# Machine-Learning for Neuroimaging

State of the art

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#### Introduction



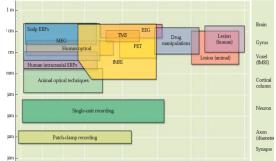
## Neuroimaging 101

#### **NMR Techniques**

- MRI
  - GMD
- fMRI
  - BOLD

#### Other techniques

- PET
- MEG/EEG
- NIRS/SPIR
- **.**.



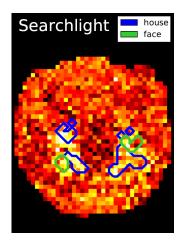


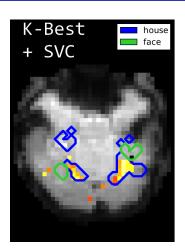


### **Techniques**



### Selecting part of interest







## Regression and Classification

- Support Vector Machine
  - Find an hyperplane that maximizes the distance of the closest points.
  - Focus on the max margin size.
- Logistic Regression
  - Find an hyperplane that maximize the probability (P(Y = y|X)) for points to be on the good side.
  - Uses a likelihood function.

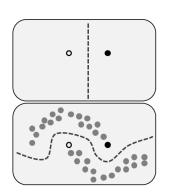
Table 1 | Five fold cross validation accuracy scores obtained for different values of parameter C (± SD), best scores are shown in bold.

C value	0.0005	0.001	0.005	0.01	0.05	0.1
ℓ <sub>1</sub> Logistic regression	0.50 ± 0.02	0.50 ± 0.02	0.57 ± 0.13	0.63 ± 0.11	<b>0.70</b> ± 0.12	0.70 ± 0.12
ℓ <sub>2</sub> Logistic regression	$0.60 \pm 0.11$	$0.61 \pm 0.12$	$0.63 \pm 0.13$	$0.63 \pm 0.13$	0.64 ± 0.13	$0.64 \pm 0.13$
ℓ <sub>1</sub> SVM classifier (SVC)	$0.50 \pm 0.06$	$0.55 \pm 0.12$	$0.69 \pm 0.11$	$0.71 \pm 0.12$	$0.69 \pm 0.12$	$0.68 \pm 0.12$
$\ell_2$ SVM classifier (SVC)	$0.67 \pm 0.12$	$0.67 \pm 0.12$	$0.67 \pm 0.12$	$0.66 \pm 0.12$	$0.65 \pm 0.12$	$0.65 \pm 0.12$



# Semi-supervised learning

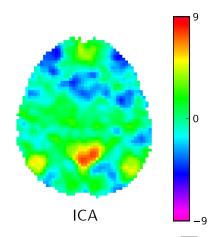
- Inputs : Labeled and non-labeled data
- Clustering for classification LDS: Low-density separation TSVM: Transductive SVM





## Extracting brain's network with ICA

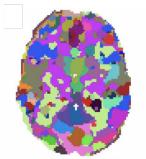
- Limits Neuroimaging
- Need for cognitive ontology (BrainMap Database)
- Improving Reverse Inference





## Clustering with K-means approach

■ Each voxels is assigned to the nearest center, forming clusters



Kmeans, 1000 clusters



Kmeans, 100 clusters

#### Case studies

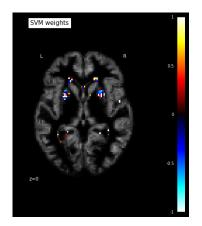


Case studies

Diseases and condition prediction

## Age and dementia prediction

- Record MRI based on GMD.
- Train a dementia classification model and an age regression model.
- Look at the involved parts of the brain.



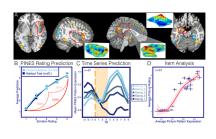


Case studies

Diseases and condition prediction

## Feelings and emotions

- Showing uncomfortable pictures while recording fMRI (BOLD).
- Getting "bad feeling grades" back (1 to 5)
- Train the model and evaluate it using K-Fold cross-validation



Chang LJ, Gianaros PJ, Manuck SB, Krishnan A, Wager TD (2015) A Sensitive and Specific Neural Signature for Picture-Induced Negative Affect. PLoS Biol 13(6): e1002180. https://doi.org/10.1371/journal.pbio.1002180

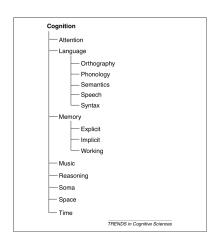
<sup>2.</sup> https://github.com/intv0id/ADST2018/blob/master/SentimentAnalysis/SentimentAnalysis.ipynb

#### Conclusion



#### Reverse inference Limits

- Neuroimaging limits
- Need for cognitive ontology (BrainMap Database)
- Improving Reverse Inference



<sup>1.</sup> Russell A. Poldrack, Can cognitive processes be inferred from neuroimaging data?, In Trends EURECO Cognitive Sciences, Volume 10, Issue 2, 2006, Pages 59-63, ISSN 1364-6613, https://doi.org/10.1016/j.tics.2005.12.004.

#### The future of machine learning for Neuroimaging

- Be able to detect and understand physical disease such as ADHD
- Be able to detect and understand mental disease such as schizophrenia
- European Brain Project





Questions?

