

Script Demo for fMRItual based on NeuroRA

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Users can learn how to use [NeuroRA](#) to do research by representational analysis based on fMRI data.

This demo is based on the publicly available Haxby fMRI datasets.
(Reference: [Haxby, J. V. \(2001\). Distributed and Overlapping Representations of Faces and Objects in Ventral Temporal Cortex. Science, 293\(5539\), 2425–2430.](#)) [Nilearn](#) has been used to load this dataset and plot some results in this demo.

```
# -*- coding: utf-8 -*-

' a demo based on Haxby fMRI data '
# Users can learn how to use Neurora to do research based on fMRI data.

__author__ = 'Zitong Lu'

from nilearn import datasets, plotting
from nilearn.image import index_img, mean_img
import numpy as np
import pandas as pd
import nibabel as nib
from neurora.stuff import get_affine, datamask
from neurora.nps_cal import nps_fmri, nps_fmri_roi
from neurora.rsa_plot import plot_rdm
from neurora.rdm_cal import fmriRDM_roi, fmriRDM
from neurora.corr_cal_by_rdm import fmriRDMs_corr
from neurora.corr_to_nii import corr_save_nii

"""*****          Section 1: Loading example data          *****"""
""" Here, we use Nilearn toolbox for loading data and processing """
""" you can learn this process from Nilearn
(http://nilearn.github.io/index.html) """
```

```

# load Haxby dataset (here, we only use subject2's data for this
example)
haxby_dataset = datasets.fetch_haxby()

# load the fMRI data filename & mask data filename
func_filename = haxby_dataset.func[0]
mask_filename = haxby_dataset.mask

# read label information of the experiment
labelinfo = pd.read_csv(haxby_dataset.session_target[0], sep=' ')
labels = labelinfo['labels']

""""*****                               Section 2: Preprocessing                               *****""""

# get mask data NumPy array
maskdata = nib.load(mask_filename).get_data()

# get the size of the data
nx, ny, nz = maskdata.shape

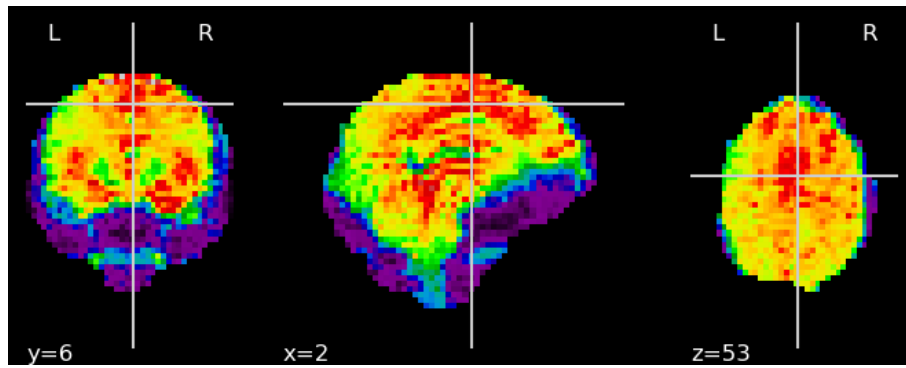
# labels of seven ategories
categories = ["face", "cat", "house", "chair", "shoe", "bottle",
"scissors"]
# numbe of conidtions: 7
ncon = len(categories)

# get fmri data under 7 conditions
# here we average the data under different conditions
fmri_data = np.full([ncon, nx, ny, nz], np.nan)

for i in range(ncon):
    img = mean_img(index_img(func_filename,
labels.isin([categories[i]])))
    fmri_data[i] = datamask(img.get_data(), maskdata)

# get fmri data under 'face'-condition
face_img = nib.Nifti1Image(fmri_data[0], affine=img.affine)
# have a look
plotting.plot_epi(face_img)
plotting.show()

