

Analysis of structural and resting-state functional MRI data using FreeSurfer

UBDA workshop.

Manson C-M. Fong
PhD in Electronic Engineering (CUHK)
Department of Chinese & Bilingual Studies
Research Centre for Language, Cognition, and Neuroscience
December 21, 2021

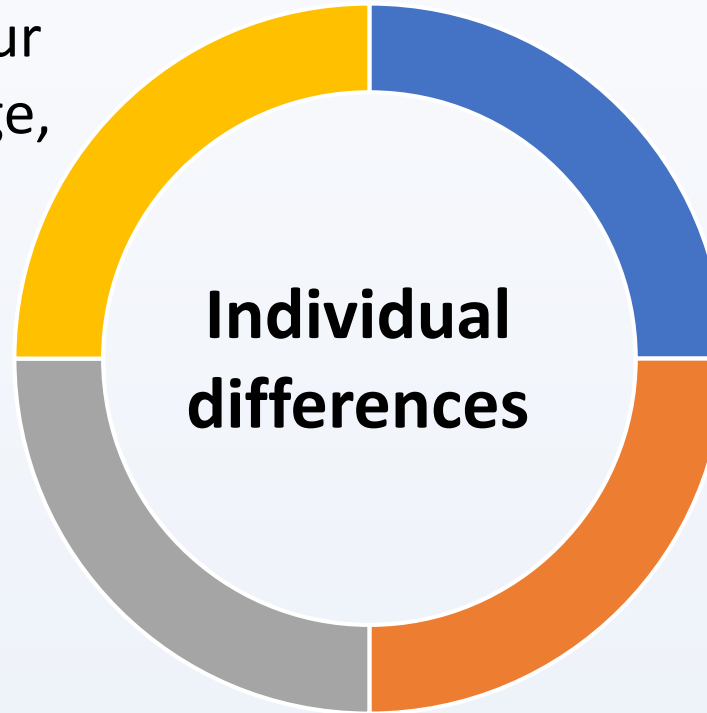
My research

Aging

- How does aging affect our brain, cognition, language, & vice versa?

Cognition

- Attention & inhibitory control
- Working memory
- Long term memory
- Fluid intelligence
- ...



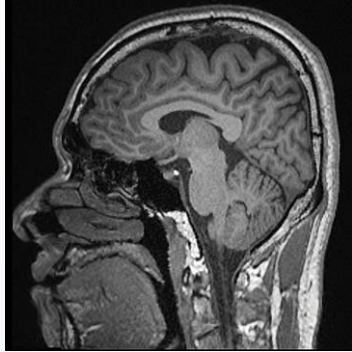
Brain

- Structural anatomy
- Functional connectivity
- Neurophysiology

Language

- Word recognition
- Semantic processing
- Topics in bilingualism

Outline



Can structural MRI inform cognitive & language functions?

Structural MRI

Anatomical correlates of language learning

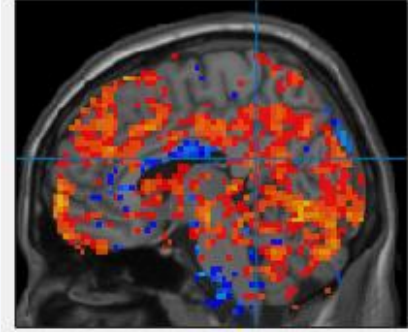
UBDA application example

Introduction

Resting-state fMRI

Brain age estimation

Ongoing analysis



Can we estimate cognitive age from resting-state fMRI?

Basic sub-divisions of the brain

Corpus callosum
(fibres connecting
both hemispheres)

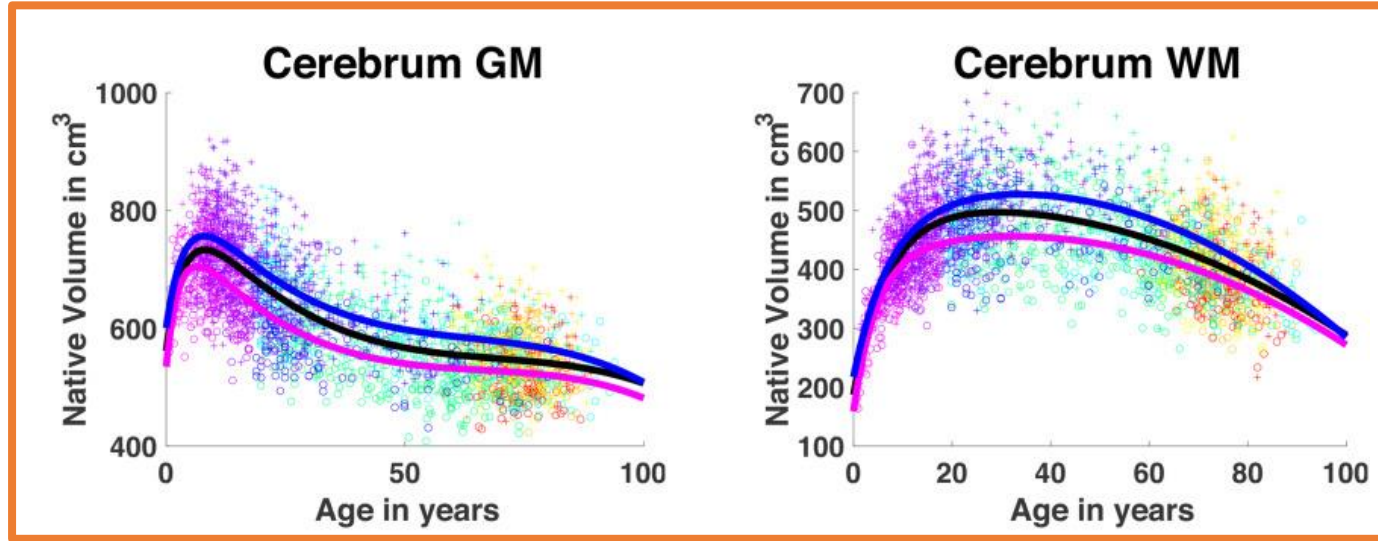
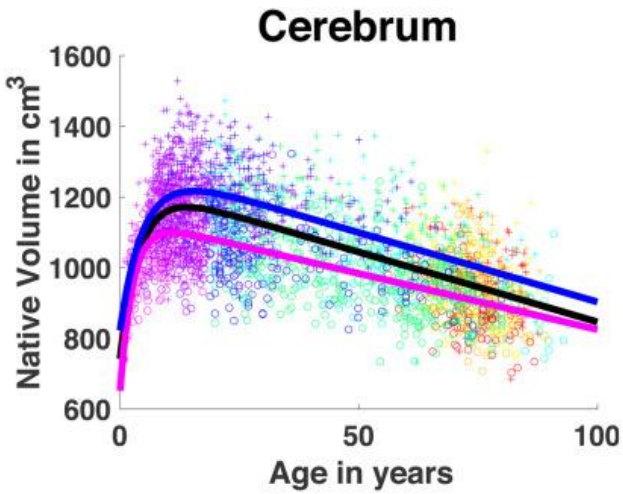
Brainstem



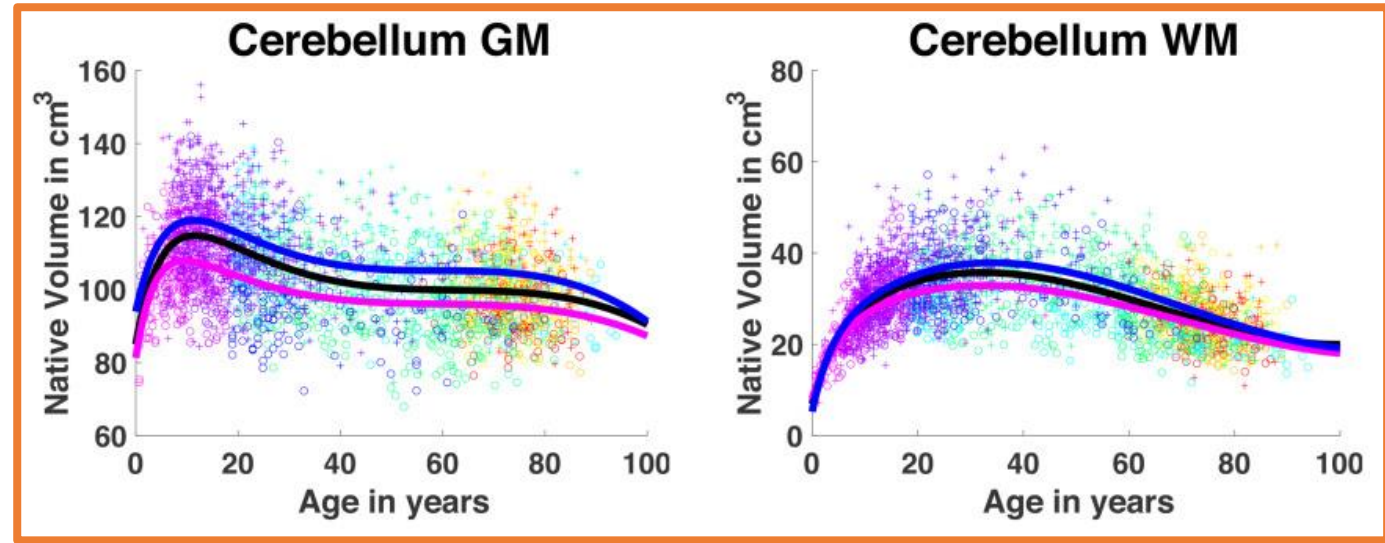
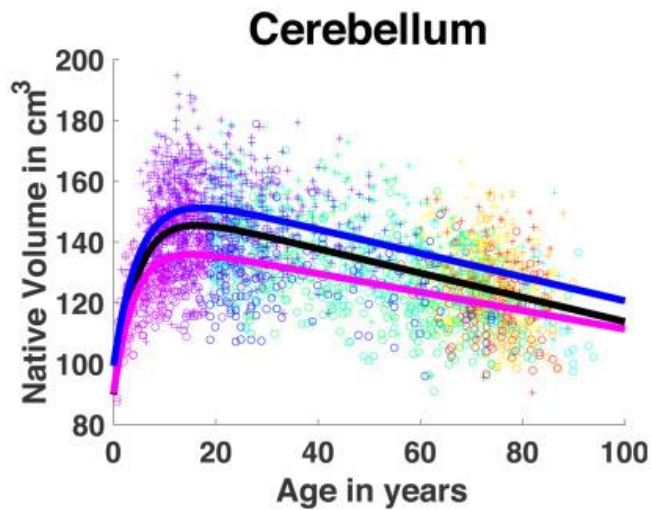
Cerebrum

