leading to risk for behaviorally defined disorders” (Le and Stein 2019).

Le and Stein, *Psychiatry Clin Neurosci*, 2019 Jul;73(7):357-369. doi: 10.1111/pcn.12839. Epub 2019 May 21.



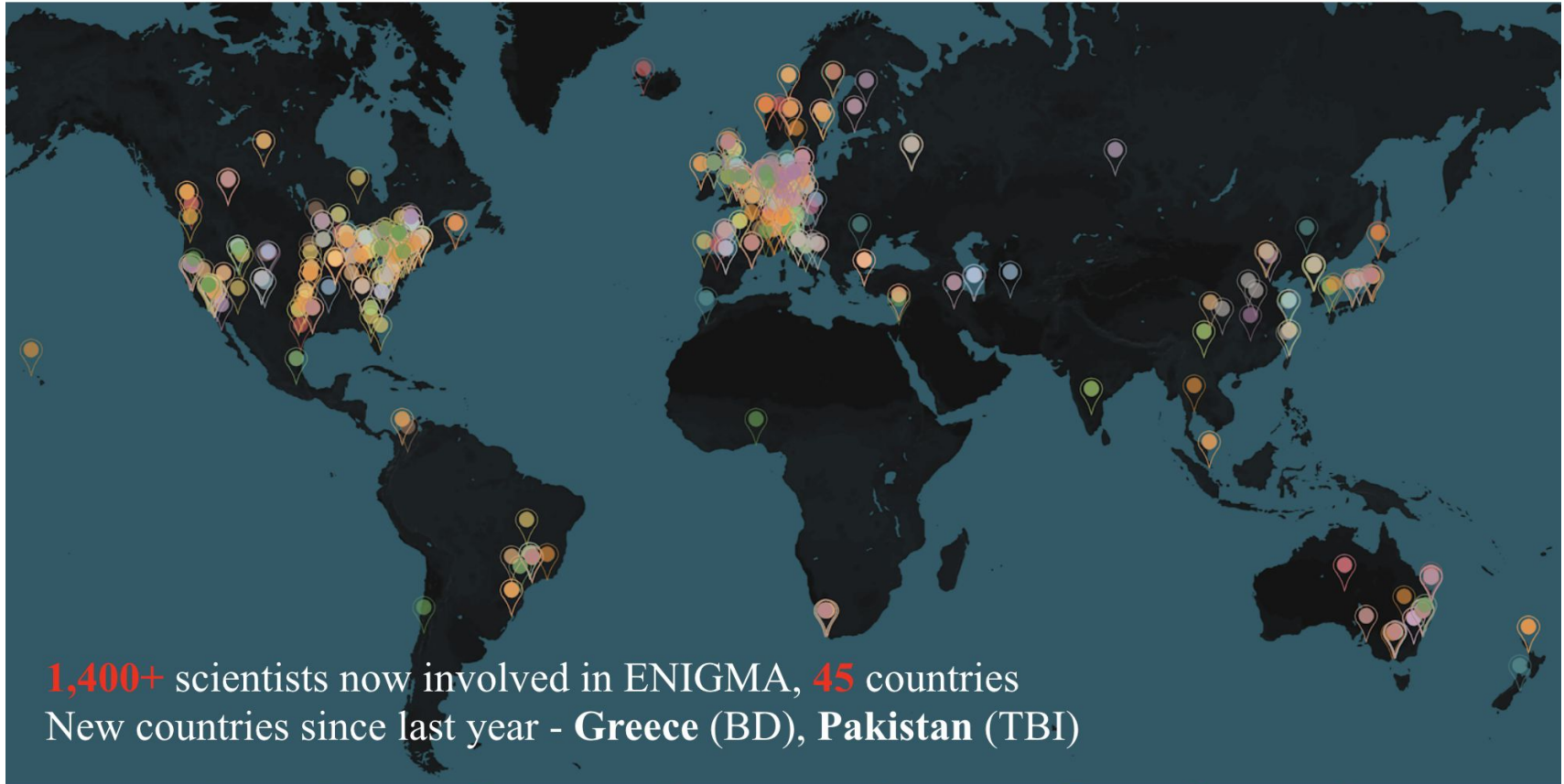
Imaging Genetics study designs



Imaging Genetics study designs

Need for large N!

- **UK Biobank** (Elliott et al., 2018)
- **CHARGE** (Cohorts for Heart and Aging Research in Genomic Epidemiology)
(Bis et al., 2012)
- **ENIGMA** (Enhancing Neuroimaging Genetics through Meta-Analysis)
(Thompson et al., 2020)



1,400+ scientists now involved in ENIGMA, **45** countries
 New countries since last year - **Greece (BD)**, **Pakistan (TBI)**

