

Computational Neuroscience Laboratory

Machine Learning for Neuroimaging

Autumn 2023

10/18/2023

Hands on Statistics & ML (Python)

Intro into Jupyter Notebook

- Jupyter Notebook is an open-source web application
- Create and share documents that contain live code, equations, visualizations, and text
- The name, Jupyter, comes from the core supported programming languages: Julia, Python, and R





Getting Started with Jupyter

Install Jupyter using pip (or Anaconda)

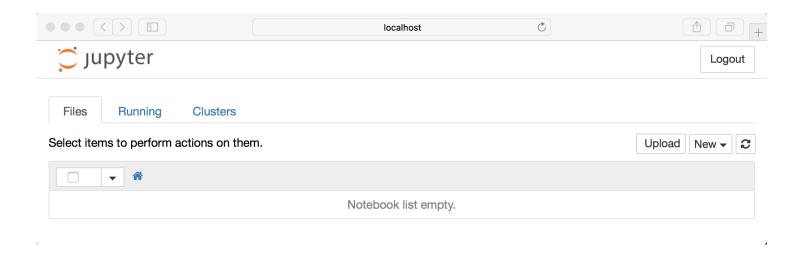
```
$ pip install jupyter
```

Run Jupyter

```
$ jupyter notebook
```

Create New Notebook

 Jupyter will run on your default browser (or open a new tab) to the following URL: http://localhost:8888/tree