Cerebellum

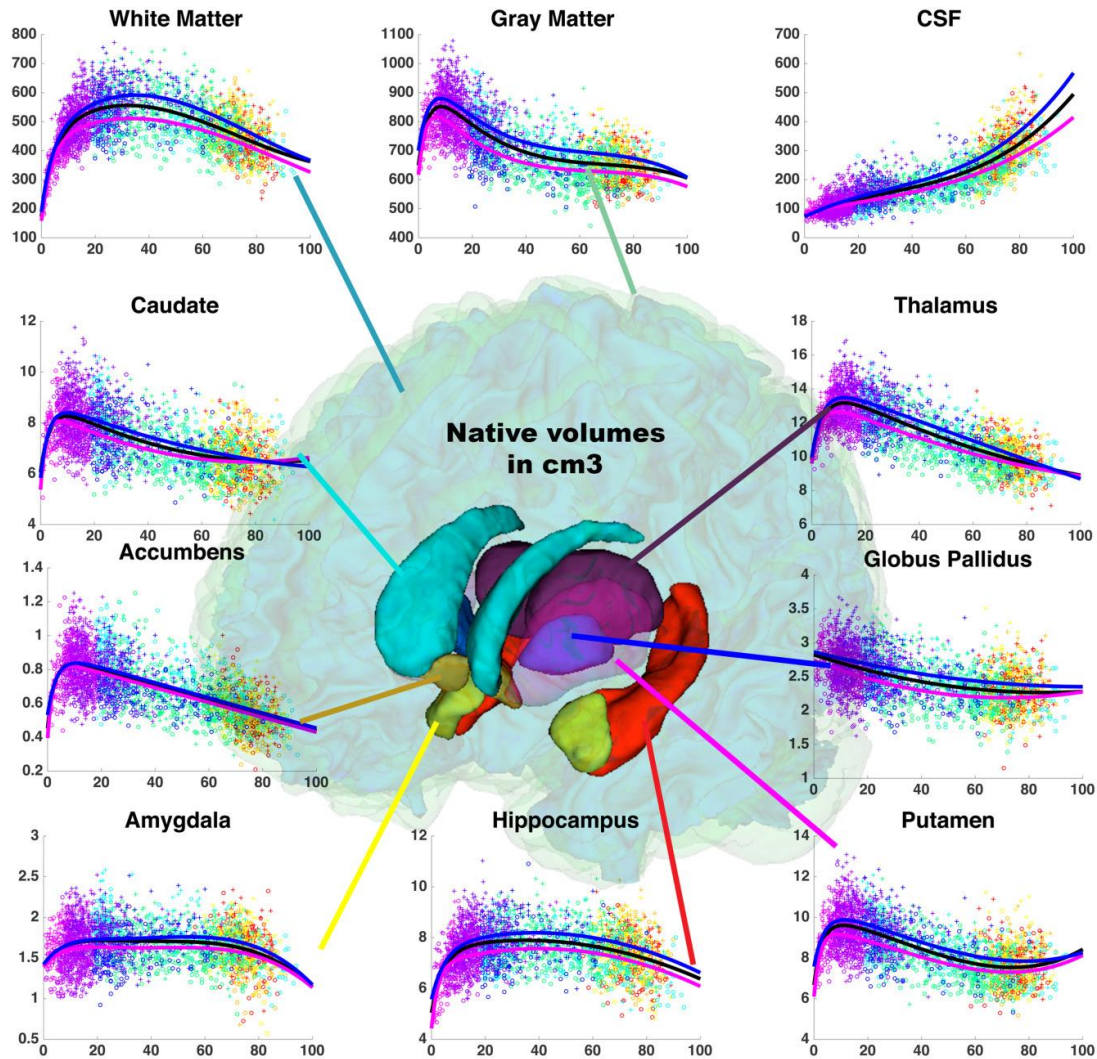
Age-related brain changes



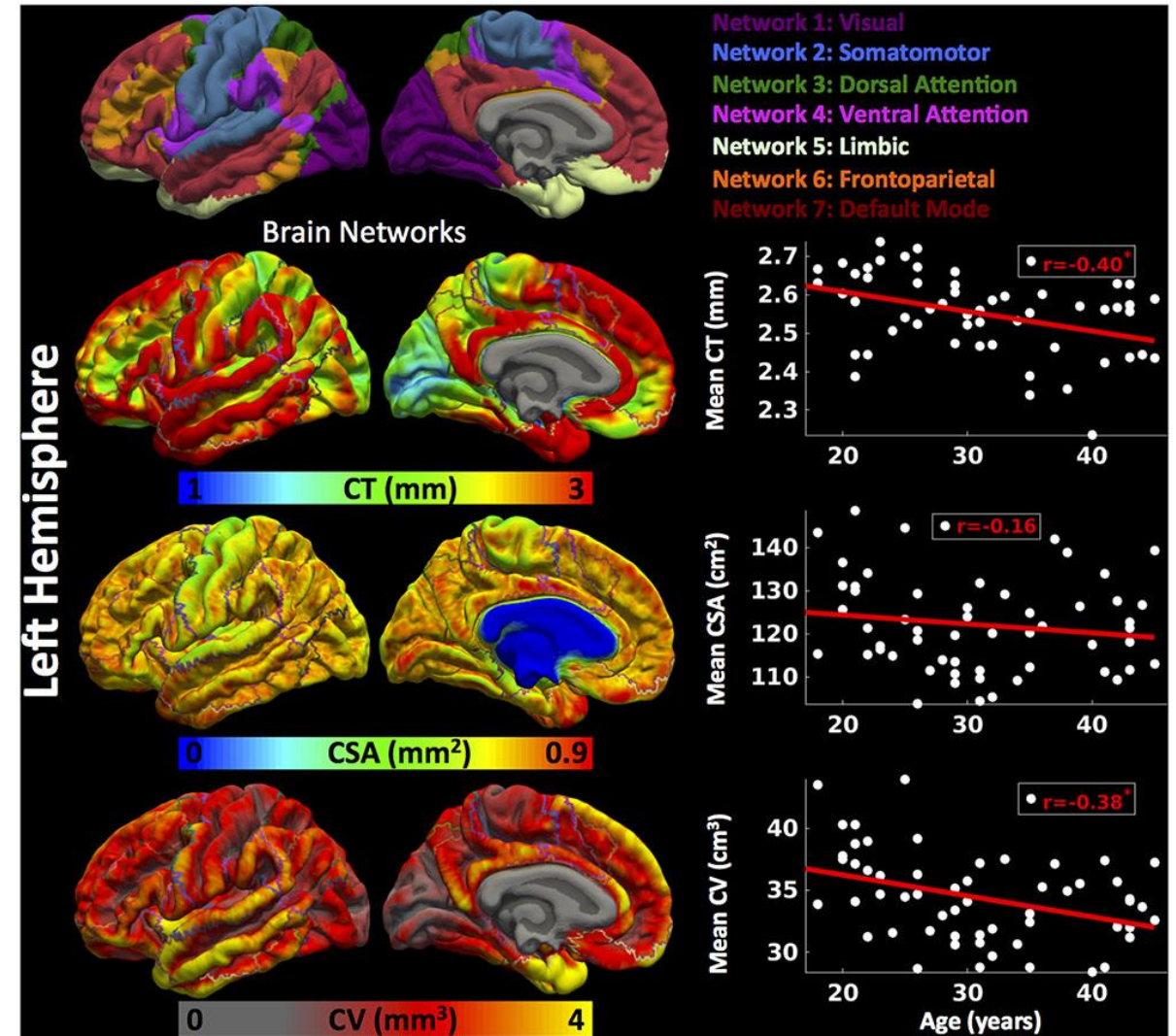
- Grey matter (GM)**
- Massive pruning of early on
- White matter (WM)**
- Peak in early middle age