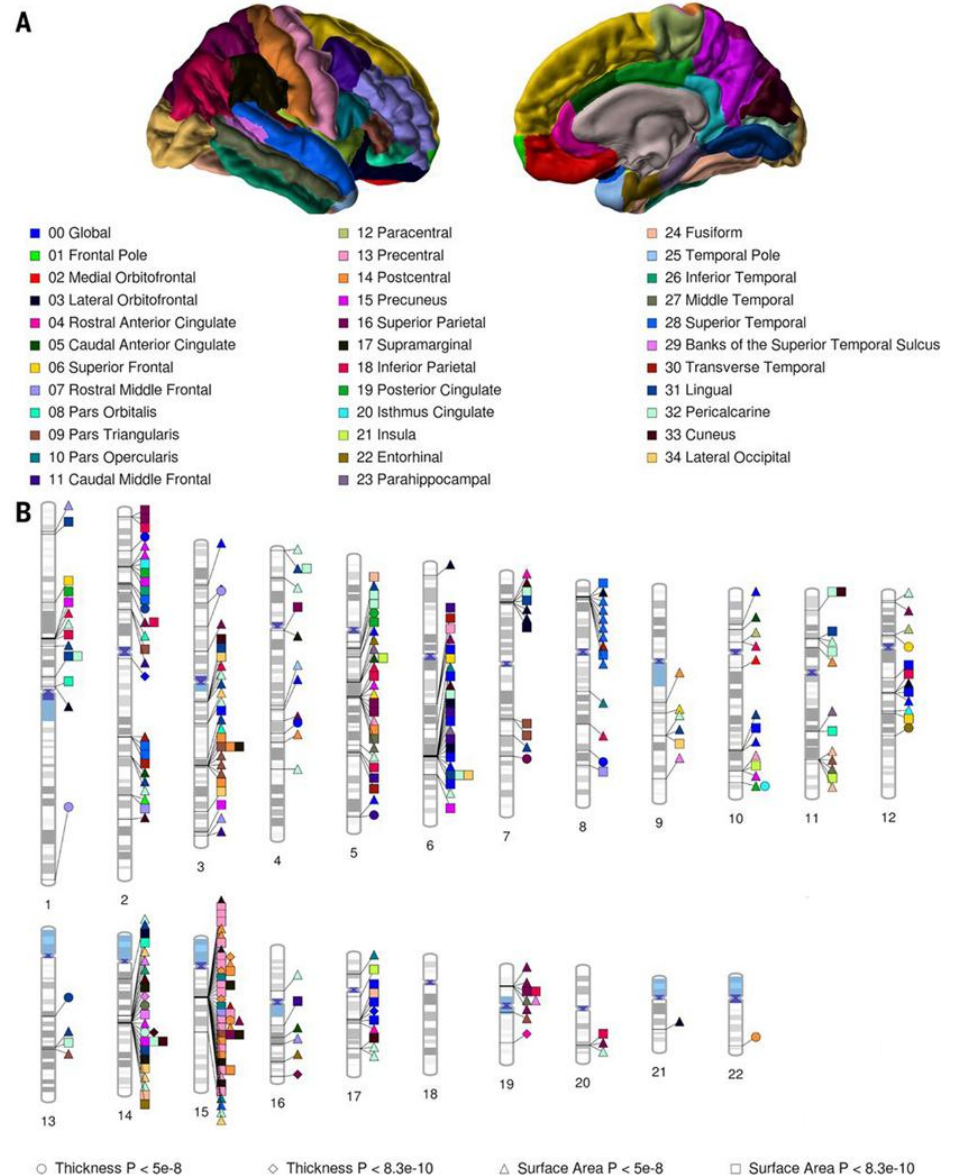
Primary Working Group	SCZ	Ataxia	Core	Epigenetics	Hippocampal subfields	OCD	Relatives	Stroke Recovery
22q Deletion Syndrome	Addiction	Brain Injury	DTI	Epilepsy	HIV	ORIGINs	rsfMRI	Suicidal Thoughts & Behaviors
MDD	ADHD	BrainAGE	Early Onset Psychosis	Evolution	Irritability	Parkinson's	Schizotypy	Sulci
BD	Anxiety	Clinical High Risk	Eating Disorders	Frontotemporal Dementia	Laterality	Plasticity	Sleep Disorders	Tourette's Syndrome
	ASD	CNVs	EEG	GWAS	Lifespan	PTSD	Storage Disease	Transgender Persons

<https://enigma.ini.usc.edu/>

The genetic architecture of the human cerebral cortex

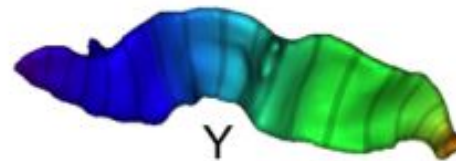
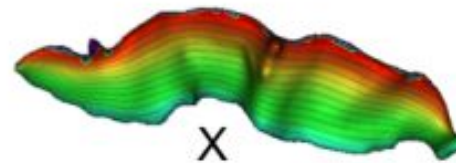
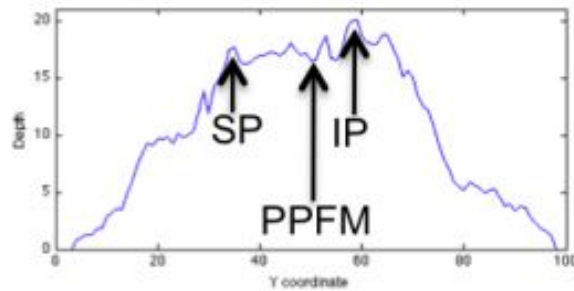
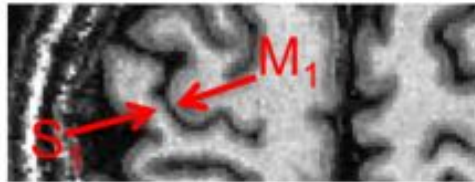
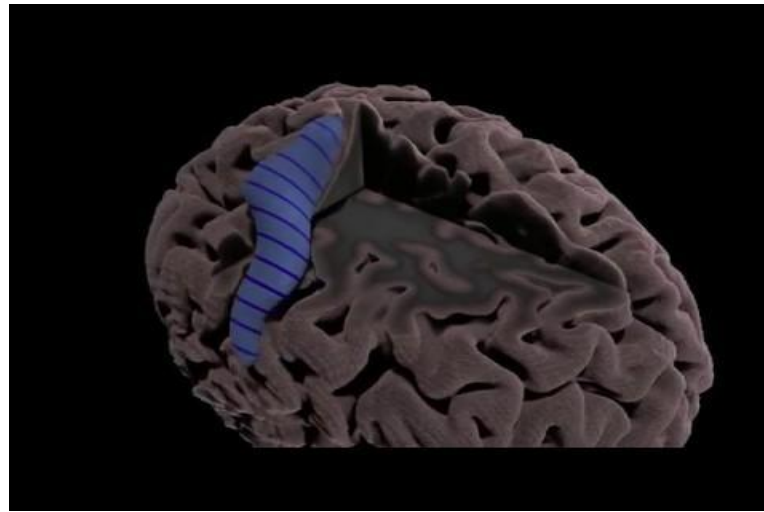
- Genome-wide association meta-analysis
- MRI data from **51,665 individuals** across **60 cohorts**
- Surface area and average thickness of the whole cortex and 34 cortical regions with known functional specializations were analyzed