```



```
# reshaoe the data: [ncon, nx, ny, nz] -> [ncon, nsubs, nx, ny, nz]
# here just one subject's data
fmri_data = np.reshape(fmri_data, [ncon, 1, nx, ny, nz])

""""Section 3: Calculating the neural pattern similarity (for ROI)""""

# get mask of 'mask_face' in the dataset
mask_face_filename = haxby_dataset.mask_face[0]
mask_face_data = nib.load(mask_face_filename).get_data()

# get input data under two condition
# here, "face"-condition vs. "cat"-condition
nps_fmri_data = fmri_data[[0, 6]]

# calculate the neural pattern similarity (NPS) for ROI between two
stimulus
nps_roi = nps_fmri_roi(nps_fmri_data, mask_face_data)

# print the NPS result
print(nps_roi)
```

9.994484681410596982e-01, 5.880839542727496111e-43

```
""""Section 4: Calculating the neural pattern similarity (Searchlight)""""

# calculate the neural pattern similarity (NPS) between two stimulus
nps = nps_fmri(nps_fmri_data)

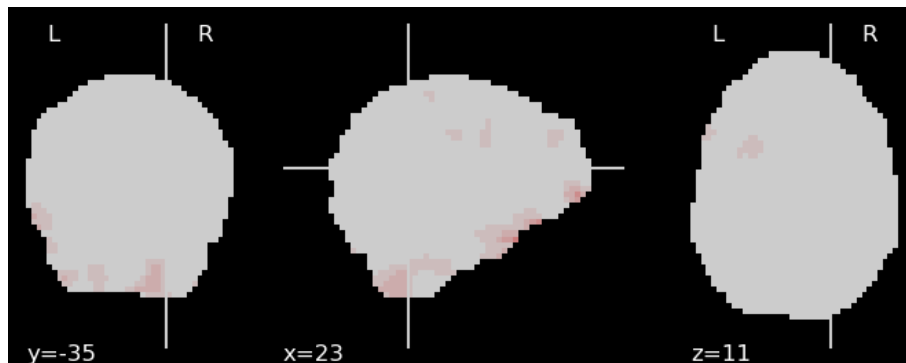
# convert the NPS results into a .nii file
savefilename = "nps_img"
affine = get_affine(mask_filename)
```

```

corr_save_nii(nps[0], filename=savefilename, affine=affine, size=[nx,
ny, nz], plotrlt=True, plotrlt=False)

# have a look
plotting.plot_epi(savefilename+".nii")
plotting.show()

```



***** Section 5: Calculating the RDM for ROI and Plotting *****

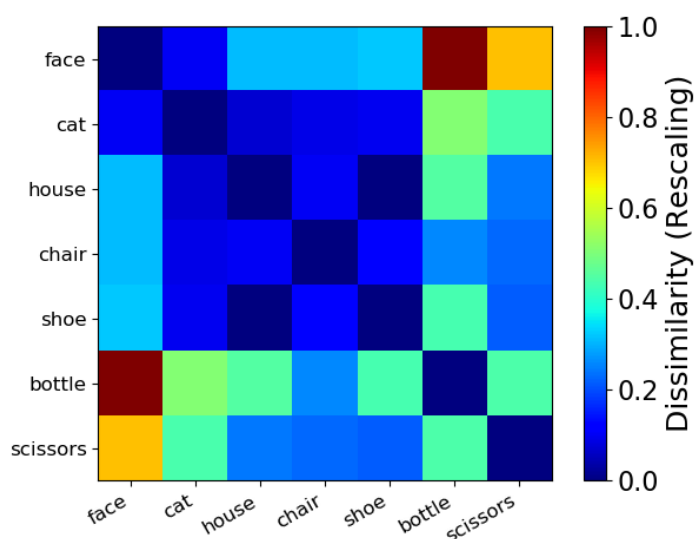
```

# get mask of "mask_vt" in the dataset
mask_vt_filename = haxby_dataset.mask_face[0]
mask_vt_data = nib.load(mask_vt_filename).get_data()

# calculate the RDM for ROI
rdm_roi = fmriRDM_roi(fmri_data, mask_vt_data)

# plot the RDM
plot_rdm(rdm_roi, rescale=True, conditions=categories)

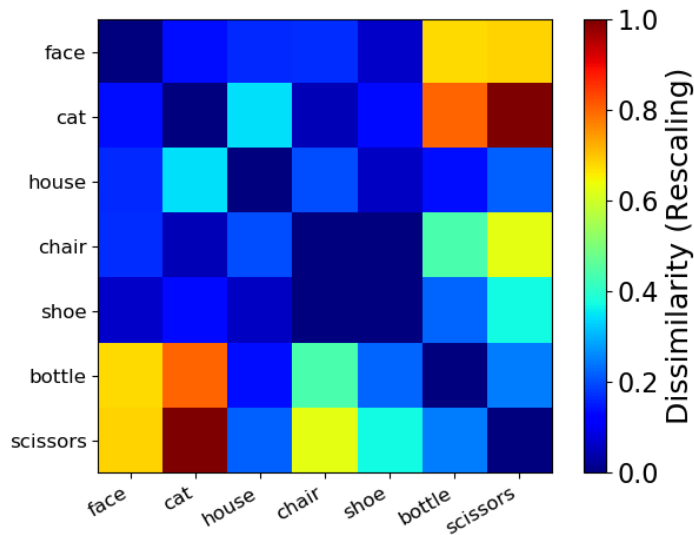
```