Subcortical changes

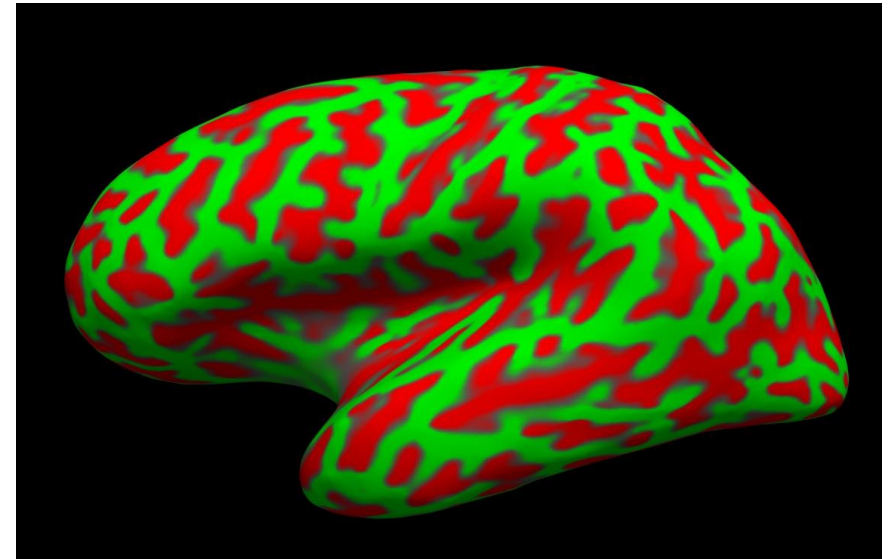


Cortical changes

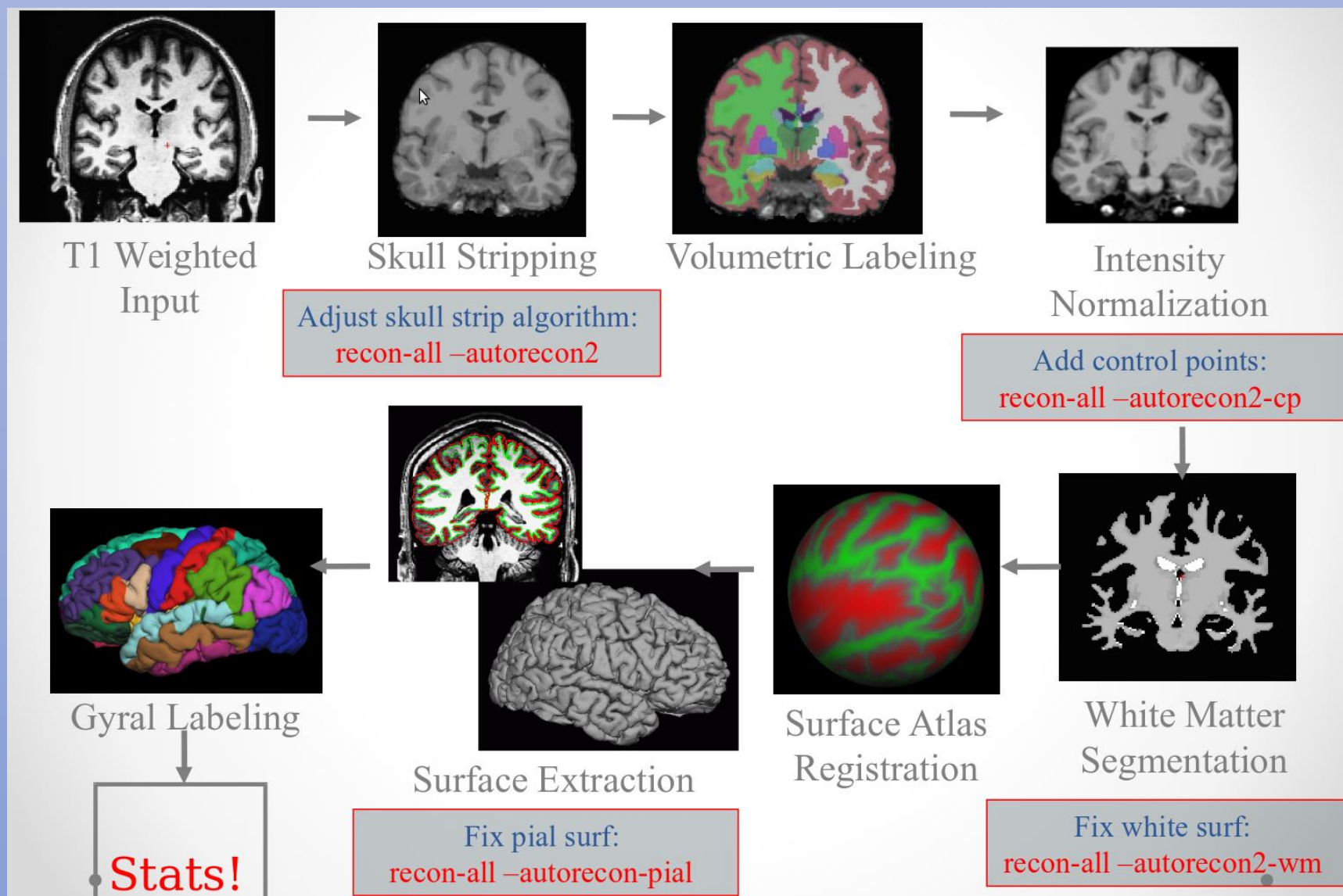


FreeSurfer

- A package for automatically calculating the anatomical parameters of the brain
- Developed at MGH Martinos Center (Boston) over the past 20-25 years
- Main functions
 - Sub-cortical volume
 - Surface-based analysis of the cortex
 - Cortical thickness
 - Cortical surface area
 - Cortical volume
 - Mean curvature (also known as gyrification ratio)



FreeSurfer processing pipeline



Percentage change per decade (both gender)

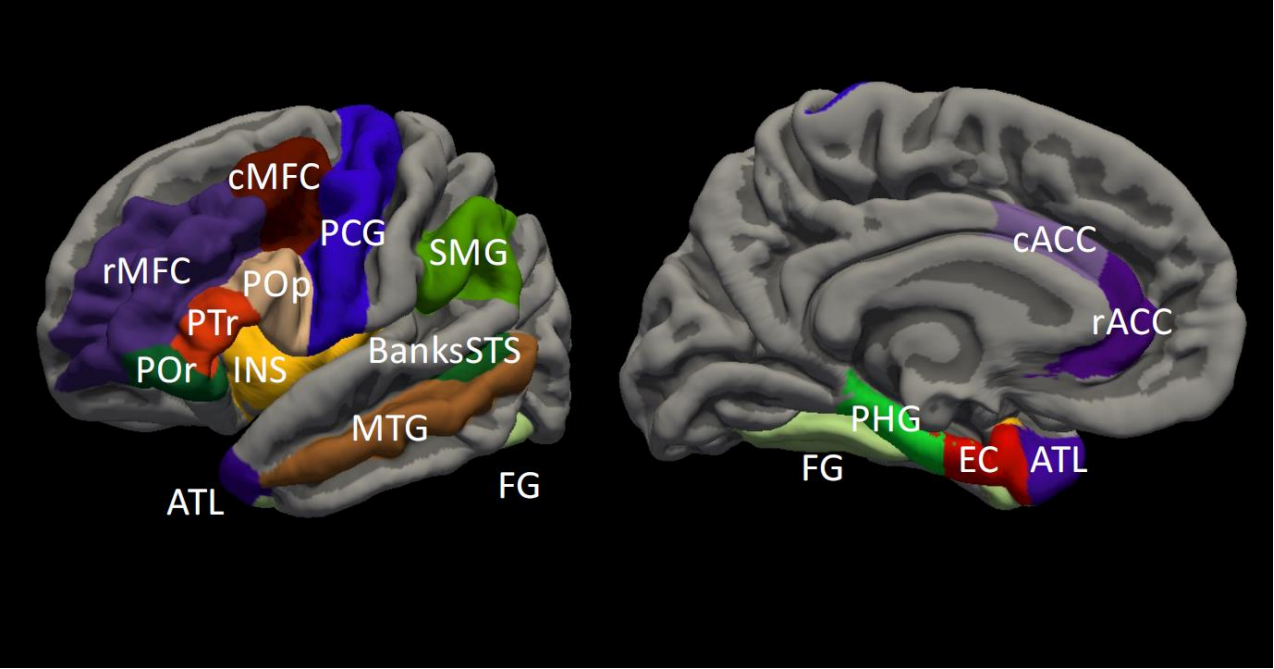
Percentage change per decade for the total sample based on raw volumes.

	18–29 to 30–39	30–39 to 40–49	40–49 to 50–59	50–59 to 60–69	60–69 to 70–79	70–79 to 80–95	18–29 to 80–95
Cerebral Cor	–5.5	–.8	–7.9	–6.2	–6.1	–1.0	–24.7
Cerebral WM	3.9	1.7	–4.6	–5.2	–7.9	–8.5	–19.7
Lat Vent	18.9	7.4	8.2	40.6	39.2	20.8	226.5
Inf Lat Vent	11.2	–2.6	1.0	46.8	66.7	43.5	283.9
Cerebel WM	.8	–.5	–3.7	–5.7	–5.2	–6.5	–19.3
Cerebel Cor	–.9	–1.1	–6.1	–4.0	–7.0	.3	–17.6
Thalamus	.2	–2.9	–6.4	–4.0	–6.0	–5.0	–21.9
Caudate	–6.7	–2.5	–2.8	–1.2	1.8	4.4	–7.2
Putamen	–9.6	–5.3	–4.7	–5.5	–1.2	–5.1	–27.8
Pallidum	–6.7	–4.7	–5.7	–2.3	–3.1	–8.4	–27.3
Hippocampus	1.3	.6	–3.2	–7.8	–8.1	–9.7	–24.5
Amygdala	–2.8	–1.2	–5.4	–5.9	–8.6	–3.1	–24.3
Accumbens	–15.3	–7.0	–2.8	–7.2	–4.4	2.5	–30.4
3rd Vent	–3.9	4.1	13.6	21.0	27.8	6.3	86.7
4th Vent	–1.7	–7.4	2.6	9.1	5.5	–2.3	5.0
Brainstem	3.4	.1	–3.4	–.8	–5.2	–5.4	–11.0
CSF	3.8	2.6	–2.2	13.4	9.2	4.2	34.4
Total volume	–1.1	.1	–6.1	–5.4	–6.8	–4.3	–21.6

Cor: cortex; WM: white matter; Lat: lateral; Inf: inferior; Vent: ventricles; CSF: cerebrospinal fluid; Total volume: the sum of all the other structures (CSF and ventricles not included).

Foreign language learning as *cognitive training*

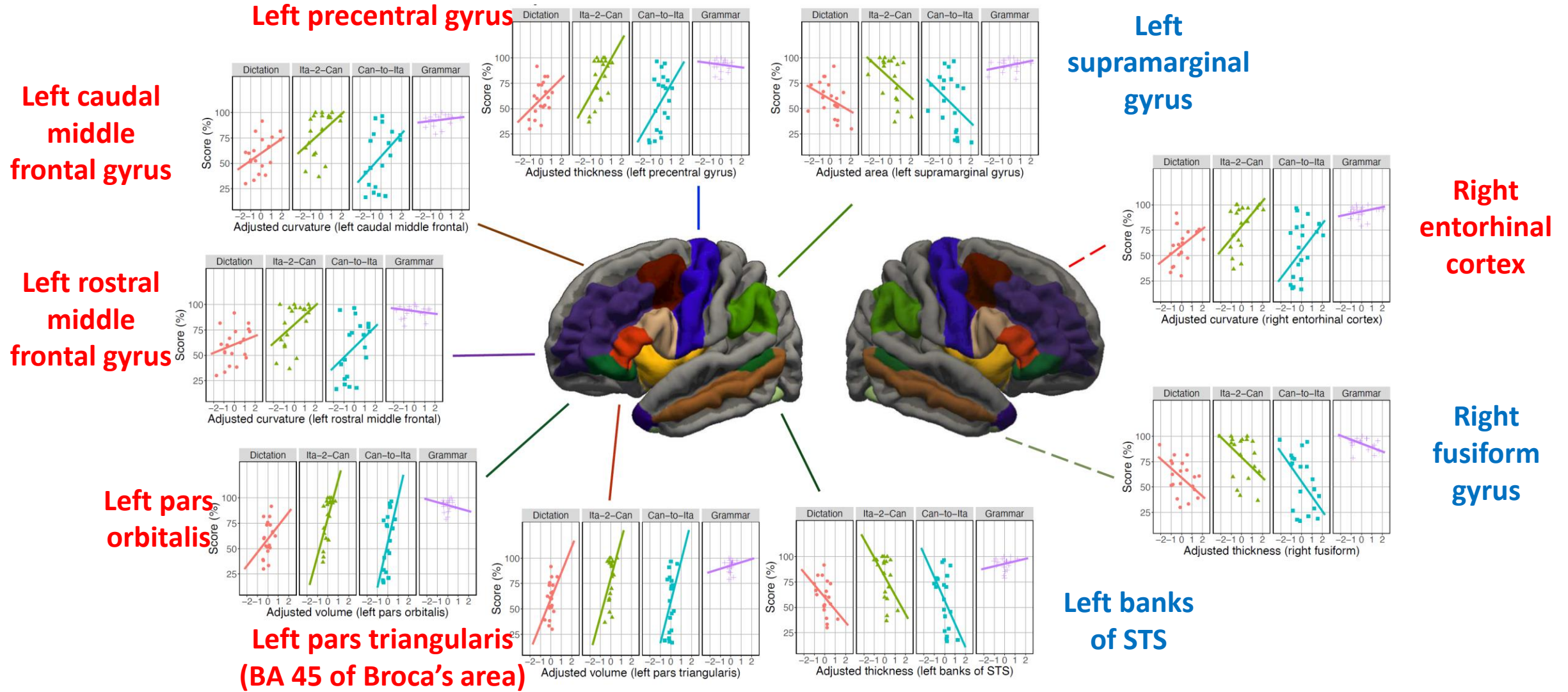
- Sep 2020 to May 2021
- Research participants
 - 25 older participants aged 58-69
- Italian vocabulary learning sessions
- Longitudinally follow their foreign language (Italian) learning performance for about 5 weeks