Grasby et al., Science 2020



The genetic architecture of the human cerebral cortex

Central sulcus depth profile

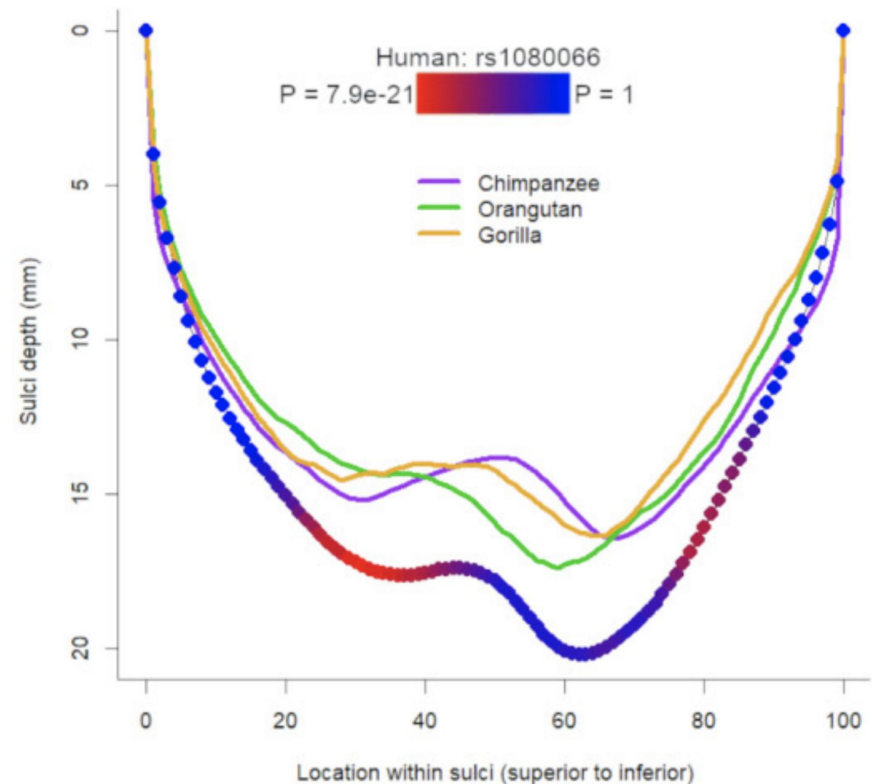


Central sulcus depth profile

- Central sulcal depth profile in UK Biobank

Strongest regional association was observed on chromosome 15q14 with the precentral SA (rs1080066). The maximal association between rs1080066 and sulcal depth was observed around the *pli de passage fronto-pariétal moyen*.

○ N ~ 10,000



Grasby et al., Science 2020

Outline

Context

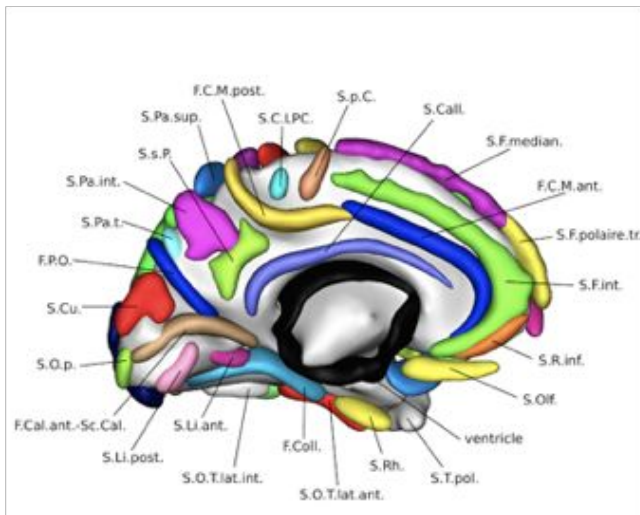
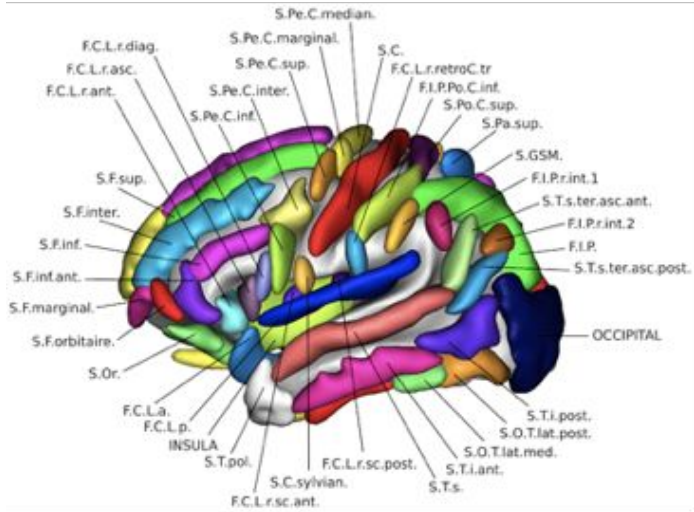
Why Imaging Genetics ?

Genetics factors driving brain morphology

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- **Sulcal-based morphometry**
- Brain folding heritability
- The genetic architecture of brain folding

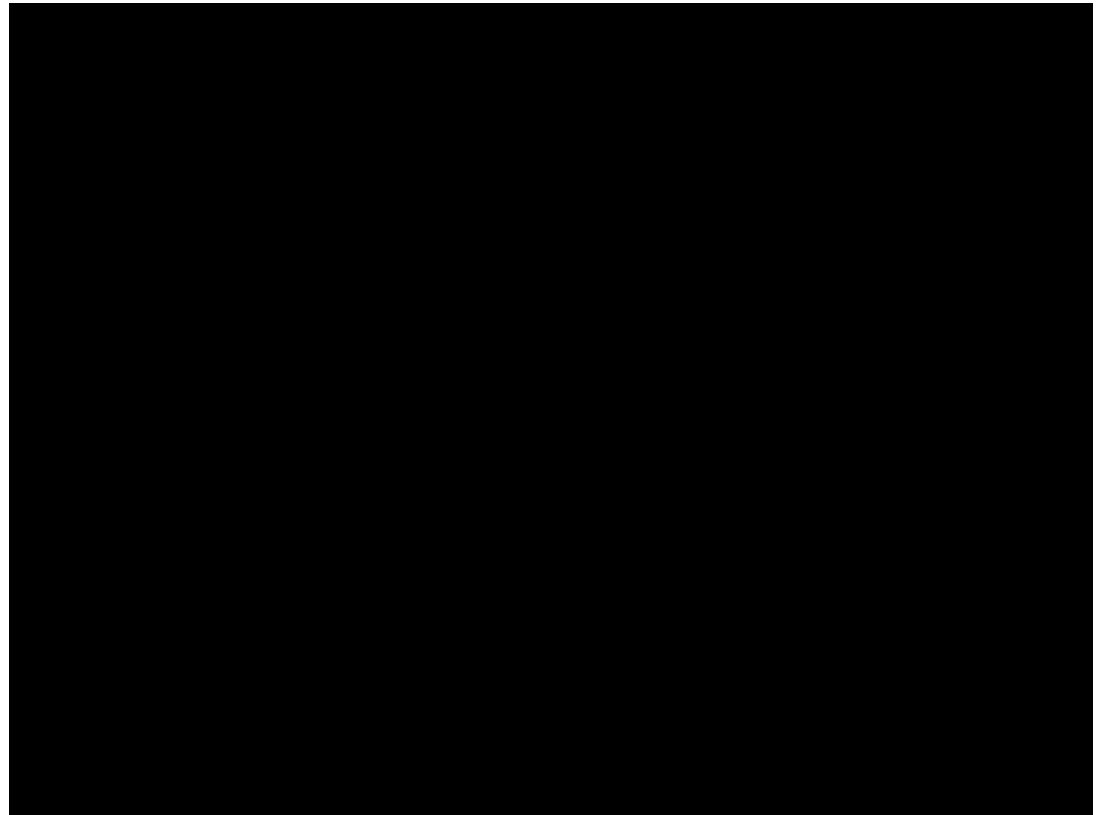
Sulcal-based morphometry

Brain folding



BrainVISA segmentation and labelling

<https://brainvisa.info/web/>



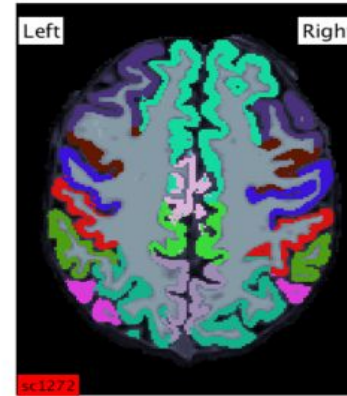
Perrot et al., Med Image Anal. 2011 Aug;15(4):529-50. [doi: 10.1016/j.media.2011.02.008](https://doi.org/10.1016/j.media.2011.02.008).

