

Tutorial of NeuroRA Version 1.0.5

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The Tutorial of NeuroRA provides information on how to use the NeuroRA including its easy-to-use functions.

Before you read it, we assume that you are familiar with Python, especially the matrix operations based on Numpy. If not, don't worry. You only need to spend a little time learning the basic Python syntax and this toolkit is easy to understand.

If there is anything wrong, difficult to understand or having any useful advice during reading it, you can contact me (zitonglu1996@gmail.com) and I will be happy and thankful to know about it.

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This tutorial consists of these parts:

- Introduction & Installation
- Data Conversion
- Calculate the RDM (Representational Dissimilarity Matrices)
- Calculate the correlation coefficient between RDMs
- Visualization for results
- Save as a NIfTI file (for fMRI)
- Others

Part 1: Introduction

NueorRA is a Python toolbox for multimode neural data representational analysis.



Overview

Representational Similarity Analysis (RSA) has become a popular and effective method to measure the representation of multivariable neural activity in different modes.

NeuroRA is a novel and easy-to-use toolbox based on Python, which can do some works about RSA among nearly all kinds of neural data, including behavioral, EEG, MEG, fNIRS, ECoG, electrophysiological and fMRI data.

Installation

- `pip install NeuroRA`

Required Dependencies:

Numpy: a fundamental package for scientific computing

Matplotlib: a Python 2D plotting library

NiBabel: a package providing read +/- write access to some common medical and neuroimaging file formats

Part 2: Data Conversion

Type of Neural Bata	Data Conversion Scheme
fMRI	<p>Use <i>Nibabel</i> (https://nipy.org/nibabel/) to load fMRI data.</p> <pre>import nibabel as nib fmrifilename = "demo.nii" # the fmri data file name with full address data = nib.load(fmrifilename).get_fdata() # load fMRI data as ndarray</pre>
EEG/MEG	<p>Use <i>MATLAB EEGLab</i> (http://sccn.ucsd.edu/eeglab/) to do preprocessing and obtain .mat files, and use <i>Scipy</i> (https://www.scipy.org) to load EEG data (.mat).</p> <pre>import scipy.io as sio filename = "demo.mat" # the EEG/MEG data file name with full address data = sio.loadmat(filename)["data"] # load EEG/MEG data as ndarray</pre> <p>Or use <i>MNE</i> (https://mne-tools.github.io) to do preprocessing and return <i>ndarray</i>-type data.</p>
fNIRS	<p>For raw data from device, use <i>Numpy</i> (http://www.numpy.org) to load fNIRS data (.txt or .csv).</p> <pre>import numpy as np txtfilename = "demo.txt" # the fNIRS data file name with full address csvfilename = "demo.csv" data = np.loadtxt(txtfilename) # load fNIRS data as ndarray data = np.loadtxt(csvfilename, delimiter, usecols, unpack)</pre>
ECoG/sEEG	<p>Use <i>Brainstorm</i> (https://neuroimage.usc.edu/brainstorm/) to do preprocessing and obtain .mat files, and use <i>Scipy</i> to load ECoG data (.mat).</p>
Electrophysiology	<p>Use <i>pyABF</i> (https://github.com/swarden/pyABF) to load electrophysiology data (.abf).</p> <pre>import pyabf abf = pyabf.ABF("demo.abf") # the electrophysiology data file name with full address abf.setSweep(sweepNumber, channel) # access sweep data data = abf.sweepY # get sweep data with sweepY</pre>

Two functions, *NumPy.reshape()* & *NumPy.transpose()*, are recommended for further data transformation

Part 3: Calculate the RDM

Module *rdm_cal.py*:

◆ `bhvRDM(bhv_data, sub_opt=0, data_opt=1)`

A function for calculating the RDM based on behavioral data

Parameters:

- bhv_data:** The behavioral data. If `data_opt=0`, the shape of `bhv_data` must be `[n_cons, n_subs]`. `n_cons`, `n_subs` represent the number of conditions & the number of subjects.
- sub_opt:** Calculate the RDM for each subject or not. `1` or `0`.
- data_opt:** If `data_opt=1`, each subject's each trial has a value of data. If `data_opt=0`, each subject has a value of data, here ignore the effect of trials.

Returns:

- rdm/rdms:** If `sub_opt=0`, return only one `rdm` (shape: `[n_cons, n_cons]`). If `sub_opt=1`, return `rdms` (shape: `[n_subs, n_cons, n_cons]`).

◆ `eegRDM(EEG_data, sub_opt=0, chl_opt=0, time_opt=0)`

A function for calculating the RDM based on EEG/MEG/fNIRS data

Parameters:

- EEG_data:** The EEG/MEG/fNIRS data. The shape of `EEG_data` must be `[n_cons, n_subs, n_trials, n_chls, n_ts]`. `n_cons`, `n_subs`, `n_trials`, `n_chls`, `n_ts` represent the number of conditions, subjects, trials, channels, frequencies and time-points.
- sub_opt:** Calculate the RDM for each subject or not. `1` or `0`.
- chl_opt:** Calculate the RDM for each channel or not. `1` or `0`.
- time_opt:** Calculate the RDM for each time-point or not. `1` or `0`.