***** Section 6: Calculating the RDM by Searchlight and Plotting *****

```
# calculate the RDMs by Searchlight
fmri_RDMs = fmriRDM(fmri_data)

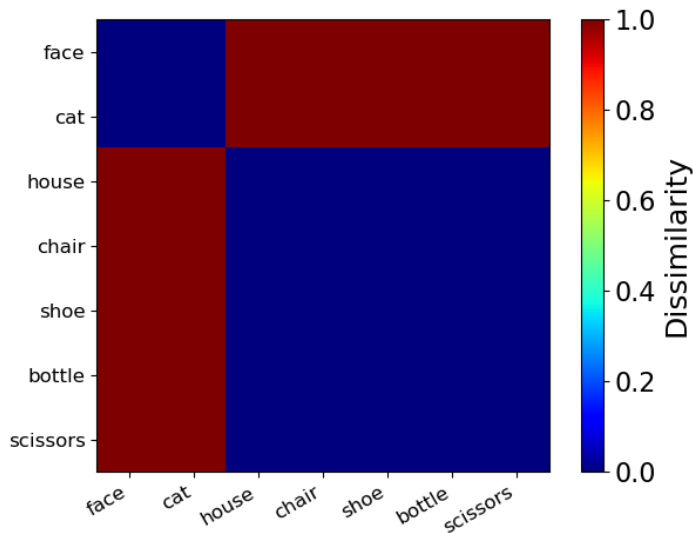
# plot one of the RDMs
plot_rdm(fmri_RDMs[20, 30, 30], rescale=True, conditions=categories)
```



```
***** Section 7: Calculating the representational similarities *****
***** between a coding model and neural activities *****

# Create a RDM for "animate-inanimate" coding model
# which means the representations of animate matters are highly similar
# and the representations of inanimate matters are highly similar
model_RDM = np.array([[0, 0, 1, 1, 1, 1, 1],
                      [0, 0, 1, 1, 1, 1, 1],
                      [1, 1, 0, 0, 0, 0, 0],
                      [1, 1, 0, 0, 0, 0, 0],
                      [1, 1, 0, 0, 0, 0, 0],
                      [1, 1, 0, 0, 0, 0, 0],
                      [1, 1, 0, 0, 0, 0, 0]])

# plot the model RDM
plot_rdm(model_RDM, conditions=categories)
```