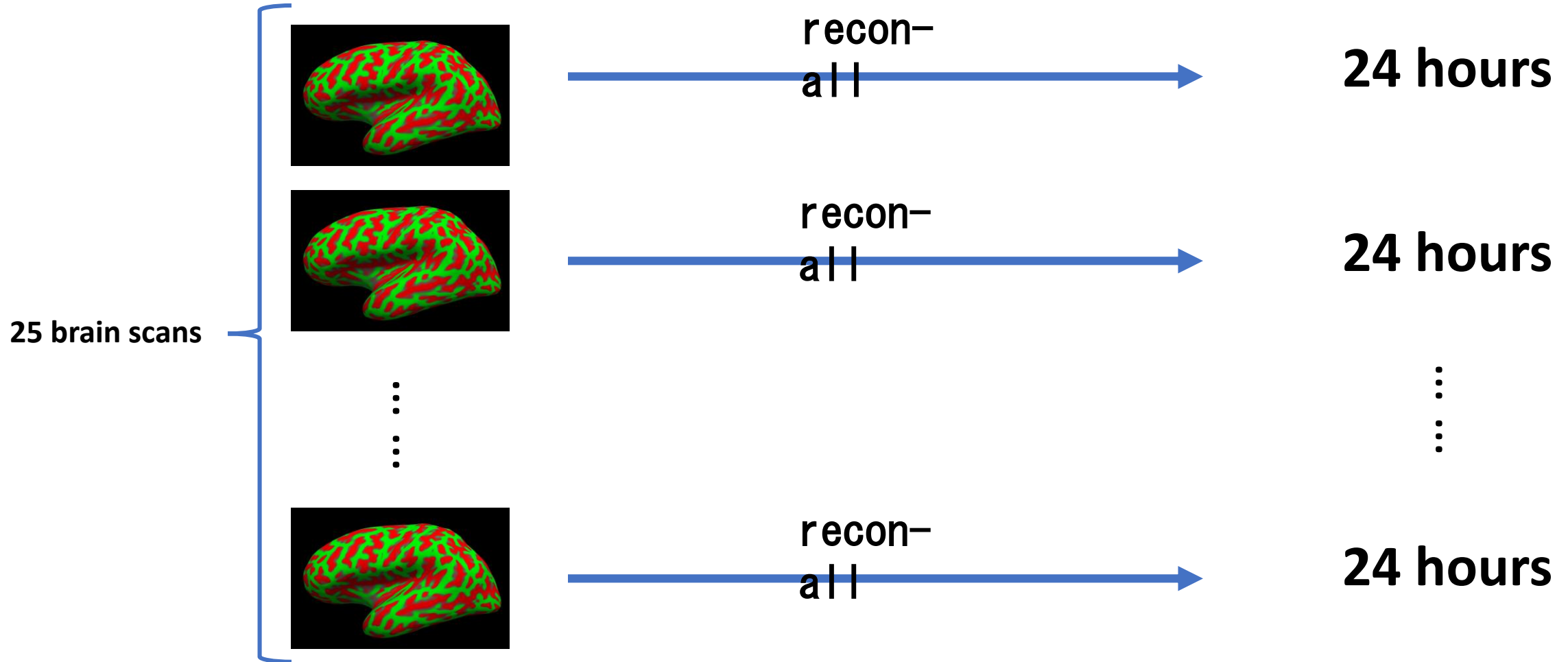
Cortical ROIs

Rostral middle frontal cortex	rMFC	Supramarginal gyrus	SMG
Caudal middle frontal cortex	cMFC	Middle temporal gyrus	MTG
Pars opercularis	POp	Rostral anterior cingulate	rACC
Pars triangularis	PTr	Caudal anterior cingulate	cACC
Pars orbitalis	POr	Fusiform gyrus	FG
Anterior temporal lobe	ATL	Parahippocampal gyrus	PH
Insula	INS	Entorhinal cortex	EC
Precentral gyrus	PCG	Banks of the Superior Temporal Sulcus	BankSTS

Language retention performance post-learning



UBDA application example: *brain segmentation*



UBDA example: *brain segmentation using FreeSurfer*

- Command-line analysis pipeline (**recon-all**)

- **recon-all** -all -i ./T1/f.nii -s L101 -brainstem-structures -
parallel -openmp 6

options of the command

options related to parallel processing
(here, we used 6 processors or threads)

- With parallel computing, significant speedup to **~ 6 hours per brain scan**

UBDA example: *brain segmentation using FreeSurfer*

PBS setting

Use q2s01

Nodes=1:ppn=16

- Request 1 node
- 16 processors per node

Setup environment

- Setup Home directory (~/.freesurfer)
- Setup Subject directory (~/.SLVLsubjects)

Timekeeping & log

- Display the end time
- Remove some log files

```

GNU nano 2.3.1 File: runSMRI_16.sh
#!/bin/sh
cd ~/SLVL/
subid=$1

cd ~/SLVL/$subid
recon-all -all -i ./T1/f.nii -s $subid -brainstem-structures -parallel -openmp 16

GNU nano 2.3.1
#!/bin/sh
#PBS -N slvlanalysis
#PBS -l nodes=1:ppn=16
#PBS -l walltime=100:00:00
#PBS -q q2s01
#PBS -V
#PBS -S /bin/bash
#module load freesurfer-7.2.0
#module load python-2.7.15
module load openmpi-3.0.1-gcc-5.5.0
#####
export FREESURFER_HOME=~/.freesurfer
export SUBJECTS_DIR=~/.SLVLsubjects
source $FREESURFER_HOME/SetUpFreeSurfer.sh

~/SLVL/runSMRI_16.sh $var > ~/SLVL/log-$var.out

echo "+++++"
echo "process end at : "
date
rm -f /tmp/nodefile.$$
rm -f /tmp/nodes.$$
module unload openmpi-3.0.1-gcc-5.5.0
    
```

Shell script:
runSMRI_16.sh **1**

PBS script
runSMRI_cache_q2.pbs **2**

Call the shell script

3 Submit the job
qsub -v "var=L101"
runSMRI_cache_q2.pbs

Brief summary and tips (for absolute beginners)

• Procedure

- In the home directory (cd ~), install the required software or your own scripts
- **Write the shell script**
- **PBS (Portable Batch System) script (.pbs)**
 - **Setup number of nodes**
 - #PBS -l nodes=1:ppn=16
 - **Select the queue (q2s01, q4s01, qgpu01, qmic01)**
 - **Setup environmental variables**
 - **Call the shellscript (.sh)**
- **Submit the job to the queue**

• Useful software

- **Terminal software: MobaXterm (Windows) or Hyper (Mac)**
- **FileZilla:** file transfer between local computer and server

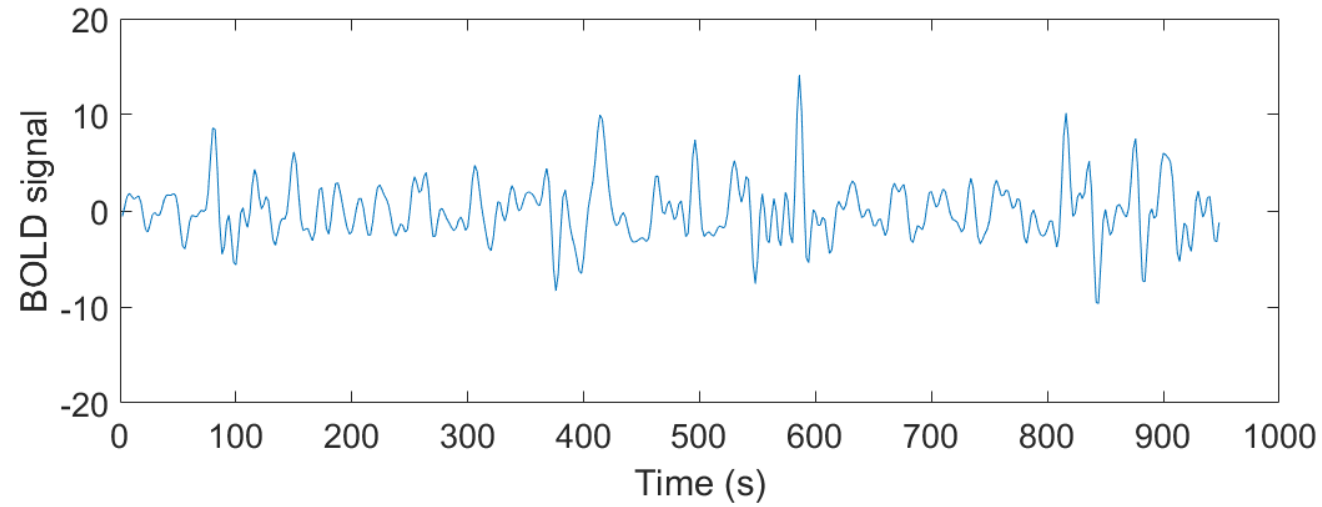
• Most useful commands

- `history | grep "qsub"`
 - to keep track of the submitted jobs
- `qstat`
 - to see what jobs are currently running