Returns:

- rdm/rdms:** If `sub_opt=0` and `chl_opt=0` and `time_opt=0`, return only one

`rdm` (shape: [n_cons, n_cons]).

If `sub_opt=0` and `chl_opt=0` and `time_opt=1`, return `rdms` (shape: [n_ts, n_cons, n_cons]).

If `sub_opt=0` and `chl_opt=1` and `time_opt=0`, return `rdms` (shape: [n_chls, n_cons, n_cons]).

If `sub_opt=1` and `chl_opt=0` and `time_opt=0`, return `rdms` (shape: [n_subs, n_cons, n_cons]).

If `sub_opt=0` and `chl_opt=1` and `time_opt=1`, return `rdms` (shape: [n_chls, n_ts, n_cons, n_cons]).

If `sub_opt=1` and `chl_opt=0` and `time_opt=1`, return `rdms` (shape: [n_subs, n_ts, n_cons, n_cons]).

If `sub_opt=1` and `chl_opt=1` and `time_opt=0`, return `rdms` (shape: [n_subs, n_chls, n_cons, n_cons]).

If `sub_opt=1` and `chl_opt=1` and `time_opt=1`, return `rdms` (shape: [n_subs, n_chls, n_ts, n_cons, n_cons]).

◆ `ecogRDM(ele_data, nchls, opt="all")`

A function for calculating the RDM based on ECoG/electrophysiological data

Parameters:

`ele_data`: The ECoG/electrophysiological data. The shape of `ele_data` must be [n_cons, n_trials, n_chls, n_ts]. n_cons, n_trials, n_chls, n_ts represent the number of conditions, trials, channels, frequencies and time-points.

`nchls`: The number of channels of the data.

`opt`: Calculate the RDM for each channel or for each time-point or not. "**channels**" or "**time**" or "**all**".

Returns:

`rdm/rdms`: If `opt="channels"`, return `rdms` (shape: [n_chls, n_cons, n_cons]).

If `opt="time"`, return `rdms` (shape: [n_ts, n_cons, n_cons])

◆ `fmriRDM(fmri_data, ksize=[3, 3, 3], strides=[1, 1, 1])`

A function for calculating the RDM based on fMRI data

Parameters:

fmri_data: The fmri data. The shape of **fmri_data** must be [n_cons, n_subs, nx, ny, nz]. n_cons, n_subs, nx, ny, nz represent the number of conditions, the number of subjects, the size of the fMRI data.

ksize: **ksize**=[kx, ky, kz] represents that the calculation unit contains k1*k2*k3 voxels.

strides: **strides**=[sx, sy, sz] represents the moving steps along the x, y, z.

Returns:

rdms: Return **rdms** for each calculation unit. The shape of **rdms** is [n_x, n_y, n_z, n_cons, n_cons]. Here, n_x, n_y, n_z represent the number of calculation units along the x, y, z.

Part 4: Calculate the Correlation Coefficient

Module *rsa_corr.py*:

◆ `rsa_correlation_spearman(RDM1, RDM2)`

A function for calculating the Spearman correlation coefficient between two RDMs

Parameters:

RDM1: The shape of **RDM1** must be [n_cons, n_cons].

RDM2: The shape of **RDM2** must be [n_cons, n_cons].

Returns:

corr: The **corr** contains two values: correlation coefficient and p-value. The shape of **corr** is [2,].

◆ `rsa_correlation_pearson(RDM1, RDM2)`

A function for calculating the Pearson correlation coefficient between two RDMs

Parameters:

RDM1: The shape of **RDM1** must be [n_cons, n_cons].

RDM2: The shape of **RDM2** must be [n_cons, n_cons].

Returns:

corr: The **corr** contains two values: correlation coefficient and p-value. The shape of **corr** is [2,].

◆ `rsa_correlation_kendall(RDM1, RDM2)`

A function for calculating the Kendalls tau correlation coefficient between two RDMs

Parameters:

RDM1: The shape of **RDM1** must be [n_cons, n_cons].

RDM2: The shape of **RDM2** must be [n_cons, n_cons].

Returns:

corr: The **corr** contains two values: correlation coefficient and p-value. The shape of **corr** is [2,].

◆ `rsa_similarity(RDM1, RDM2)`

A function for calculating the Cosine Similarity between two RDMs

Parameters:

RDM1: The shape of **RDM1** must be [n_cons, n_cons].

RDM2: The shape of **RDM2** must be [n_cons, n_cons].

Returns:

similarity: The Cosine Similarity.

◆ `rsa_distance(RDM1, RDM2)`

A function for calculating the Euclidean Distances between two RDMs

Parameters:

RDM1: The shape of **RDM1** must be [n_cons, n_cons].

RDM2: The shape of **RDM2** must be [n_cons, n_cons].

Returns:

dist: The Euclidean Distance.

◆ `rsa_permutation (RDM1, RDM2, iter=1000)`

A function for permutation test between two RDMs

Parameters:

RDM1: The shape of **RDM1** must be [n_cons, n_cons].

