Tutorial: Pereira, Neuroimage 2009 00000000 Scenarios



Classification of fMRI data

Roma Siugzdaite & Ruth Seurinck

Department of Data-Analysis

April 17th, 2014

イロト イポト イヨト イヨト

э

Roma Siugzdaite & Ruth Seurinck Classification of fMRI data

Outline

1 Toolboxes

- 2 Tutorial: Pereira, Neuroimage 2009
 - Creating and selecting features
 - Pre-processing features
 - Choosing a (classification) algorithm
 - Cross-validation
 - Assessing the results
- 3 Scenarios
 - Haxby Data (Pronto manual)
 - IXI Data (Pronto manual)
 - Ishai Data (Beta images)
 - Motion After Effect (Beta images)

э

- Multi Voxel Pattern Analysis: MVPA
 - requires Matlab (OctaveMVPA available)
 - imports images from SPM, AFNI and Brainvoyager
 - https://code.google.com/p/princeton-mvpa-toolbox/
- pyMVPA
 - runs on Python
 - includes a searchlight (moves across the brain to calculate local multivariate information content)
 - http://www.pymvpa.org/
- Functional Real-time Interactive Endogenous Neuromodulation and Decoding: FRIEND
 - FSL toolbox
 - reference paper: Sato et al., PLOS one 2013
 - http://www.nitrc.org/projects/friend/

- Pattern Recognition for Neuroimaging Toolbox: PRoNTo (what we will use)
 - requires Matlab and SPM8
 - reference paper: Schrouff et al., Neuroinformatics 2013
 - http://www.mlnl.cs.ucl.ac.uk/pronto/

◆□▶ ◆□▶ ◆三▶ ◆三▶ 三日 うのの

Two specials:

- Representational Similarity Analysis: RSA http://www.mrc-cbu.cam.ac.uk/methods-andresources/toolboxes/
- 2 Cortical surface-based searchlight decoding (Hayneslab, Berlin) https://sites.google.com/site/hayneslab/links/

3

General principle

Infer variable from multivariate brain information

- categorical variable: CLASSIFICATION (Scenario 1: Did the subject see houses or faces?)
- 2 scale variable: REGRESSION (Scenario 2: Does the brain contain information that can predict age?)

◆□▶ ◆□▶ ◆三▶ ◆三▶ 三日 うのの

Feature Selection

Selecting informative voxels and isolating relevant information:

- use a brain mask to isolate only voxels within the brain
- further feature selection in PRoNTo: use a priori mask image (e.g. only fusiform gyrus)
- classification: association with class labels? (e.g. faces or houses)

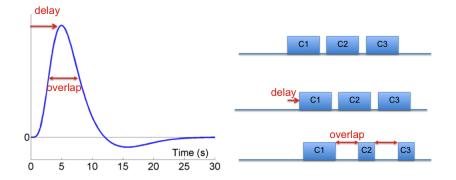
Creating and selecting features

Associating scans with classes

- **1** use the entire timeseries of the experiment: what scans should be associated with what event?
 - HRF-correction (delay and overlap)
 - block design works well, fast event-related design does not
 - choose scans that coincide with event of interest
 - overlapping scans should be omitted, given sluggish HRF

3

| Toolboxes | Tutorial: Pereira, Neuroimage 2009 | Scenarios |
|---------------------------------|------------------------------------|-----------|
| 000 | ००●०००००० | 00000 |
| Creating and selecting features | | |



Roma Siugzdaite & Ruth Seurinck

Classification of fMRI data

◆□▶ ◆□▶ ◆臣▶ ◆臣▶ 臣 のへで

- use the entire timeseries of the experiment: what scans should be associated with what event?
- **2** use Beta-images from a regular GLM analysis you perform prior to classification
 - recommended procedure for fast-event related design
 - results depend on the goodness-of-fit of the univariate GLM

3

イロト イポト イヨト イヨト

Pre-processing features

Pre-processing features

- not the same as pre-processing images prior to GLM!
- adapting features/voxels to make them comparable across the brain
- independently for training and test data (see cross-validation)
- different possibilities in PRoNTo: minimum requirement is to mean centre the data

Choosing a (classification) algorithm

Algorithms

PRoNTo has several algorithms for both regression and classification

- regression: for the moment only if you have one image per subject and do a group analysis
- 2 classification: can be performed both at the individual and the group level

Cross-validation

- Dividing the data in *n* folds (e.g. runs, subjects, ...).
- Use *n* − 1 folds as training data, the remaining fold is the test data
- Rotate which fold is the test data.
- Result for each fold and an overall result

<ロ> (四) (四) (三) (三) (三)

Assessing the results

Is my result statistically signficant?

We focus on the classification:

- we get a classification accuracy: % events that has been classified correctly
- PRoNTo:
 - regular accuracy (each category has equal frequencies)
 - balanced accuracy (corrected accuracy if one category occurs more)
 - an accuracy for each class/category
 - a permutation test for the accuracy with a p-value
- classification accuracies for each subject can be combined at the group-level in a regular test (e.g. t-test, non-parametric test, ...) compared to chance level (e.g. 50 % for two-class problem)

Locating informative voxels: the weights map

If your algorithm uses a linear model, you can display the weights, or the relative amount of information a voxel conveys, in a map.

You can never focus on certain peaks and coordinates, as it is a relative contribution within a multivariate analysis!

Haxby Data (Pronto manual)

Single-subject Block Design

- one subject
- viewing 8 categories: faces, houses, cats, chairs, bottles, scissors, shoes and scrambled pictures limited to two-class problem of houses and faces
- block design, entire timeseries available
- 12 runs, runs used as folds
- Leave-One-Block-Out cross-validation (run = block)

Regression example

- one group of 102 subjects
- one anatomical image and the age per subject
- subjects used as folds
- Leave-One-Subject-Out cross-validation
- correlation and MeanSquaredError (MSE) as output measure

Ishai Data (Beta images)

Toolboxes

What to do with a classification problem in a fast-event related design?

- Ishai et al., Journal of Cognitive Neuroscience 2000 (could be analysed as a block design as well)
- one subject, three classes: faces, houses, chairs and their scrambled pictures
- 12 runs, only a Beta-image for each category and each run
- for a two-class problem: faces versus houses
 - model two groups, one with the 12 Beta-images of the face category, one with the 12 Beta-images of the house category
 - make sure the images are in the same order across groups!
 - Leave-One-Subject-per-Group-Out cross-validation (run = subject)

Motion After Effect (Beta images)

Toolboxes

What if you have only one run per subject?

- Seurinck et al., Journal of Cognitive Neuroscience 2011 (could be analysed as a block design as well)
- motion after effect localizer: rotating disk with continuous direction (MAE) and rotating disk with alternating direction (noMAE)
- 13 subjects, only a Beta-image for each category (MAE and noMAE)
- for a two-class problem: MAE versus noMAE
 - model two groups, one with the 13 Beta-images of the MAE category, one with the 13 Beta-images of the noMAE category
 - make sure the images are in the same order across groups!
 - Leave-One-Subject-per-Group-Out cross-validation (subject = subject)

< ロ > < 同 > < 回 > < 回 > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ >

Let's try...

Roma Siugzdaite & Ruth Seurinck Classification of fMRI data ・ロト・日本・モート ヨー シタぐ