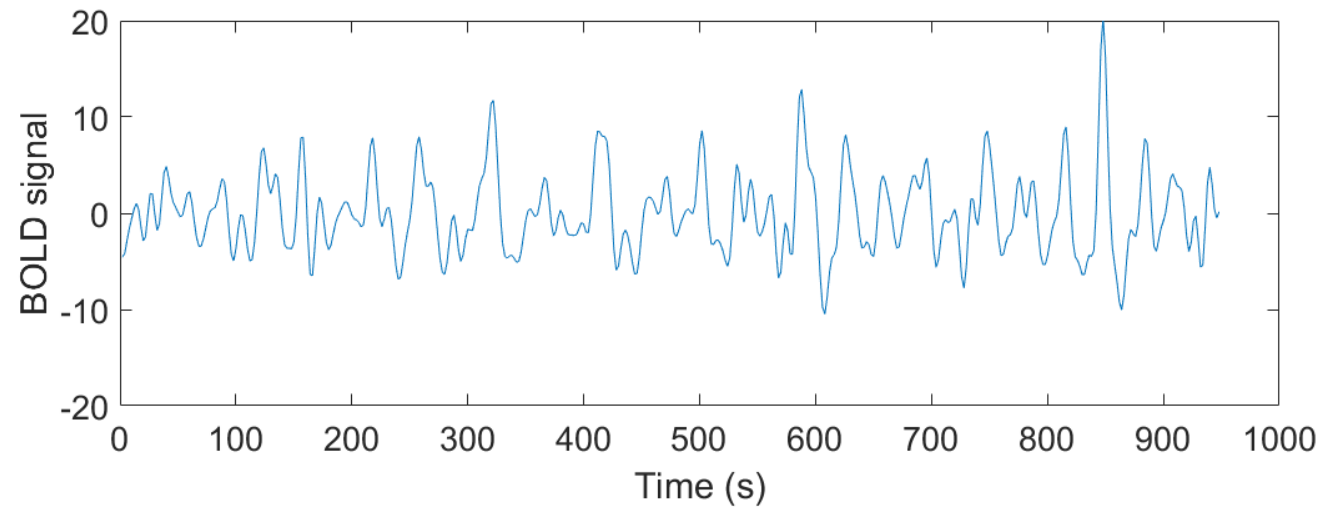
Functional connectivity

Temporal correlations between remote neurophysiological events (Friston, K. J.)

ROI1: Left precentral gyrus
(primary motor cortex,
controls right hand)



ROI2: Right precentral gyrus
(primary motor cortex,
controls left hand)



Resting-state functional connectivity

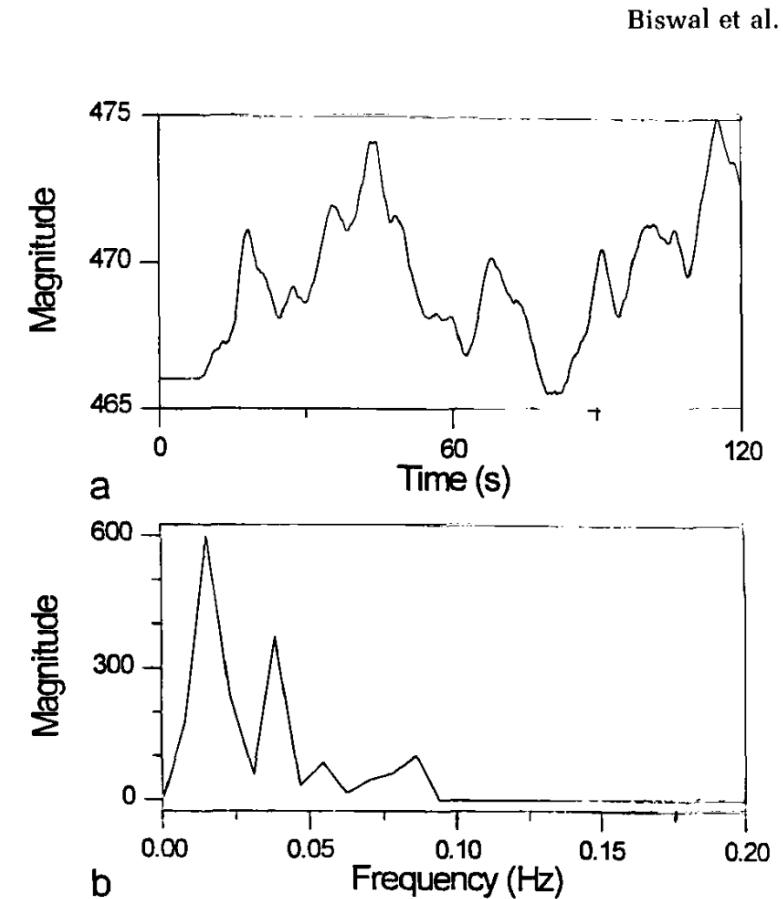
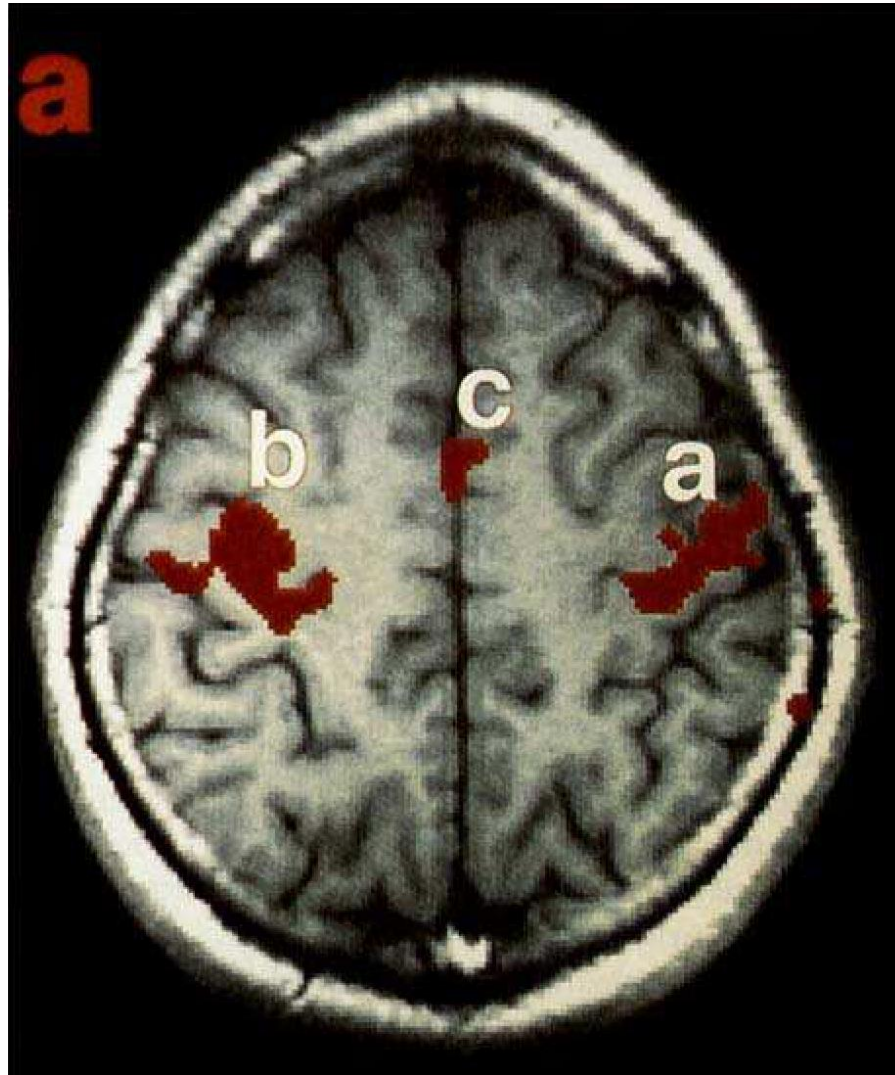
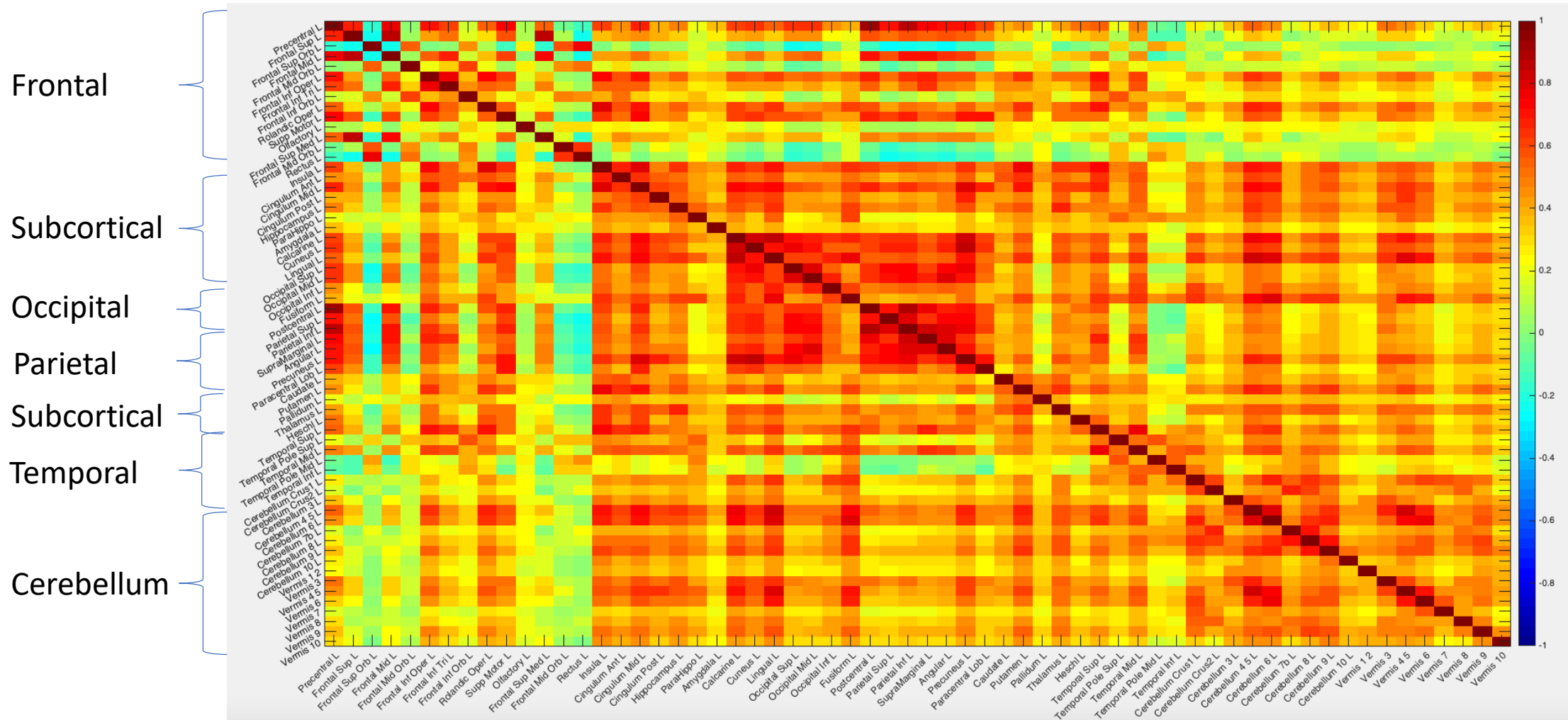
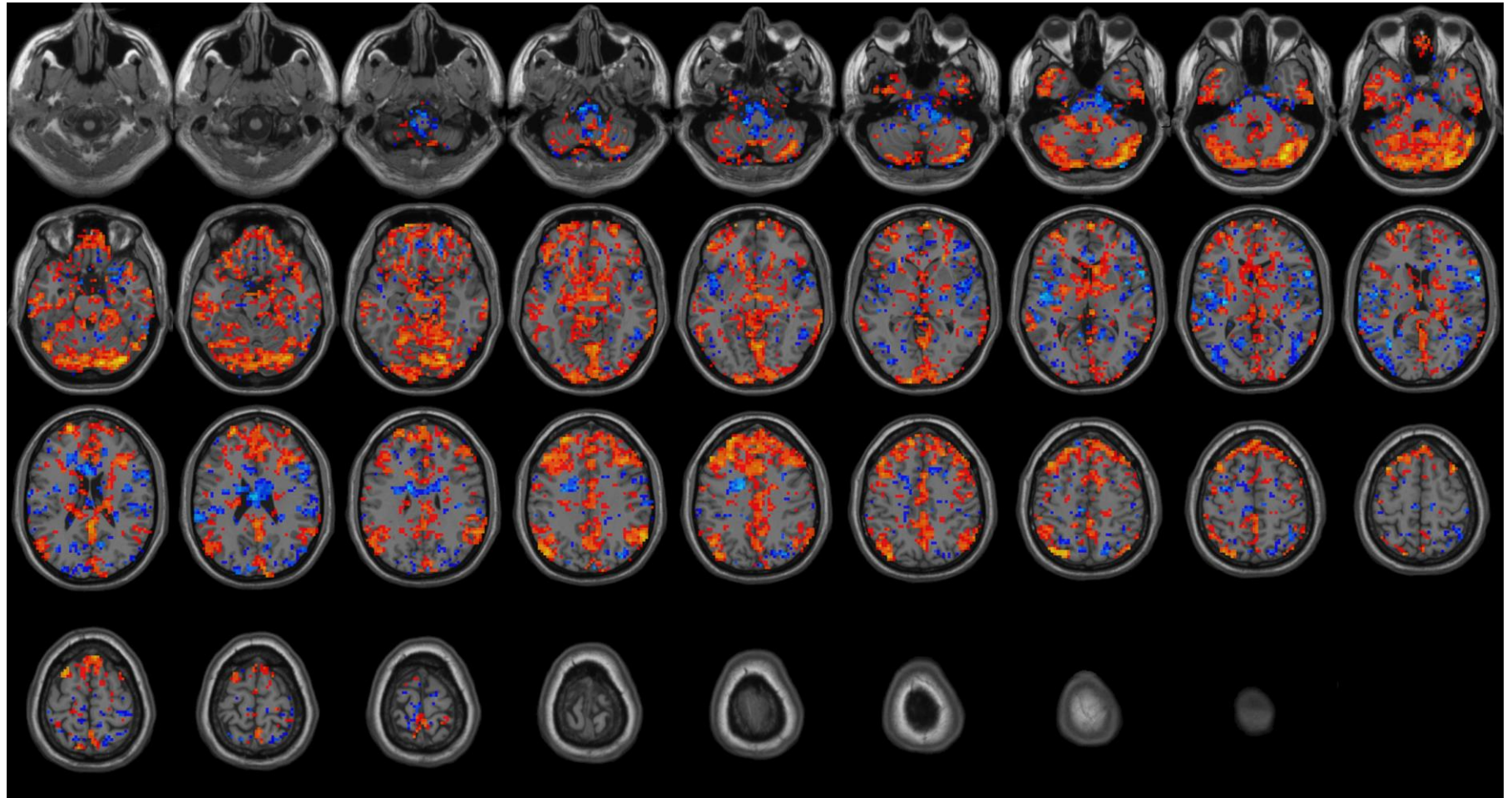
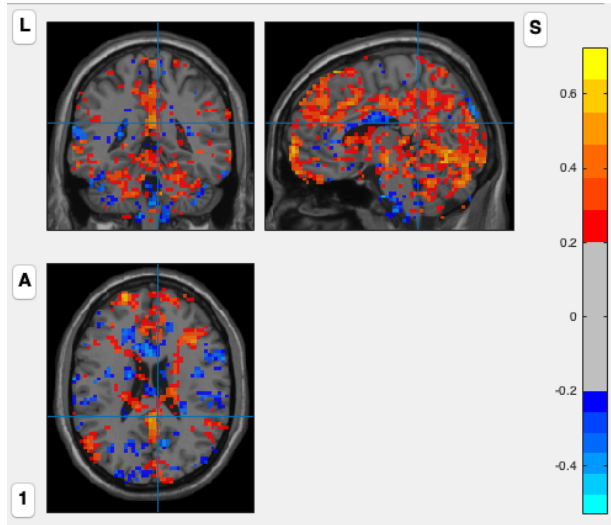