RDM2: The shape of **RDM2** must be [n_cons, n_cons].

iter: The number of iterations.

Returns:

p: The p-value.

Module ***corr_cal.py***:

◆ `bhvANDeeg_corr(bhv_data, EEG_data, sub_opt=0, bhv_data_opt=1, chl_opt=0, time_opt=0, method="spearman")`

A function for calculating the Similarity/Correlation Coefficient between behavioral data and EEG/MEG/fNIRS data

Parameters:

method: "spearman" or "pearson" or "kendall" or "similarity" or "distance". They correspond to different methods of calculation.

Returns:

corr/corrs: The correlation coefficients corresponding to the RDMs.

◆ `bhvANDecog_corr(bhv_data, ele_data, chls_num, ecog_opt="allin", method="spearman")`

A function for calculating the Similarity/Correlation Coefficient between behavioral data and ECoG/electricophysiological data

Returns:

corr/corrs: The correlation coefficients corresponding to the RDMs.

◆ `bhvANDfmri_corr(bhv_data, fmri_data, bhv_data_opt=1, ksize=[3, 3, 3], strides=[1, 1, 1], method="spearman")`

A function for calculating the Similarity/Correlation Coefficient between behavioral data and fMRI data

Returns:

corrs: The correlation coefficients corresponding to the RDMs. The shape of **corrs** is [n_x, n_y, n_z, 2].

◆ `eegANDfmri_corr(eeg_data, fmri_data, chl_opt=0, ksize=[3, 3, 3], strides=[1, 1, 1], method="spearman")`

A function for calculating the Similarity/Correlation Coefficient between EEG/MEG/fNIRS data and fMRI data

Returns:

corrs: The correlation coefficients corresponding to the RDMs.

Module ***corr_cal_by_rdm.py***:

◆ `eegrdms_corr(demo_rdm, EEG_rdms, method="spearman")`

A function for calculating the Similarity/Correlation Coefficient between RDMs based on EEG/MEG/fNIRS data and a demo RDM

Parameters:

demo_rdm: The shape must be [n_cons, n_cons].

EEG_rdms: The shape must be [n_ts, n_cons, n_cons] or [n_chls, n_cons, n_cons].

Returns:

corrs: The correlation coefficients corresponding to the RDMs.

◆ `fmrirdms_corr(demo_rdm, fMRI_rdms, method="spearman")`

A function for calculating the Similarity/Correlation Coefficient between RDMs based on fMRI data and a demo RDM

Parameters:

demo_rdm: The shape must be [n_cons, n_cons].

fmri_rdm: The shape must be `[n_x, n_y, n_z, n_cons, n_cons]`.

Returns:

corrs: The correlation coefficients corresponding to the RDMs. The shape of **corrs** is `[n_x, n_y, n_z, 2]`.

Part 4: Visualization for Results

Module *rsa_plot.py*:

◆ `plot_rdm_1(rdm) / plot_rdm_2(rdm)`

Two functions for plotting the RDM

◆ `plot_corrs_by_time(corrs, labels, time_unit=[0, 1])`

A function for plotting the correlation coefficients by time sequence

Parameters:

- corrs:** `corrs` represent the correlation coefficients point-by-point. Its shape must be `[n_cons, ts, 2]` or `[n_cons, ts]`.
- labels:** `labels` represent the names of conditions of RSA results.
- time_unit:** `time_unit=[start_t, t_step]`. Here, `start_t` represents the start time and `t_step` represents the time between two adjacent time-points.

◆ `plot_corrs_hotmap(eegcorrs, chllabels, time_unit=[0, 1], smooth=True)`

A function for plotting the correlation coefficients by time sequence

Parameters:

- eegcorrs:** `eegcorrs` represent each channels' correlation coefficients point-by-point. Its shape must be `[n_chls, ts, 2]` or `[n_chls, ts]`.
- chllabels:** `labels` represent the names of channels.
- time_unit:** `time_unit=[start_t, t_step]`. Here, `start_t` represents the start time and `t_step` represents the time between two adjacent time-points.
- smooth:** `True` or `False` represents smoothing the results or not.

Part 5: Save as a NIfTI file (for fMRI)

Module *corr_to_nii.py*:

◆ `corr_save_nii(corr, filename, affine, size=[60, 60, 60], ksize=[3, 3, 3],
strides=[1, 1, 1], p=1, r=0, similarity=0, distance=0)`

A function for saving the correlation coefficients as a .nii file

Parameters:

- corr:** `corr` represent the correlation coefficients. Its shape must be `[n_x, n_y, n_z, 2]`.
- filename:** The filename of the NIfTI file. Don't need a suffix.
- affine:** An affine array that tells you the position of the image array data in a reference space.
- p, r, similarity, distance:** They represent the threshold value for calculation.

Returns:

- img_nii:** The matrix form of the NIfTI file.

Part 6: Others

Module ***stuff.py***:

◆ `limtozero(x)`

A function for zeroing the value close to zero.

Parameters:

`x`: A value.

Returns:

`0`

◆ `get_affine(file_name)`

A function for getting the affine.

Parameters:

`file_name`: The file_name of a fMRI file.

Returns:

`affine`