
A Gentle Introduction To Machine Learning

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Outline

- Why Use Machine Learning?
- Workflow
- Resources
- Final Comments



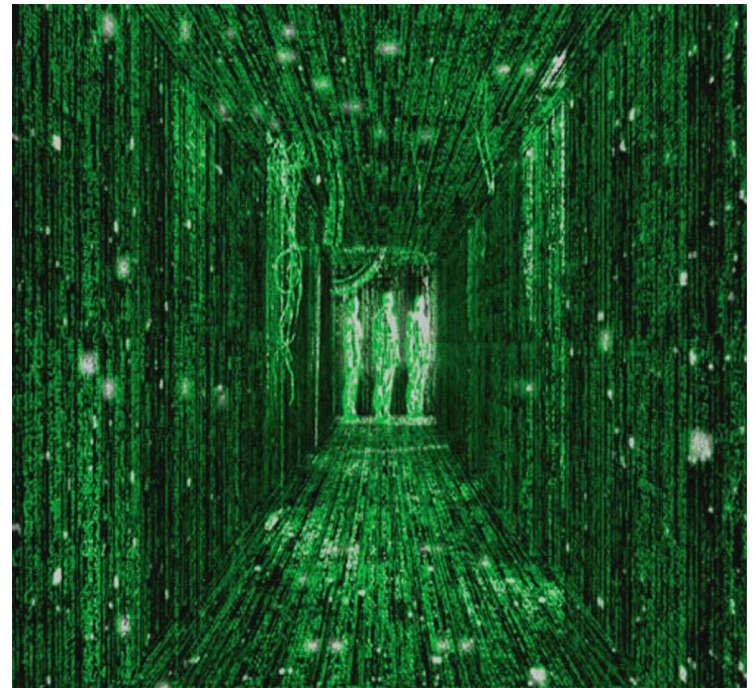
Why Use Machine Learning?

- Drowning in data
- Computers are cheap, humans are expensive
- Psychic superpowers (sometimes)



Types of Problems

- Regression (Supervised)
 - Predict housing prices
- Classification (Supervised)
 - Handwritten digit recognition
- Clustering (Unsupervised)
 - Document tagging



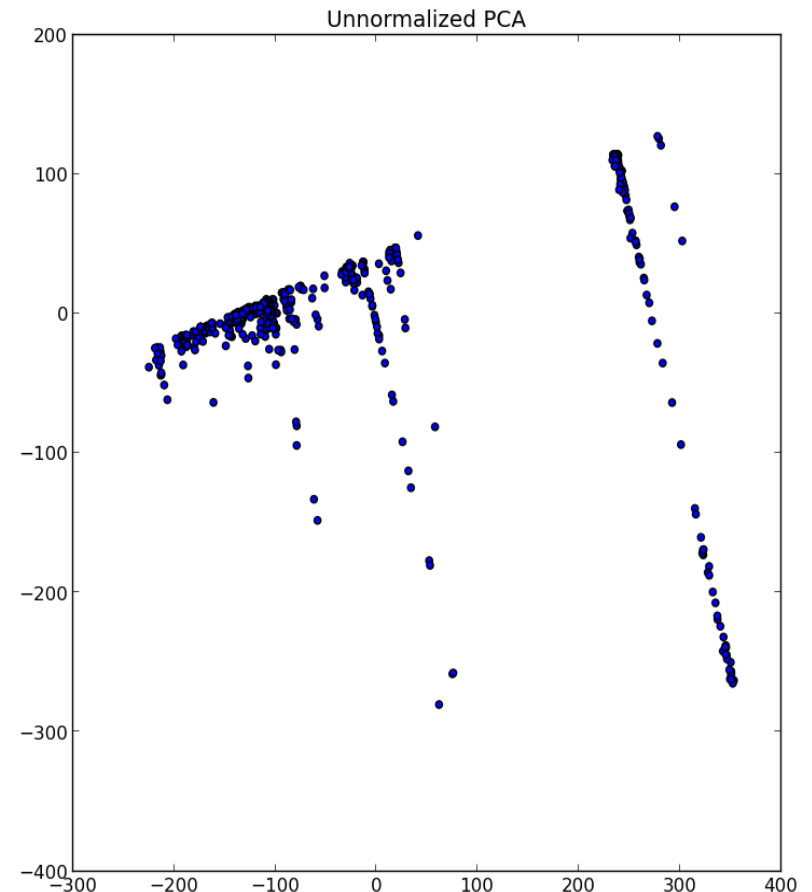
Where to Start?