FIG. 1. (a) Representative time course of motor cortex physiological fluctuations after FIR filtering that passes frequencies <0.08 Hz. (b) Fourier transform of the time course.

Functional connectivity matrix



Example of functional connectivity pattern



UBDA application example: *brain segmentation*

PBS setting

- Use q2s01
- Nodes=1:ppn=16
 - Request 1 node
 - 16 processors per node

Setup environmental variables

- Load FreeSurfer
- Setup Home directory (~ / freesurfer)
- Setup Subject directory (~ / SLVLsubjects)
- Source to set path

Call the shell script

Timekeeping & log

- Display the end time
- Remove some log files

```

GNU nano 2.3.1 File: runRSFMRI q2.pbs
#!/bin/sh
#PBS -N slvlnalysis
#PBS -l nodes=1:ppn=16
#PBS -l walltime=100:00:00
#PBS -q q2s01
#PBS -V
#PBS -S /bin/bash
#module load freesurfer-7.2.0
#module load python-2.7.15
module load openmpi-3.0.1-gcc-5.5.0
#####

export FS_MCRR00T=$FREESURFER_HOME/MCrv84
export LD_LIBRARY_PATH="$LD_LIBRARY_PATH": "$FS_MCRR00T/runtime/glnxa64"
export LD_LIBRARY_PATH="$LD_LIBRARY_PATH": "$FS_MCRR00T/bin/glnxa64"
export LD_LIBRARY_PATH="$LD_LIBRARY_PATH": "$FS_MCRR00T/sys/os/glnxa64"
export LD_LIBRARY_PATH="/usr/lib/x86_64-linux-gnu": "/lib/x86_64-linux-gnu": "$LD_LIBRARY_PATH"
export FS_SPMREG_USE_BIN=1
export FS_MKCON_USE_BIN=1
export FS_SXA3_USE_BIN=1

export FREESURFER_HOME=~ / freesurfer
export SUBJECTS_DIR=~ / SLVLsubjects

source $FREESURFER_HOME / SetUpFreeSurfer.sh

~/SLVL/runRSFMRI.sh $var > ~/SLVL/log-func-$var.out

echo "+++++"
echo "process end at : "
date
rm -f /tmp/nodefile.$$
rm -f /tmp/nodes.$$
module unload openmpi-3.0.1-gcc-5.5.0

```

RunRSFMRI.sh

```
GNU nano 2.3.1 File: runRSFMRI.sh
#!/bin/sh
cd ~/SLVL/
subid=$1

# preproc-sess -s $subid -fsd Rest -fwhm 5 -surface fsaverage lhrh -mni305-2mm -per-run

fcseed-config -segid 1010 -fcname L_Posteriorcingulate.dat -fsd Rest -mean -cfg L_Posteriorcingulate.config -overwrite
fcseed-config -segid 2010 -fcname R_Posteriorcingulate.dat -fsd Rest -mean -cfg R_Posteriorcingulate.config -overwrite

fcseed-sess -s $subid -cfg L_Posteriorcingulate.config
fcseed-sess -s $subid -cfg R_Posteriorcingulate.config

# fcseed-config -wm -fcname wm.dat -fsd Rest -pca -cfg wm.config
# fcseed-sess -s $subid -cfg wm.config

# fcseed-config -vcsf -fcname vcsf.dat -fsd Rest -pca -cfg vcsf.config
# fcseed-sess -s $subid -cfg vcsf.config

mkanalysis-sess -analysis fc.lpccseed.surf.lh -surface fsaverage lh -fwhm 5 -notask -taskreg L_Posteriorcingulate.dat 1 -nuisreg vcsf.dat 5 \
-nuisreg wm.dat 5 -mcextreg -polyfit 5 -hpf 0 -nskip 4 -fsd Rest -TR 2 -per-run

mkanalysis-sess -analysis fc.lpccseed.surf.rh -surface fsaverage rh -fwhm 5 -notask -taskreg L_Posteriorcingulate.dat 1 -nuisreg vcsf.dat 5 \
-nuisreg wm.dat 5 -mcextreg -polyfit 5 -hpf 0 -nskip 4 -fsd Rest -TR 2 -per-run

mkanalysis-sess -analysis fc.rpccseed.surf.lh -surface fsaverage lh -fwhm 5 -notask -taskreg R_Posteriorcingulate.dat 1 -nuisreg vcsf.dat 5 \
-nuisreg wm.dat 5 -mcextreg -polyfit 5 -hpf 0 -nskip 4 -fsd Rest -TR 2 -per-run

mkanalysis-sess -analysis fc.rpccseed.surf.rh -surface fsaverage rh -fwhm 5 -notask -taskreg R_Posteriorcingulate.dat 1 -nuisreg vcsf.dat 5 \
-nuisreg wm.dat 5 -mcextreg -polyfit 5 -hpf 0 -nskip 4 -fsd Rest -TR 2 -per-run

selxavg3-sess -s $subid -a fc.lpccseed.surf.lh
selxavg3-sess -s $subid -a fc.lpccseed.surf.rh
selxavg3-sess -s $subid -a fc.rpccseed.surf.lh
selxavg3-sess -s $subid -a fc.rpccseed.surf.rh
```

