

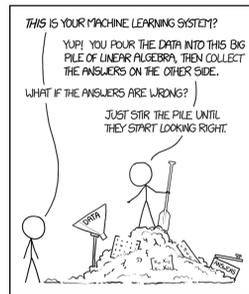
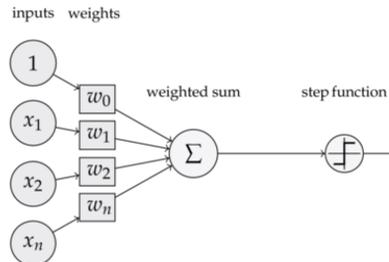
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Why?

- Many modern approaches to neuroimaging data analysis (i.e., Multivariate Pattern Analysis) rely on machine learning-based classifiers
- Classical approaches (e.g., Support Vector Machines or Logistic Regression) are inherently limited in the kinds of classification they can perform, and become computationally costly with large amount of data
- Modern neural network classifiers are more flexible and computationally tractable, at the cost of more researcher degrees of freedom without any established best practices
- To facilitate the exploration of deep learning in neuroscience, we need tools and structure for researchers to better engage with and manage additional complexity

What is deep learning?

- Linear algebra, but more fundable
- Start with a perceptron:



<https://tex.stackexchange.com/questions/104334/tikz-diagram-of-a-perceptron>

- Wire a whole bunch of them together, sprinkle in some other data transformations to taste
- Feed some feature values into the first layer, propagate results to the end
- Compare the last layer to your class labels and use backpropagation to yell at the model
- Repeat until you have solved consciousness and/or created SkyNet

Have questions?

- Drop us a line, we'll be happy to discuss your project or help you get set up.
- Email:
 - delineate.toolbox@unl.edu
 - kkuntzelman2@unl.edu
 - matthew.r.johnson@unl.edu
- Project website:
 - <http://delineate.it>
- Project git repository:
 - <https://bitbucket.org/delineate/delineate>



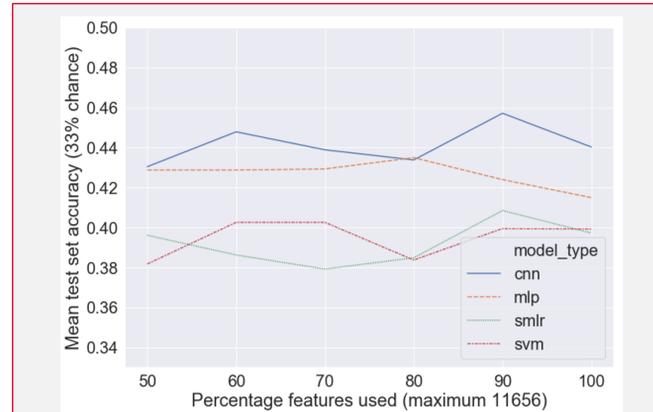
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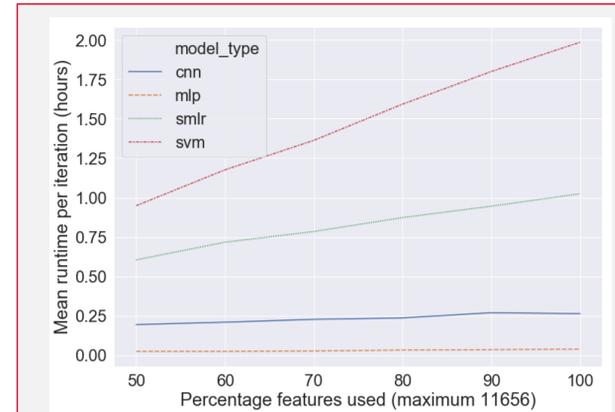
All with convenient JSON specification!
No need to write new code for a new model.



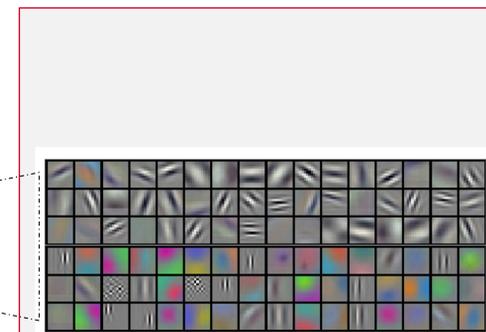
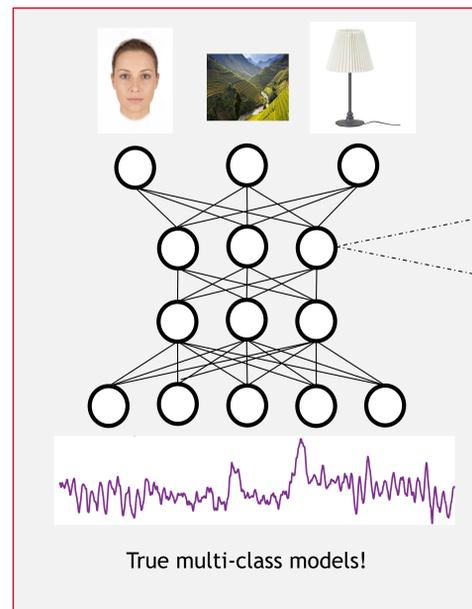
Features



Classify simple stuff as accurately as classical MVPA!



...but way faster!



Rediscover feature detectors!
(yet again:
Comon, 1994;
Olshausen & Field, 1996;
Bell & Sejnowski, 1997;
etc...)

What do you need to get started?

Hardware

- Linux workstation (Windows should work too)
- A decent GPU (nVIDIA with CUDA capability)

Software

- Python
- GPU drivers for parallel computing
- Theano or Tensorflow (These do the actual work but are Not Fun to interact with)
- Keras (Makes things friendlier if you know how to code)
- DeLINEATE (<http://delineate.it> or <https://bitbucket.org/delineate/delineate/overview>)

Files

- Data
- A loader function to get your data into the right shape (you can model this off of included examples)
- JSON model specification

Ongoing Development

- We take requests!
- Support for non-sequential models
- Support for additional backends
- Some sort of GUI, maybe, if we have to
- Better documentation
- Include prefab published models

Talk to actual users

- Poster B100, right now
- Poster B102, right now
- Poster D104, Monday morning (gross)

References & Acknowledgements

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