Sulci segmentation and labeling protocol

- 1) FreeSurfer

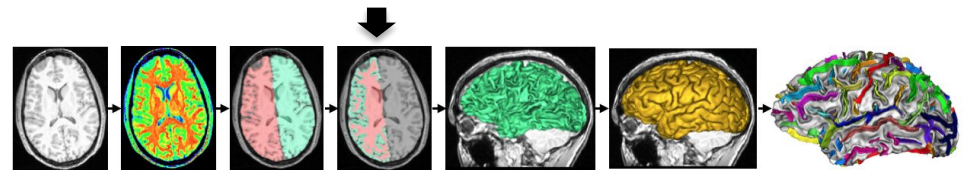
Gray/white matter segmentation

<http://freesurfer.net/>



QC-ed FreeSurfer → BrainVISA

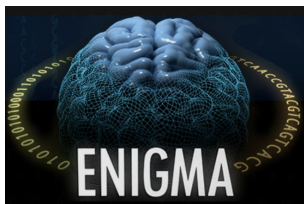
FreeSurfer Gray/white matter maps



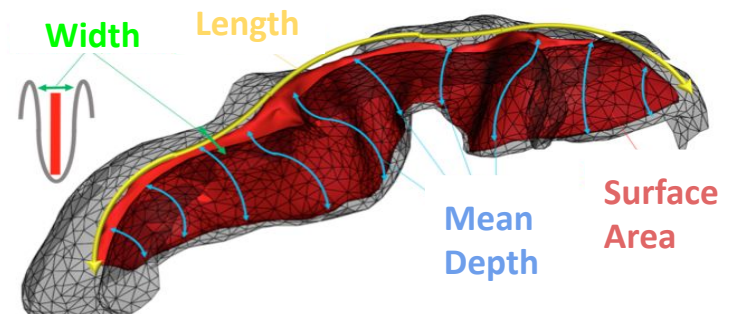
- 2) BrainVISA

<http://brainvisa.info/web/index.html>

From FS gray/white matter surface segments and labels 123 sulci

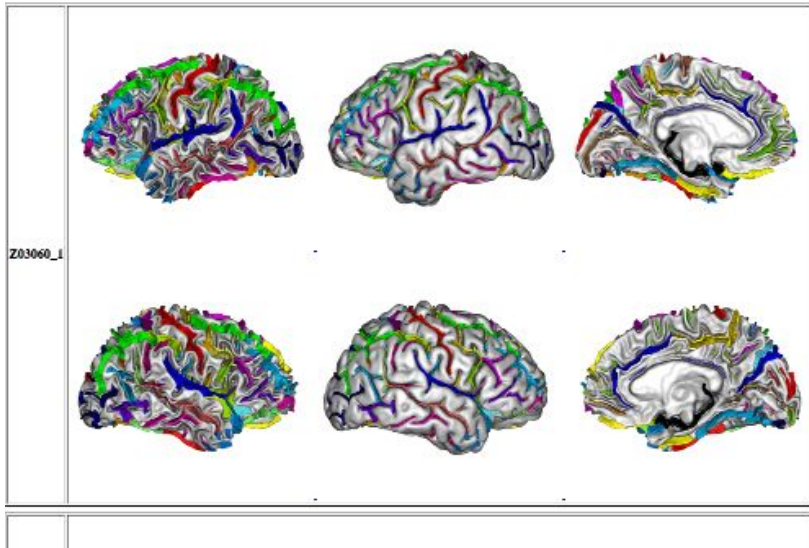


- T1-weighted images
- 123 left/right regions
- Automatic

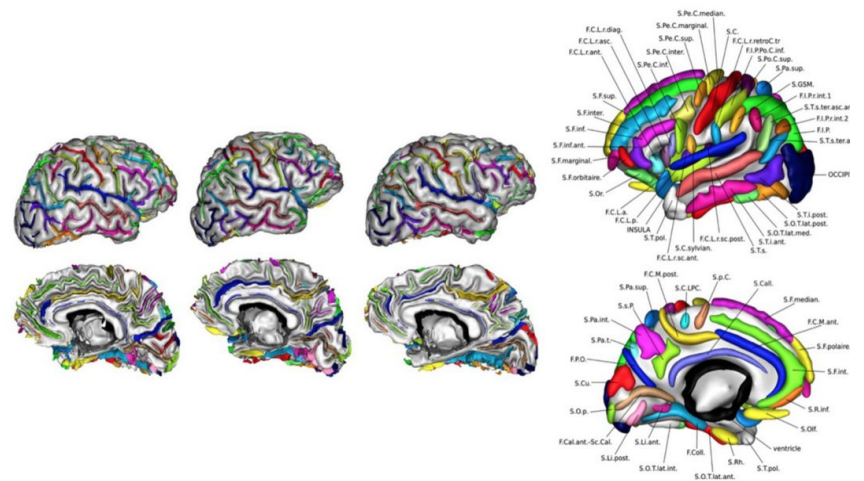


Sulci segmentation and labeling protocol

Quality Control Webpage



Template

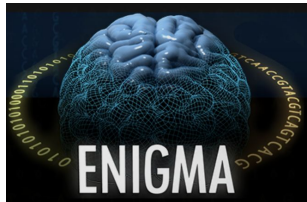


Reproduced from Perrot, M., Riviere, D., & Mangin, J. F. (2011). Cortical sulci recognition and spatial normalization. Medical Image Analysis, 15(4), 529 - 550.