RISA project: Brain age estimation

1 Construct brain age models using large public datasets

<http://adni.loni.usc.edu/>



<https://www.ukbiobank.ac.uk/>



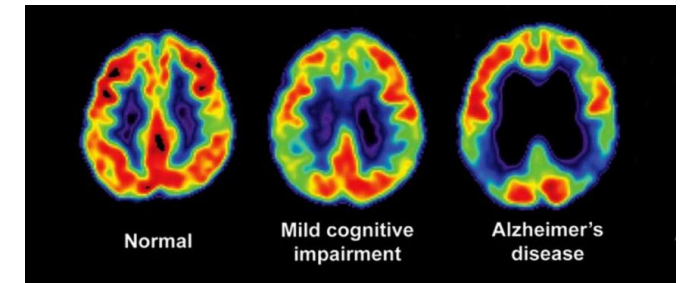
<http://www.humanconnectomeproject.org/>



2 Collect our own data to construct localized brain age models, & test their predictive powers for brain functions & prospective brain changes



3 Apply the brain age models in two clinical populations

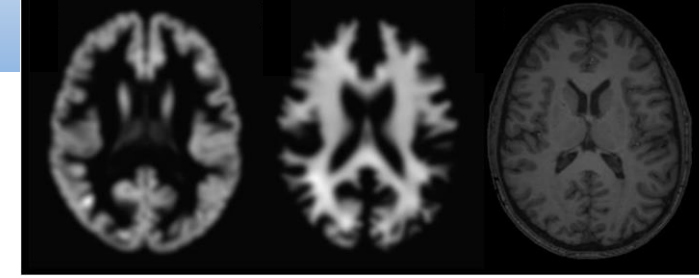


Mild cognitive impairment



Insomnia disorder

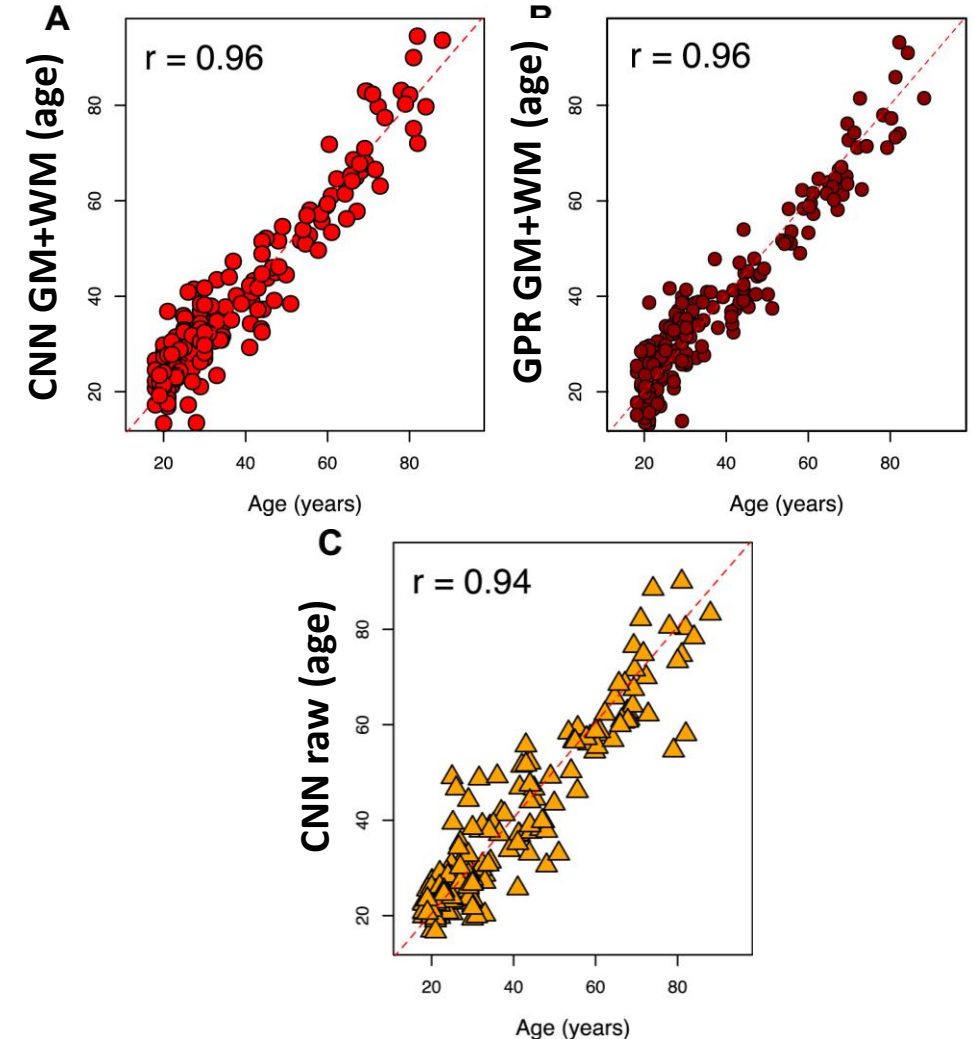
Brain age (structural neuroimaging)



**Convolutional
neural network**

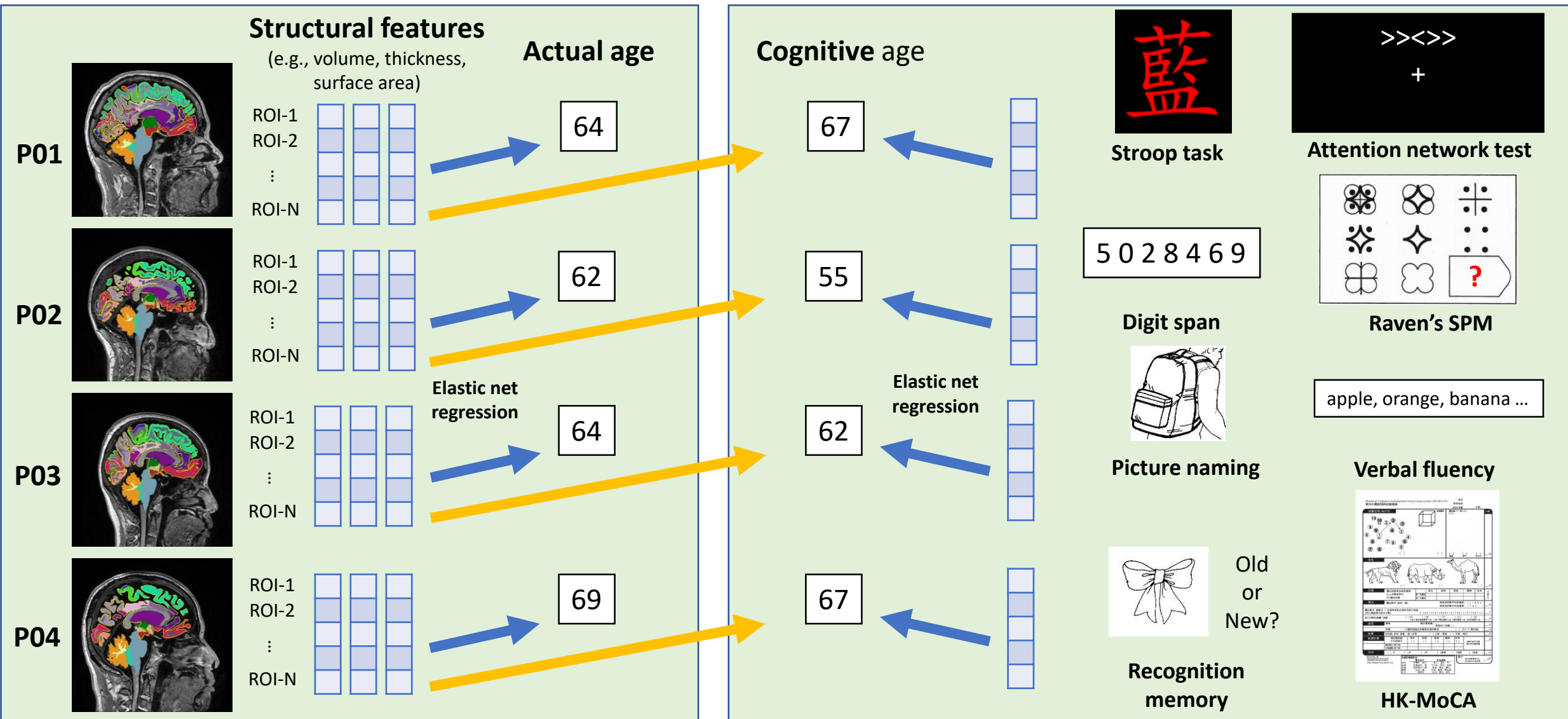
Method	Input data	MAE (years)	r	R ²	RMSE
CNN	GM	4.16	0.96	0.92	5.31
	WM	5.14	0.94	0.88	6.54
	GM+WM	4.34	0.96	0.91	5.67
	Raw	4.65	0.94	0.88	6.46
GPR	GM	4.66	0.95	0.89	6.01
	WM	5.88	0.92	0.84	7.25
	GM+WM	4.41	0.96	0.91	5.43
	Raw	11.81	0.57	0.32	15.10

**Gaussian Process
Regression**



MAE = mean absolute error, r = Pearson's r from correlation between chronological age and brain-predicted age, RMSE = root mean squared error, GM = Grey Matter, WM = White Matter.