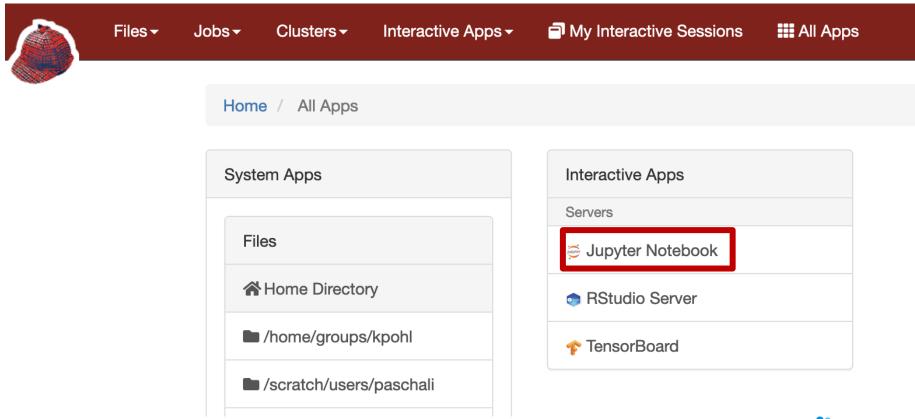


Click 'New' to create a new Jupyter Notebook



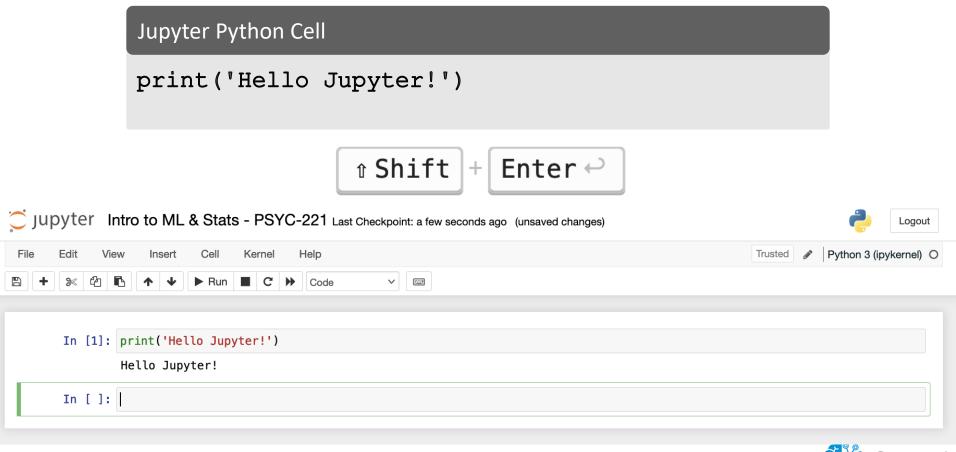
Jupyter on Sherlock

https://login.sherlock.stanford.edu/pun/sys/dashboard



Running your first cell

Add Python code to the cell and try running its contents

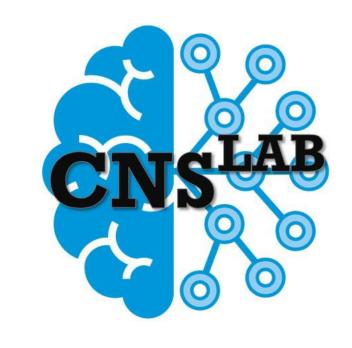


Requirements Install in a virtual environment or load modules in Sherlock

- numpy==1.24.4
- pandas==2.0.2
- scikit-image==0.22.0
- scikit-learn==1.2.2
- scipy==1.11.0
- seaborn==0.12.2
- statsmodels==0.14.0
- nibabel==5.1.0
- nilearn==0.10.1
- torch==2.0.1 (PyTorch)

- Executed (after installing <u>Anaconda</u>)
 - conda create --name tut3 python=3.11
 - conda activate tut3
 - pip install -r requirements.txt
 - pip install torch --index-url https://download.pytorch.org/whl/cu118 (taken from here)
- I also include a <u>.yml file!</u>
- Make sure to activate the environment (conda activate tut3) before calling jupyter notebook!





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Part 1: Data Preparation and Hypothesis Testing





Outline

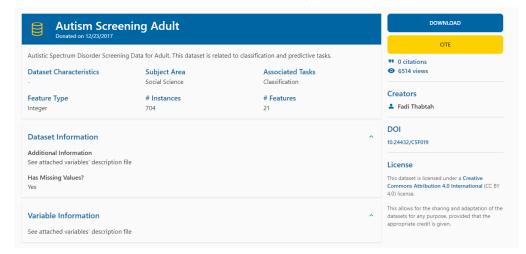
- 1. Load the data
- 2. Handle missing values and remove outliers
- 3. Test hypothesis on preconceptions we may have
- 4. Residualization: dealing with co-founder effects
- 5. Save processed data frame

Autism Screening on Adults

- Autism spectrum disorder (ASD) is a developmental disability
- Diagnosing ASD can be difficult because there is no test, e.g. blood test, to diagnose the disorder
- Many adults can and do have ASD. But it's only recently that doctors have begun diagnosing them with it

Autism Screening on Adults - Dataset

- Dataset contains survey results for more than 700 subjects
- Results from a quick referral guide for adults with suspected autism
- Download dataset from here



¹⁾ Thabtah, F. (2017). Autism Spectrum Disorder Screening: Machine Learning Adaptation and DSM-5 Fulfillment. Proceedings of the 1st International Conference on Medical and Health Informatics 2017, pp.1-6. Taichung City, Taiwan, ACM.

²⁾ Thabtah, F. (2017). ASDTests. A mobile app for ASD screening. www.asdtests.com [accessed December 20th, 2017].

³⁾ Thabtah, F. (2017). Machine Learning in Autistic Spectrum Disorder Behavioural Research: A Review. To Appear in Informatics for Health and Social Care Journal. December, 2017 (in press)

Data Loading

- Load csv file
- Visualize the top five rows (head)
- Notice data types, column values, empty cells

	A1_Score	A2_Score	A3_Score	A4_Score	A5_Score	A6_Score	A7_Score	A8_Score	A9_Score	A10_Score	 gender	ethnicity	jundice
0	1	1	1	1	0	0	1	1	0	0	 f	White- European	no
1	1	1	0	1	0	0	0	1	0	1	 m	Latino	no
2	1	1	0	1	1	0	1	1	1	1	 m	Latino	yes
3	1	1	0	1	0	0	1	1	0	1	 f	White- European	no
4	1	0	0	0	0	0	0	1	0	0	 f	?	no

Data Cleaning

- Remove row with outlier age
- Look for missing values
 - Empty cells
 - Cells with '?'
- Count amount of missing values
 - 2 missing age values
 - 95 missing ethnicities
 - 95 missing relations



Data Imputation

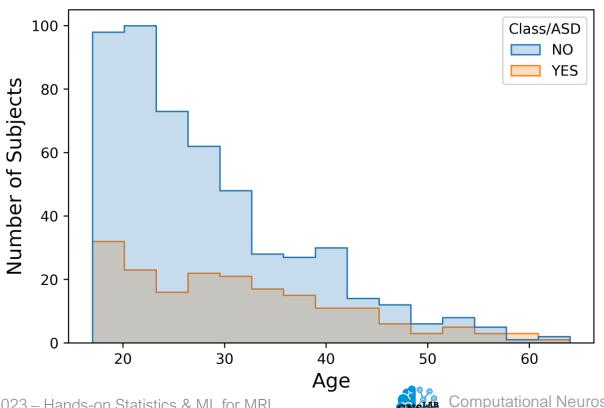
- Fill out missing age values with column mean
 - Important! This step comes after we have removed the outliers