- Know your data
 - If labeled, supervised learning
 - Unlabeled, try unsupervised
- Clean it up
 - Normalize by removing mean and dividing by variance
 - Visualize in 2D
- Separate training data
 - Try 80/20% train/test split, randomly chosen



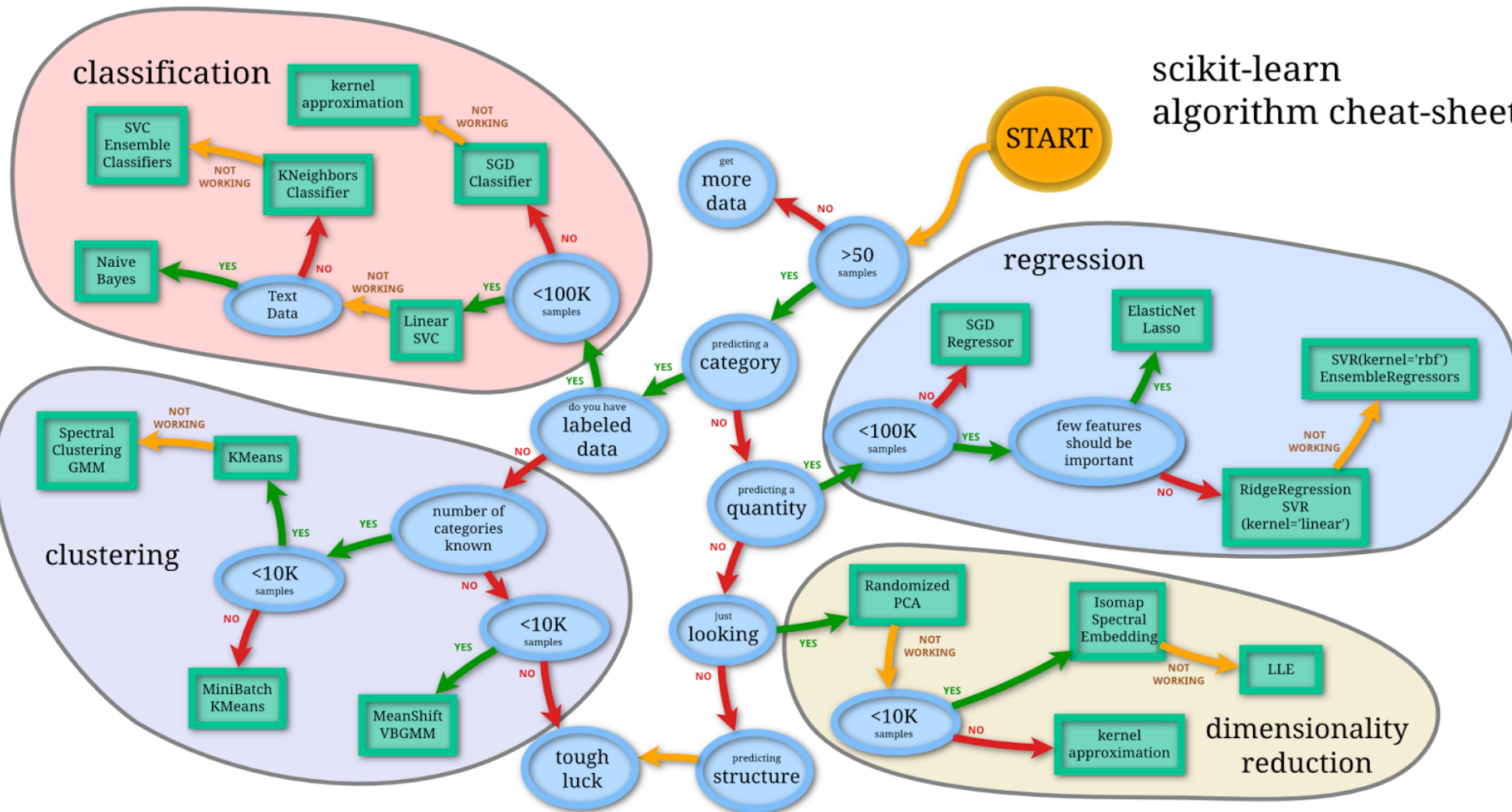
Preprocessing

- Typically normalize by subtracting mean and dividing by variance
- Use Principle Component Analysis (PCA) to keep structure while reducing dimensions
- PCA to plot N-dimensional data in 2D or 3D