Scroll through subjects with references images for comparison right on the website

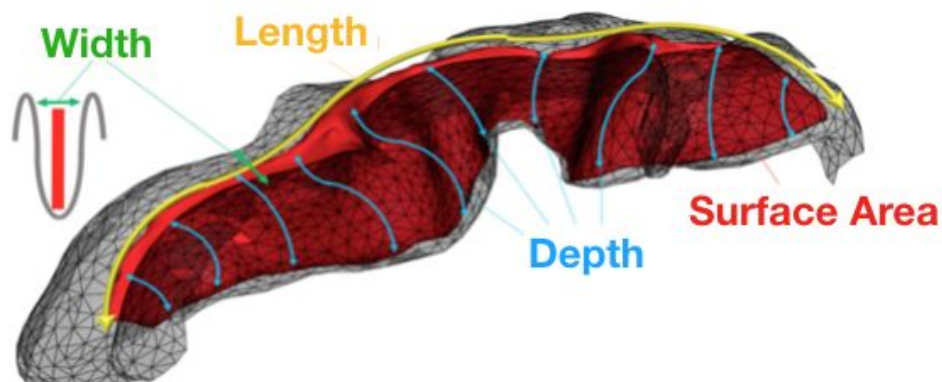
Sulci for QC: QC scale

- | |
|--|
| 5 =solid sulcus piece, no branches, consistent on WM and GM surface |
| 4 =solid sulcus piece, allow small branches, consistent on WM and GM surface (4.5 if the calcarine is splitted) |
| 3 =main sulcus is well segmented(might not be solid piece), but neighborhood needs correction |
| 2 =main sulcus piece has branches, but correct position |
| 1 =main sulcus piece has branches, incorrect position |



Reliability of sulcal traits

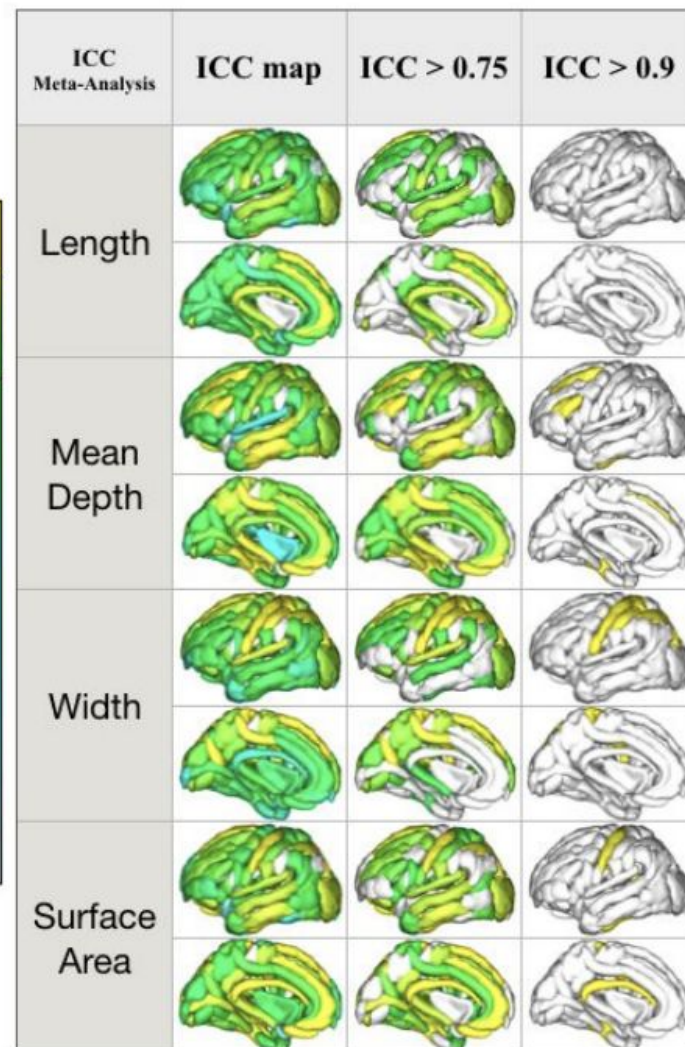
$$ICC = \frac{\sigma_{BS}^2}{\sigma_{BS}^2 + \sigma_{WS}^2}$$



A



Cohorts	Age Range (mean)	N° subject (%F)	Inter-scan Interval (days)	Field Strength [T]	Voxel size [mm] ³
KKI	22 – 61 (31.8)	21 (48%)	14	3	[1 x 1 x 1.2]
HCP	24 - 35 (30.1)	35 (44%)	90	3	[0.7 x 0.7 x 0.7]
OASIS	19 - 34 (23.3)	20 (60%)	90	1.5	[1.0 x 1.0 x 1.25]
QTIM	21 – 28 (23.2)	34 (37%)	90	4	[0.94 x 0.98 x 0.98]



Pizzagalli et al., 2020

Outline

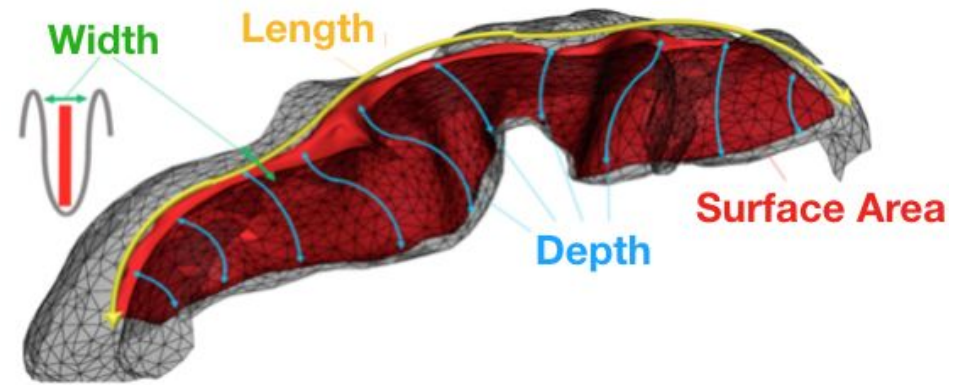
Context

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- Sulcal-based morphometry
- **Brain folding heritability**
- The genetic architecture of brain folding

Heritability of sulcal traits



Heritability estimation (narrow-sense)

$$h^2 = \frac{\sigma_g^2}{\sigma_p^2}$$

Arrows point from the labels to the terms in the equation:

- Arrows from 'Additive genetic variance' and 'Phenotypic variance' point to σ_g^2 and σ_p^2 respectively.