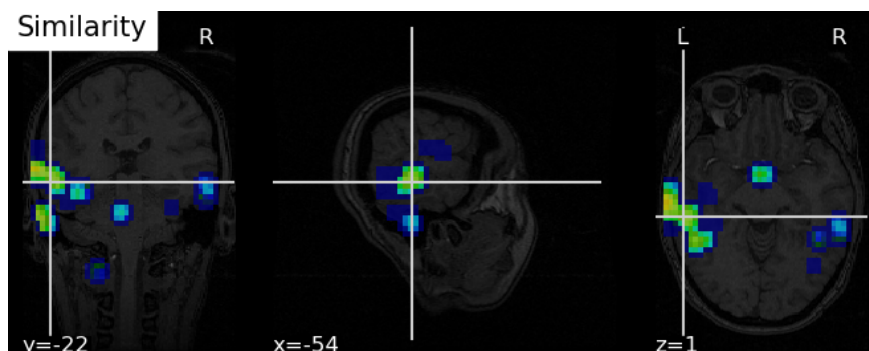
```
# calculate the similarities between model RDM and searchlight RDMs
corrs = fmri_rdm_corr(model_RDM, fmri_RDMS)

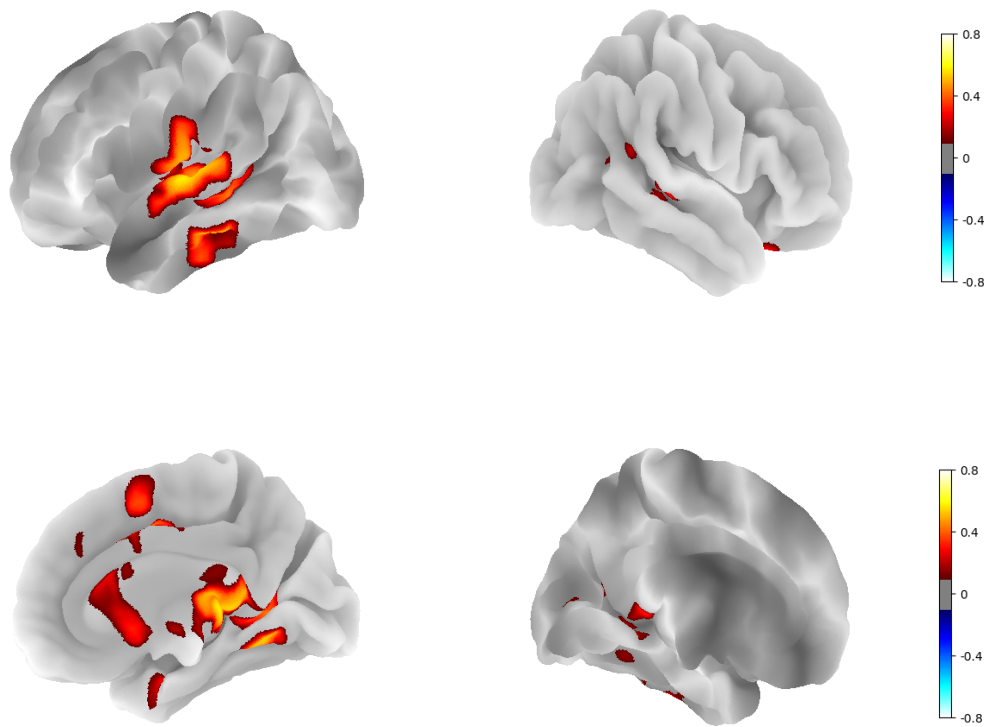
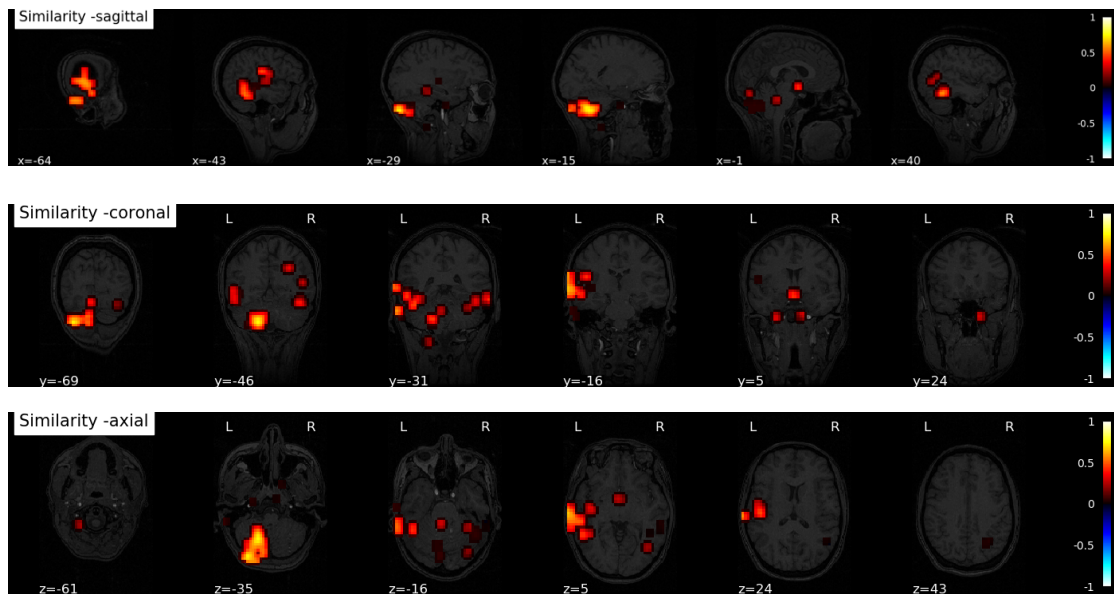
""""*****      Section 8: Saving the RSA result and Plotting      *****""""

# load the filename of anatomical image as the background for plotting
ant_filename = haxby_dataset.anat[0]

# get the affine info
affine = get_affine(mask_filename)

# save the RSA result as a .nii file
# and visualize the result automatically
# p < 0.05, FDR-correct
rsarltfilename = "demo2_rsarlt_img"
img = corr_save_nii(corrs, filename=rsarltfilename, affine=affine,
corr_mask=mask_filename, size=[40, 64, 64], p=0.001, plotrlt=True,
img_background=ant_filename)
```