- Replace missing ethnicities and relations with 'Other'
 - We do not know the actual values

Hypothesis Testing

- Dataset is split into two cohorts: controls and positive for autism (ASD)
- Null hypothesis: the age distribution of the two cohorts is equal

- Histograms of ages of two cohorts
- Hard to visually compare



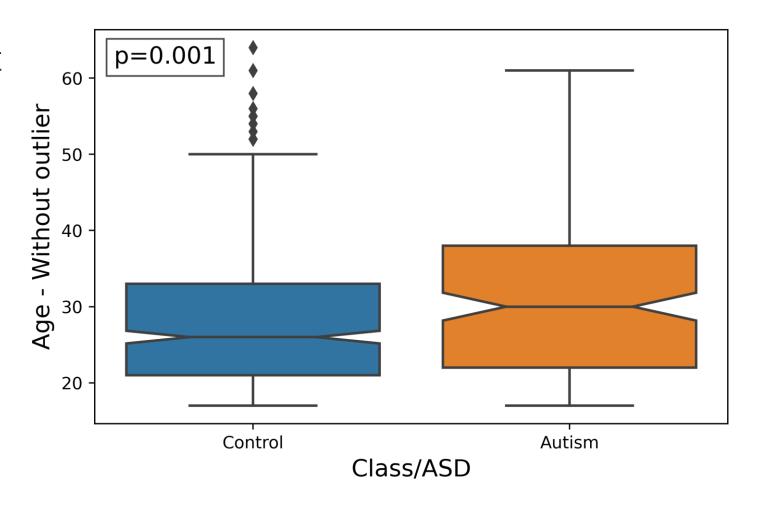
Compare age of cohorts

- Mann-Whitney U-Test
- Tests whether the distributions of two independent samples are equal
- Nonparametric test
- $p < 0.05 \rightarrow We reject the null hypothesis$

```
Jupyter Python Cell
U1, p = mannwhitneyu(controls["age"], autism["age"])
```

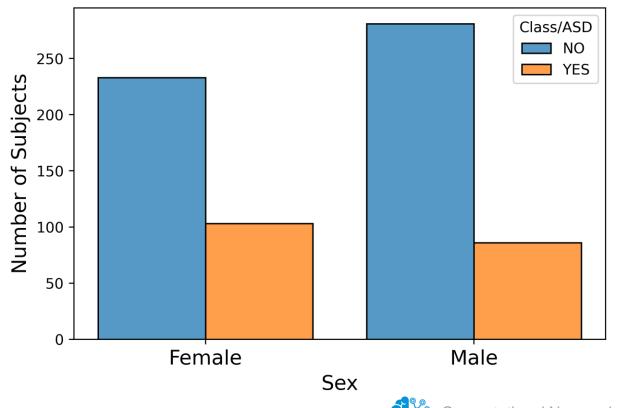
Compare age of cohorts

- The notches in the box plot do not overlap
- We can conclude, with95% confidence that thetrue medians differ



Hypothesis Testing

- Dataset is split into two cohorts: controls and positive for autism (ASD)
- Null hypothesis: the frequency of autism is the same across sexes
- Histograms of subjects of the two cohorts
- Is there a difference between Females and Males?



Compare autism across sexes

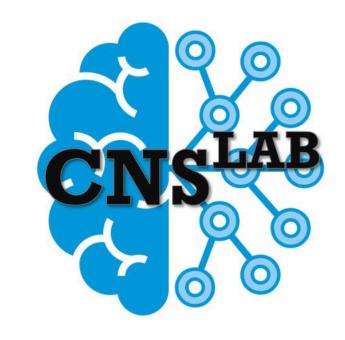
- Null hypothesis: the frequency of autism is the same across sexes
- Chi-squared test
- $p < 0.05 \rightarrow$ We reject the null hypothesis

Contingency Table

	Control	ASD	Total
Male	280	86	366
Female	232	103	335
Total	512	189	701

Removing the effect of age and sex

- Remove the effect of age and sex
 - Subtract $\beta_{age}age$ and $\beta_{sex}sex$ from raw measurements
 - Adjust 'reference point' to \overline{age} and \overline{sex}



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Part 2: Classifier Training and Evaluation with PyTorch