Left/right genetic correlation

$$Q_p = \sqrt{h_L^2} \sqrt{h_R^2} \cdot Q_G + \sqrt{1 - h_L^2} \sqrt{1 - h_R^2} \cdot Q_E$$

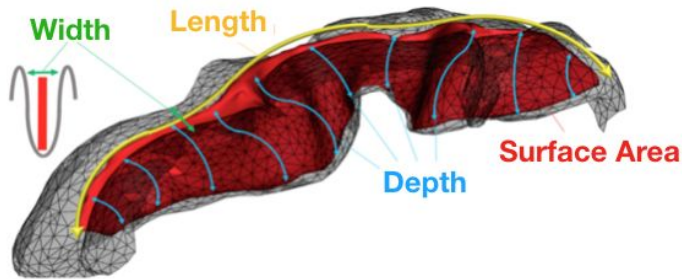
Arrows point from the labels to the terms in the equation:

- Arrows from 'Phenotypic correlation' and 'Genetic correlation' point to Q_p and Q_G respectively.
- An arrow from 'Environmental correlation' points to Q_E .

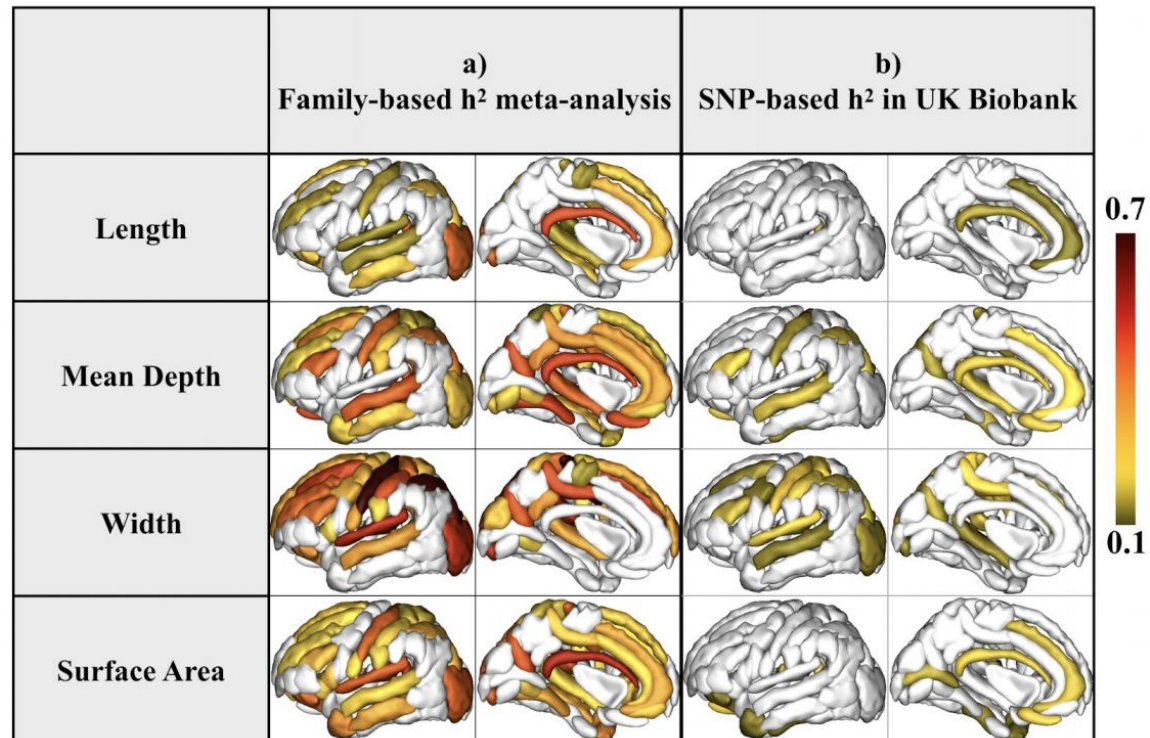
Heritability of sulcal traits

4 independent large cohorts

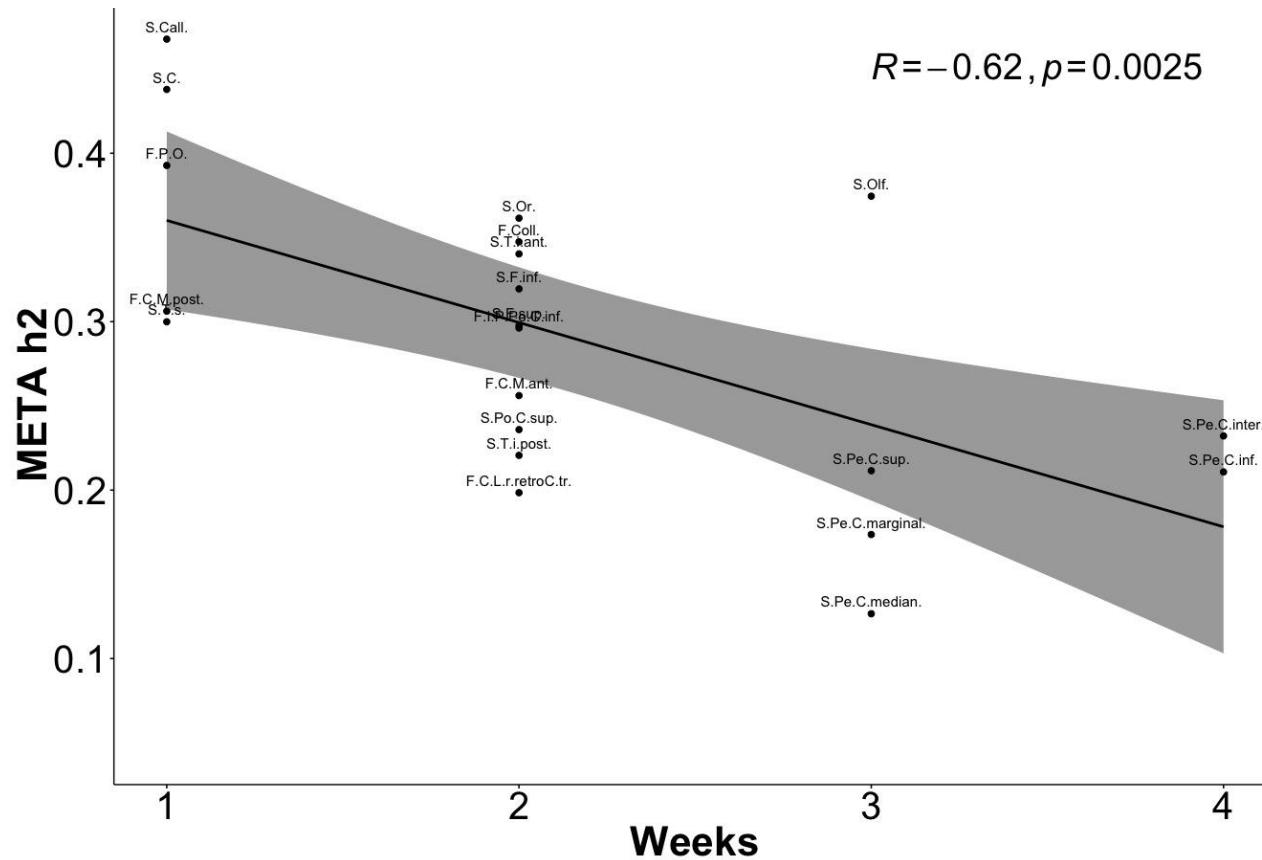
Cohort	N(%F)	Ethnicity	Age in years (mean +/- stdev [range])	Relatedness
QTIM	1,008 (37%)	Caucasian	22.7 ± 2.7 [18–30]	376 DZ 528 MZ 104 Siblings
HCP	816 (44%)	Mixed US population	29.1 ± 3.5 [22–36]	205 DZ 199 MZ and triplets 412 Siblings
GOBS	1205 (64%)	Mexican American	47.1 ± 14.2 [18–97]	71 families
UKBB	10,083 (47%)	Caucasian	62.4 ± 7.3 [45–79]	Unrelated