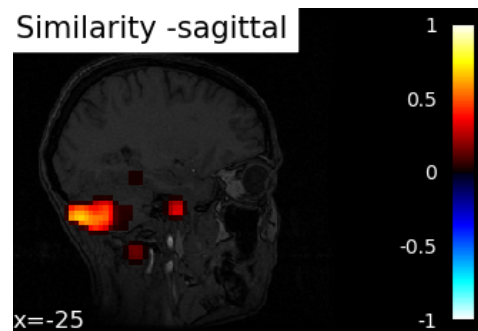




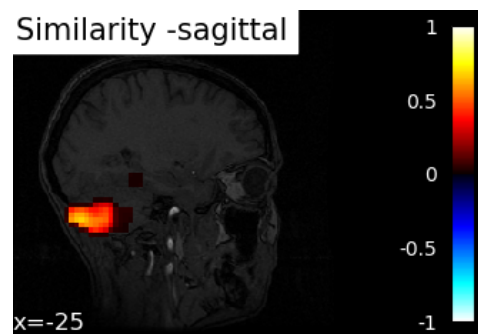
```
# Users can plot the RSA results independently by functions below
from neurora.rsa_plot import plot_brainrsa_regions
from neurora.rsa_plot import plot_brainrsa_montage
from neurora.rsa_plot import plot_brainrsa_glass
from neurora.rsa_plot import plot_brainrsa_surface

# here use a [5, 5, 5] cube to remove the significant area smaller than
it
```

```
# before filtering
plot_brainrsa_montage(rsarltfilename, slice=[[-25], 0, 0],
background=ant_filename)
```



```
# after filtering
plot_brainrsa_montage(rsarltfilename, threshold=125, slice=[[-25], 0,
0], background=ant_filename)
```



Users can see more details below:

GitHub Website: <https://github.com/neurora/NeuroRA>

NeuroRA Website: <https://neuroragithub.io/NeuroRA/>

PyPi Website: <https://pypi.org/project/neurora/>

BioRxiv Paper: <https://doi.org/10.1101/2020.03.25.008086>