Outline

- 1. Adapt data to serve as PyTorch input
- 2. Design a PyTorch architecture
- 3. Train and evaluate models
- 4. Save and restore models
- 5. Compare two models

Autism Classification

- lacktriangle Train a model f that given an input x predicts a classification label y
- In this problem x denotes the input features of the questionnaire
- y is the subject classification to control and ASD based on their responses
- We will compare two ML classifiers



Data Preparation

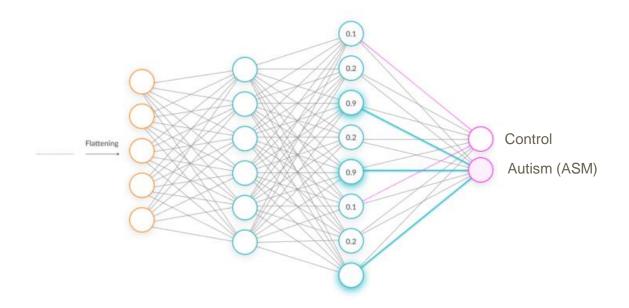
- Use the questionnaire answers A1 to A10 after residualization as input x
 to our classifier
- Transform the 'YES/NO' labels into integers

Country	Category #		Class
Control	NO		0
Autism	YES		1



Fully-connected Deep Neural Network

- Fully connected feedforward artificial neural network
- Consists of an input layer, hidden layers and an output layer

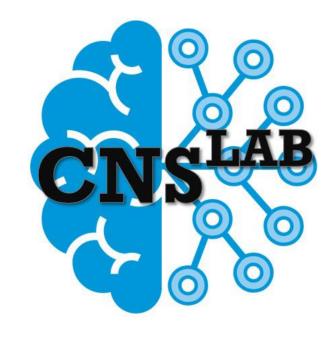


https://towardsdatascience.com/a-simple-guide-to-convolutional-neural-networks-751789e7bd88



Classifier Comparison

- Null hypothesis: the two models are equally accurate on the same cohort
- McNemar's Test
- $p < 0.05 \rightarrow We reject the null hypothesis$



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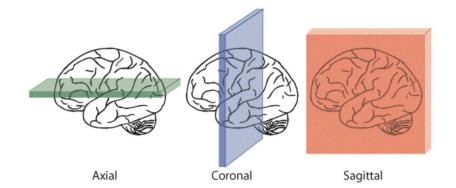
Part 3: Visualization of MRI Data

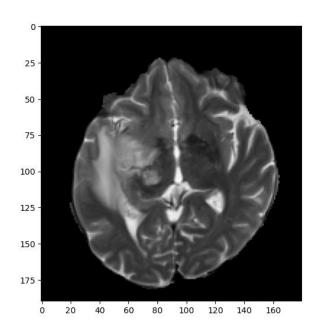


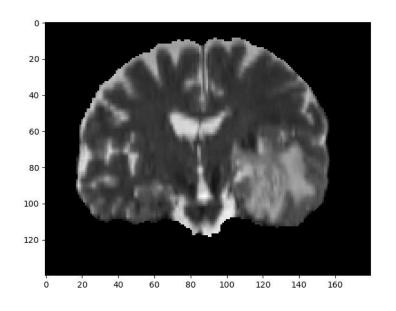


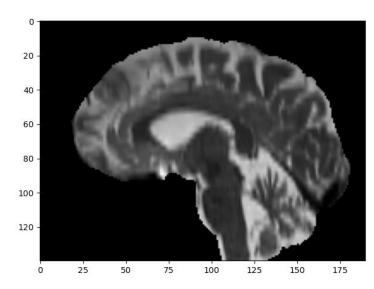


Visualization of an MRI









Goschke, T.: VL Kognitionspsychologie: Denken, Problemlösen, Sprache - Methoden der Kognitiven Neurowissenschaft: Kurze Einführung in die funktionelle Bildgebung (Summer 2013). Technical University Dresden, Dresden (2013)



Libraries for Neuroimaging Files Loading, Processing and Visualization



Community | NiBabel Home | Mailing list | License | Neuroimaging in Python

NiBabel

Read and write access to common neuroimaging file formats, including: ANALYZE (plain, SPM99, SPM2 and later), GIFTI, NIfTI1, CIFTI-2, MINC2, AFNI BRIK/HEAD, ECAT and Philips PAR/REC. In addition, NiBabel also supports FreeSurfer's MGH, geometry, annotation and morphometry files, and provides some limited support for DICOM.

NiBabel's API gives full or selective access to header information (metadata), and image data is made available via NumPy arrays. For more information, see NiBabel's <u>documentation site</u> and <u>API reference</u>.