Pizzagalli et al., 2020



Heritability of sulcal traits



Pearson's correlation between heritability (h^2) and the **appearance of sulci** (in weeks). From Dubois et al. 2018 we grouped the sulci in four groups, 26.7w ("1"), 31.0w ("2"), 34.0w ("3") and 35.7w ("4")⁶. h^2 here has been computed as the average of the heritability of sulca length, mean depth, width and surface area, as estimated by the meta analysis of the bilaterally averages shape measures. The negative correlation ($r=-0.62, p=0.0025$) suggests that sulci appearing early in brain development are those showing higher estimated heritability.

Pizzagalli et al., 2020

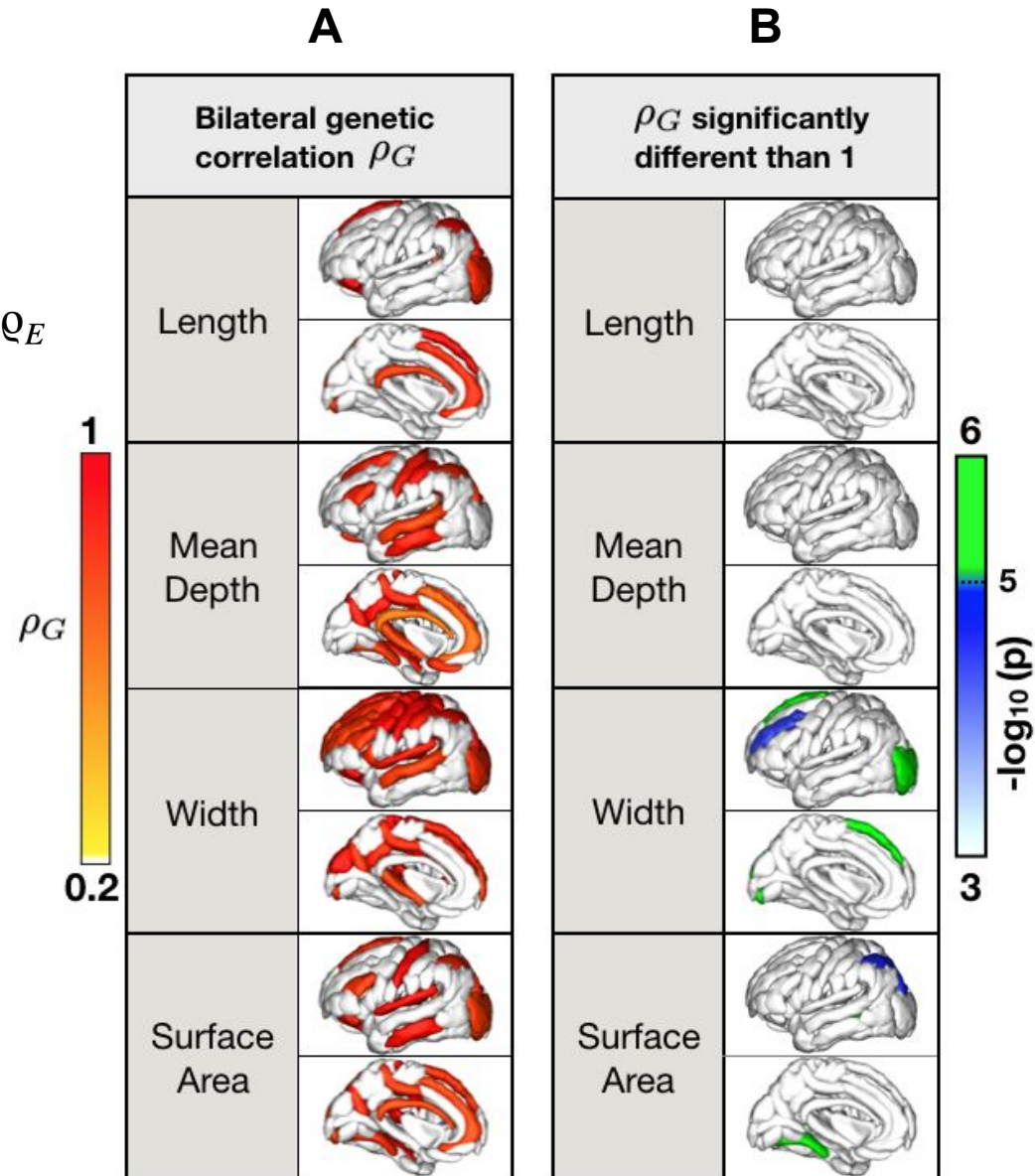
$$h_{MA-SE}^2(S) = \frac{\sum_j se_j^{-2} \times h_j^2(S)}{\sum_j se_j^{-2}}$$

Heritability of sulcal traits

Left/Right genetic correlation

$$Q_p = \sqrt{h_L^2} \sqrt{h_R^2} \cdot Q_G + \sqrt{1 - h_L^2} \sqrt{1 - h_R^2} \cdot Q_E$$

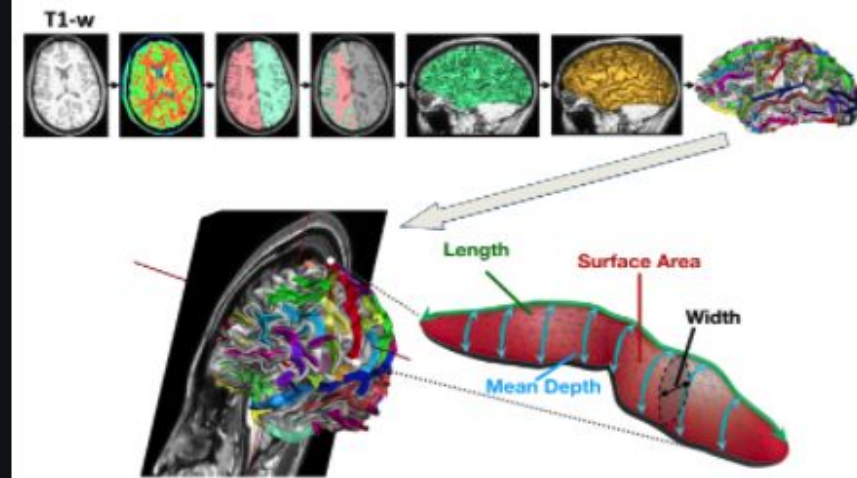
- ρ_G significantly different from zero: a significant proportion of the traits' covariance is influenced by shared genetic factors (**A**).
- Fixing $\rho_G = 1$ we tested for completely genetic overlap (**B**).