Aim 1: predicting cognitive age with brain measures

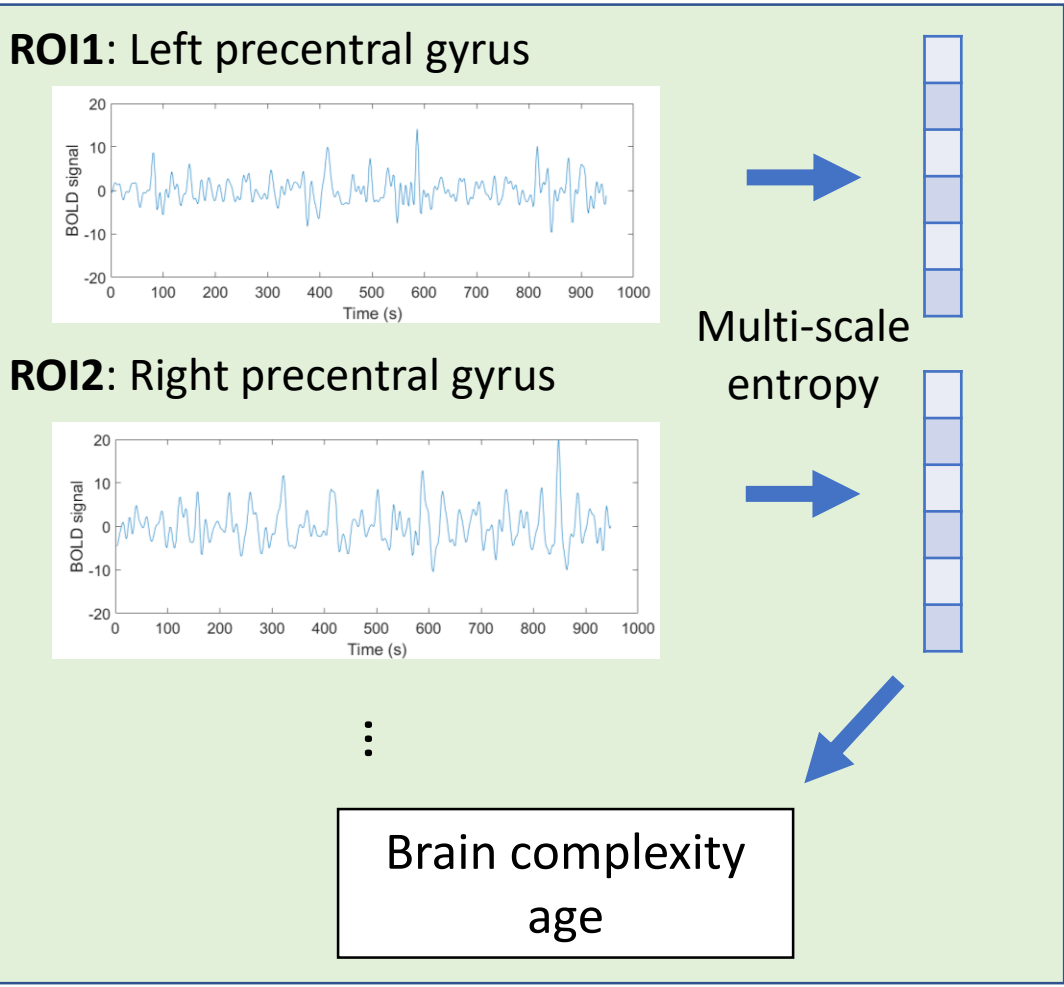


Fong, Hui, Fung, Ma, Law, Wang, & Wang (2020). *Quarterly Journal of Experimental Psychology*.

Fong, Law, Ma, Hui, & Wang (2021). *Brain and Language*.

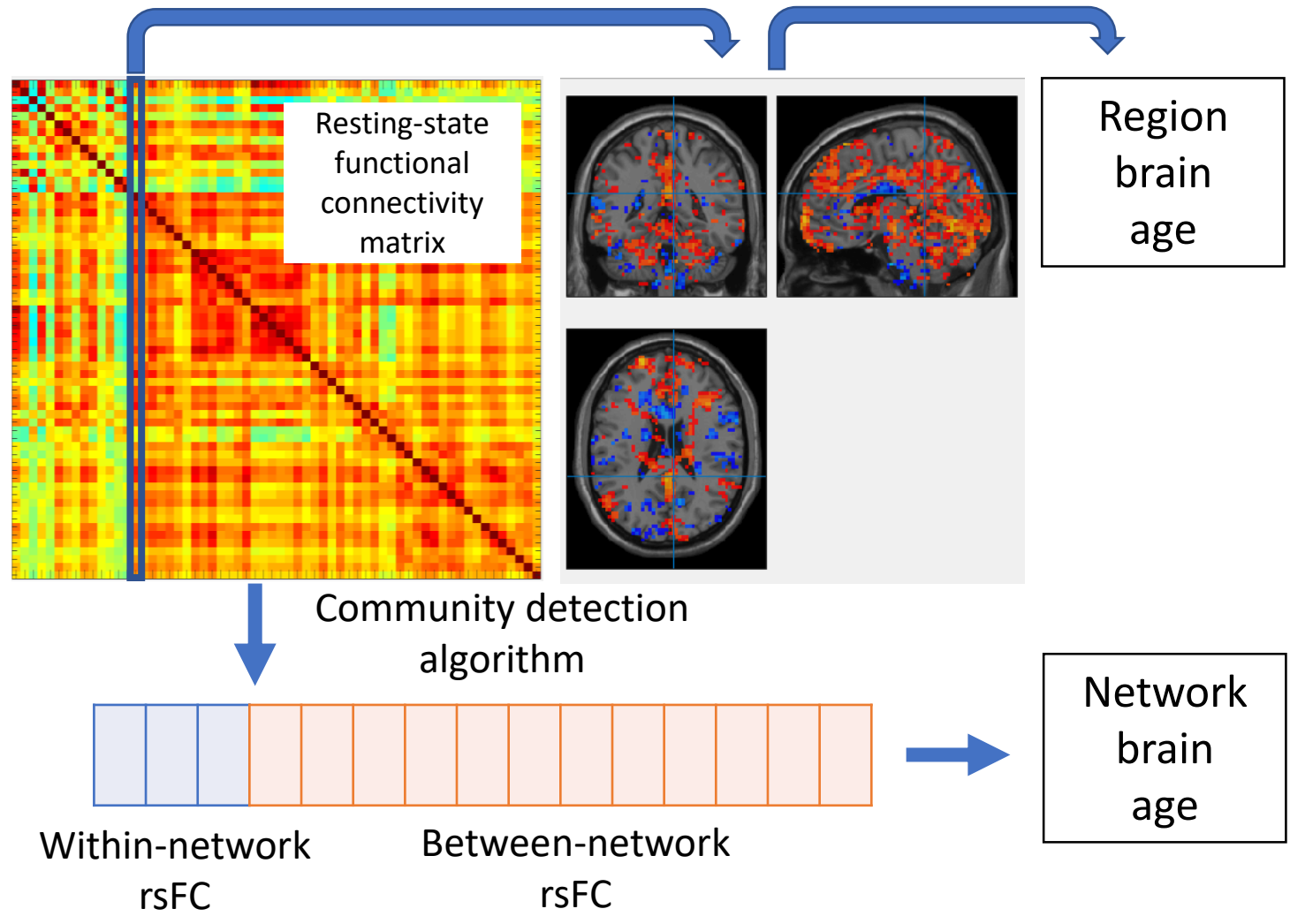
Aim 2:

brain complexity age (temporal)



Aim 3:

region & network brain age (spatial)



Ongoing experiments



Functional Connectome of Semantic Processing



招募 18-80 歲人士參與 大腦認知及語言功能研究

想認識神秘的大腦結構嗎？ 想知道你的大腦有否衰退嗎？



招募對象：

- 18歲至80歲 香港永久居民，一直以來大部分時間於香港居住，以廣東話為母語，沒有腦神經及心理疾病。



實驗內容：

1

認知能力測試 (分兩次進行)

- 認知能力測試 (1.5-2 小時)
- 語言功能測試及簡單問卷調查(1.5 小時)

2

EEG (腦電圖)



(2 小時)

磁力共振和腦電波皆為安全、對人體無痛、無害、無輻射的實驗方法。

3

MRI (磁力共振)



(預留 1.5 小時，實際掃描時間約30-45 分鐘)
[可選，隨機]



實驗詳情：

- 實驗日期：12 月至 4 月 (每人 3-4 次)
- 實驗將會在香港理工大學內舉行
- 完成後您會獲得 300-400 元 港幣的車馬費，以及認知語言能力報告 (MRI 參與者更可獲知自己大腦的基本資料)



報名方式：

有興趣參與的人士請掃 QR code，填妥表格。我們會盡快與您聯絡，安排時間。

查詢：馮小姐 (yunr.feng@connect.polyu.hk)
方博士 (cmmfong@polyu.edu.hk)



Acknowledgment

- HKRGC-GRF 15601718 & 15606119 awarded to Prof. William SY Wang (CBS)
- Research Institute for Smart Aging (RISA)
- University facility for Big Data Analytics (UBDA)
- University facility for Behavioral and Systems Neuroscience (UBSN)
- Prof. William SY Wang & other team members

References

- Bajaj, S., Alkozei, A., Dailey, N. S., & Killgore, W. D. (2017). Brain aging: uncovering cortical characteristics of healthy aging in young adults. *Frontiers in Aging Neuroscience, 9*, 412.
- Biswal, B., Zerrin Yetkin, F., Haughton, V. M., & Hyde, J. S. (1995). Functional connectivity in the motor cortex of resting human brain using echo-planar MRI. *Magnetic Resonance in Medicine, 34*(4), 537-541.
- Cole, J. H., Poudel, R. P., Tsagkrasoulis, D., Caan, M. W., Steves, C., Spector, T. D., & Montana, G. (2017). Predicting brain age with deep learning from raw imaging data results in a reliable and heritable biomarker. *NeuroImage, 163*, 115-124.
- Coupé, P., Catheline, G., Lanuza, E., Manjón, J. V., & Alzheimer's Disease Neuroimaging Initiative. (2017). Towards a unified analysis of brain maturation and aging across the entire lifespan: A MRI analysis. *Human Brain Mapping, 38*(11), 5501-5518.
- Fong, M. C-M., ..., Wang, W. S. (2020). Which cognitive functions subserve clustering and switching in category fluency? Generalisations from an extended set of semantic categories using linear mixed-effects modelling. *Quarterly Journal of Experimental Psychology, 73*(12), 2132-2147.

References

- Fong, M. C. M., Law, T. S. T., Ma, M. K. H., Hui, N. Y., & Wang, W. S. (2021). Can inhibition deficit hypothesis account for age-related differences in semantic fluency? Converging evidence from Stroop color and word test and an ERP flanker task. *Brain and Language*, 218, 104952.
- Fong, Ma, Chui, Law, Hui, Au, & Wang. (Under review). Foreign language learning in older adults: anatomical and cognitive markers of vocabulary learning success.
- Fjell, A. M., McEvoy, L., Holland, D., Dale, A. M., Walhovd, K. B., & Alzheimer's Disease Neuroimaging Initiative. (2014). What is normal in normal aging? Effects of aging, amyloid and Alzheimer's disease on the cerebral cortex and the hippocampus. *Progress in Neurobiology*, 117, 20-40.
- Ma, M. K. H., Fong, M. C. M., Xie, C., Lee, T., Chen, G., & Wang, W. S. (2021). Regularity and randomness in ageing: Differences in resting-state EEG complexity measured by largest Lyapunov exponent. *Neuroimage: Reports*, 1(4), 100054.
- Walhovd, K. B., Wfoestlye, L. T., Amlien, I., Espeseth, T., Reinvang, I., Raz, N., ... & Dale, A. M. (2011). Consistent neuroanatomical age-related volume differences across multiple samples. *Neurobiology of Aging*, 32(5), 916-932.



Happy Winter Solstice!
&
Merry Christmas!