Installation

To install NiBabel's current release with pip, run:

pip install nibabel

To install the latest development version, run:

pip install git+https://github.com/nipy/nibabel

When working on NiBabel itself, it may be useful to install in "editable" mode:

git clone https://github.com/nipy/nibabel.git
pip install -e ./nibabel

For more information on previous releases, see the release archive or development changelog.

Mailing List

Please send any questions or suggestions to the neuroimaging mailing list.

License

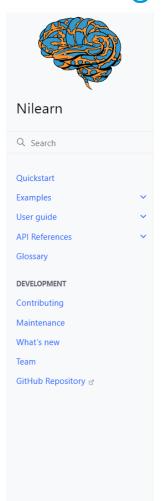
NiBabel is licensed under the terms of the MIT license. Some code included with NiBabel is licensed under the BSD license. For more information, please see the COPYING file.

NiBabel

Read and write access to common neuroimaging file formats; access header and image data



Libraries for Neuroimaging Files Loading, Processing and Visualization





Nilearn enables approachable and versatile analyses of brain volumes. It provides statistical and machine-learning tools, with instructive documentation & open community.

It supports general linear model (GLM) based analysis and leverages the scikit-learn Python toolbox for multivariate statistics with applications such as predictive modelling, classification, decoding, or connectivity analysis.



Examples

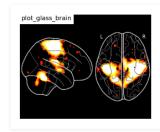
Discover functionalities by reading examples

User guide

Learn about neuroimaging analysis

0

Featured examples



Glass brain plotting

Explore how to retrieve data and plot whole brain cuts in glass mode.

Sparse rowerse covariance

Computing a connectome with sparse inverse covariance

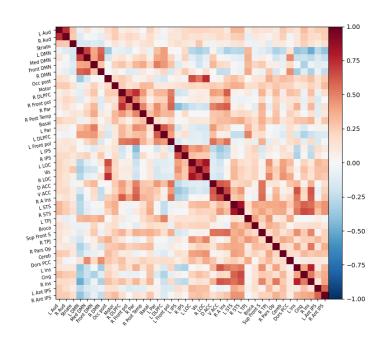
Construct a functional connectome using the sparse inverse covariance, and display the corresponding graph and matrix.

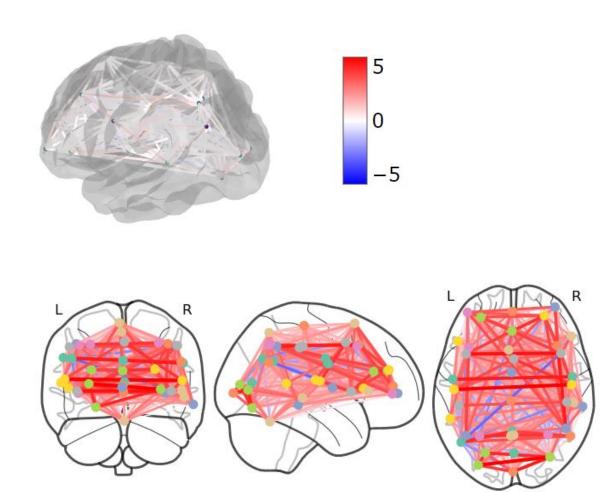
Nilearn

- Glass and surface plotting
- Very rich gallery of examples!



Visualizations of an fMRI







Libraries for Neuroimaging Files Loading, Processing and Visualization



Home · Quickstart · User Guide and Examples · Interfaces Index · Developers · About · Nipy

User Guide and Examples

Michael Notter's User Guide

Be sure to read Michael's excellent tutorials.

Examples

The following examples are <u>literate programming</u> documents which solve specific problems using Nipype. In some cases, the same problem is solved with multiple underlying tools.

- dMRI: Camino, DTI
- dMRI: Connectivity Camino, CMTK, FreeSurfer
- · dMRI: Connectivity MRtrix, CMTK, FreeSurfer
- dMRI: DTI Diffusion Toolkit, FSL
- dMRI: HARDI Diffusion Toolkit, FSL
- · dMRI: DTI, FSL
- dMRI: Group connectivity Camino, FSL, FreeSurfer
- dMRI: Group connectivity MRtrix, FSL, FreeSurfer
- dMRI: DTI MRtrix, FSL
- dMRI: Preprocessing
- <u>dMRI: TBSS on NKI RS data</u>

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Nipype

 Load, preprocess and analyze diffusion, structural and functional MRIs