Run Sulcal-Based Morphometry

This protocol has been design to extract sulcal shape descriptors from a set of subjects, using Freesurfer (<https://surfer.nmr.mgh.harvard.edu/>) and BrainVISA toolboxes (<https://brainvisa.info/web/>).

Sulci segmentation and labeling protocol



```
#!/bin/bash
SIFFILE=/path_to/fpizzaga_sulci503_fsimport.sif
FSDIR='/myfolder/freesurfer_subjs/'
BVDIR='/myfolder/sulcal_morphometry_out/'
SUBJID=sub-01

if [ ! -d "$BVDIR" ]; then
  mkdir $BVDIR
fi

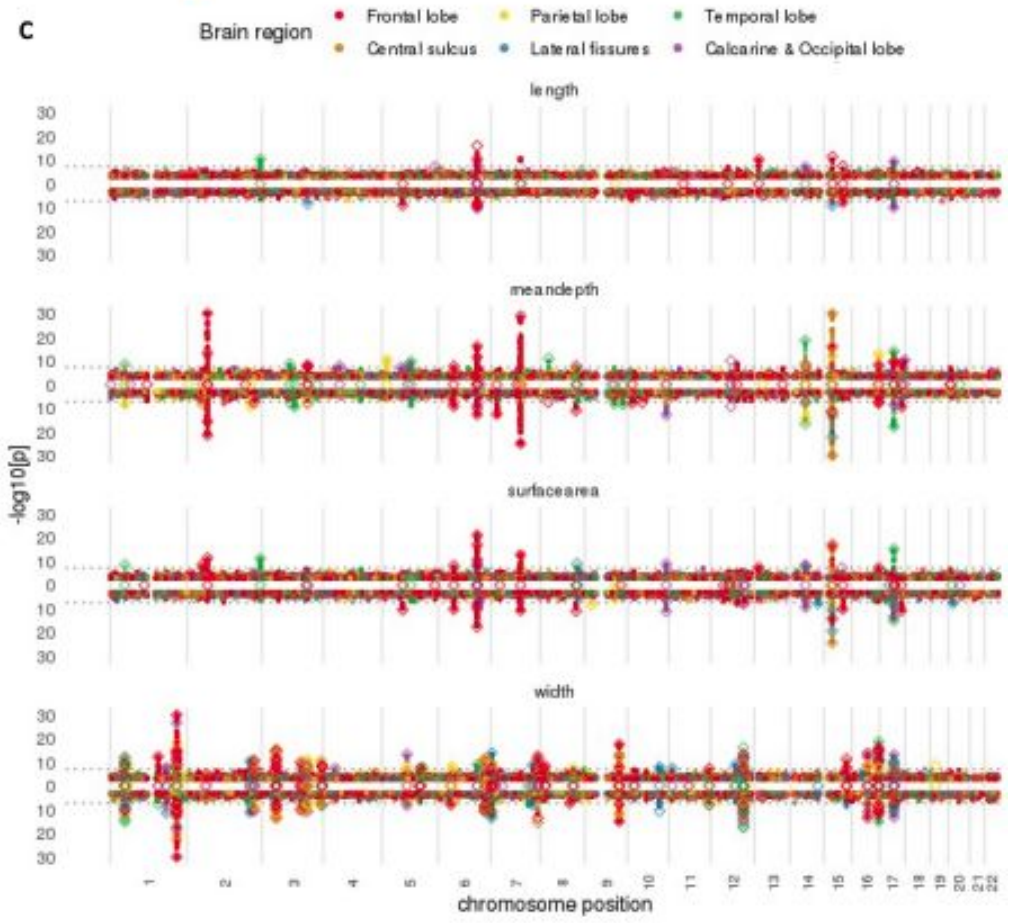
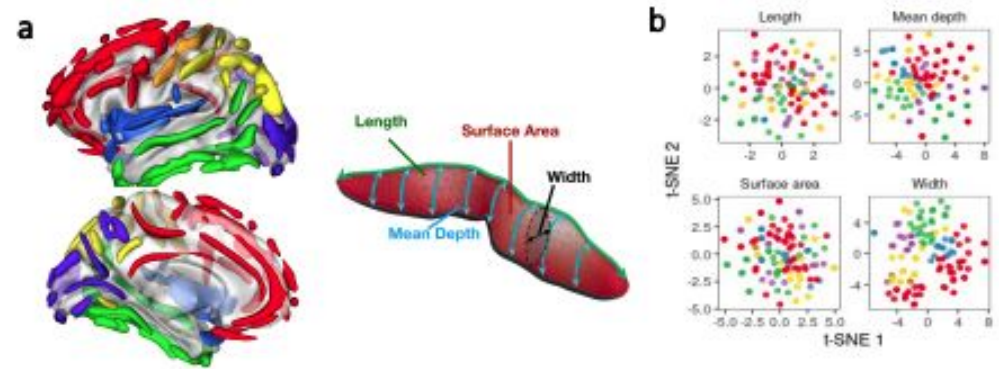
singularity run \
  -B $FSDIR:/fs_data \
  -B $BVDIR:/opt/bv_out \
  $SIFFILE $SUBJID
```

Sulcal width, surface area, length and mean depth will be extracted using command-line scripts. These scripts first import Freesurfer outputs into BrainVISA's pipeline. BrainVISA's Morphologist tool will be then used to automatically segment and label 123 sulci across the whole brain. ##Requirements

- Data pre-processed with Freesurfer
- sif scripts: please download
 - [fpizzaga_sulci503_fsimport.sif](#)
 - [fpizzaga_sulci503_sulcalMorpho.sif](#)
 - [fpizzaga_extract_bv.sif](#)
- singularity > 3.*

fabrizio.pizzagalli@unito.it

Genetic map of regional sulcal morphology in the human brain



Large-scale genome-wide associations of regional brain cortical sulcal measures derived from MRI data of **40,169 individuals** in the UK Biobank

Sun et al., <https://doi.org/10.1101/2021.10.13.463489>, biorxiv

Thank you!

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