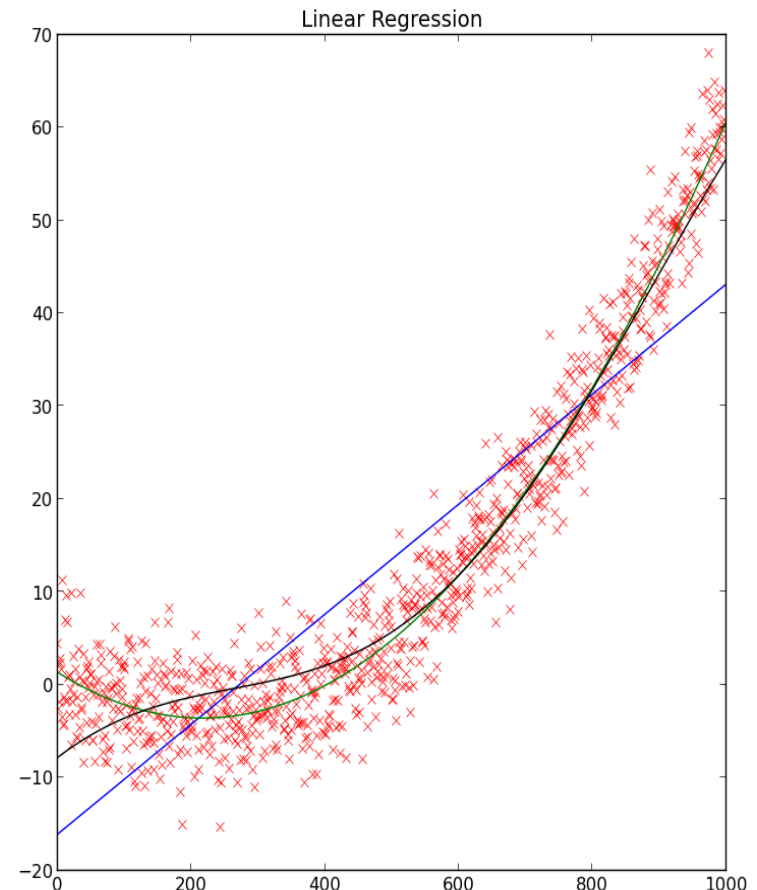
Selecting an Algorithm

scikit-learn
algorithm cheat-sheet



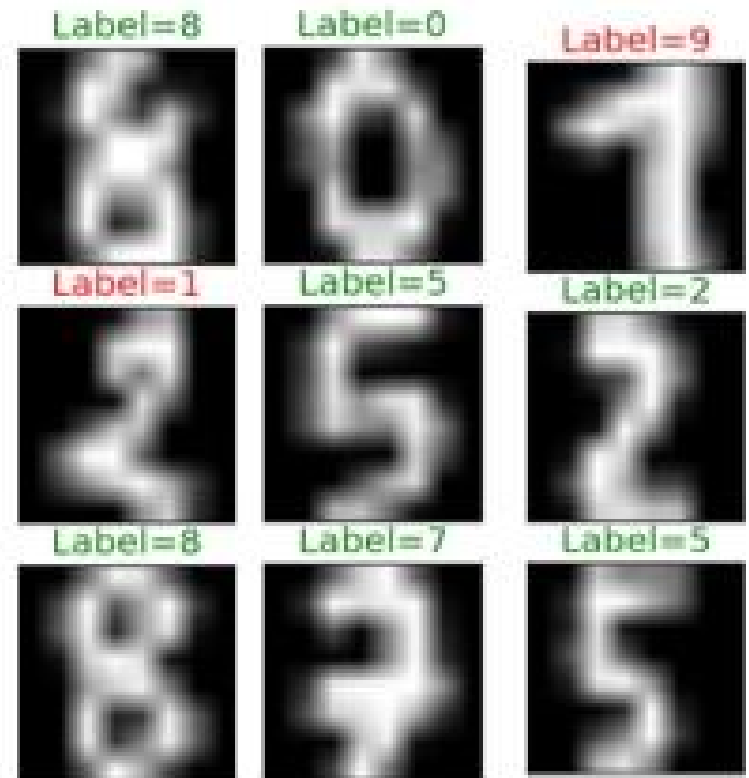
Linear Regression

- Find the "best fit" line
- Outliers will greatly affect results
- Perform regression into different basis
- Basis can be Fourier, polynomial, wavelet, etc.



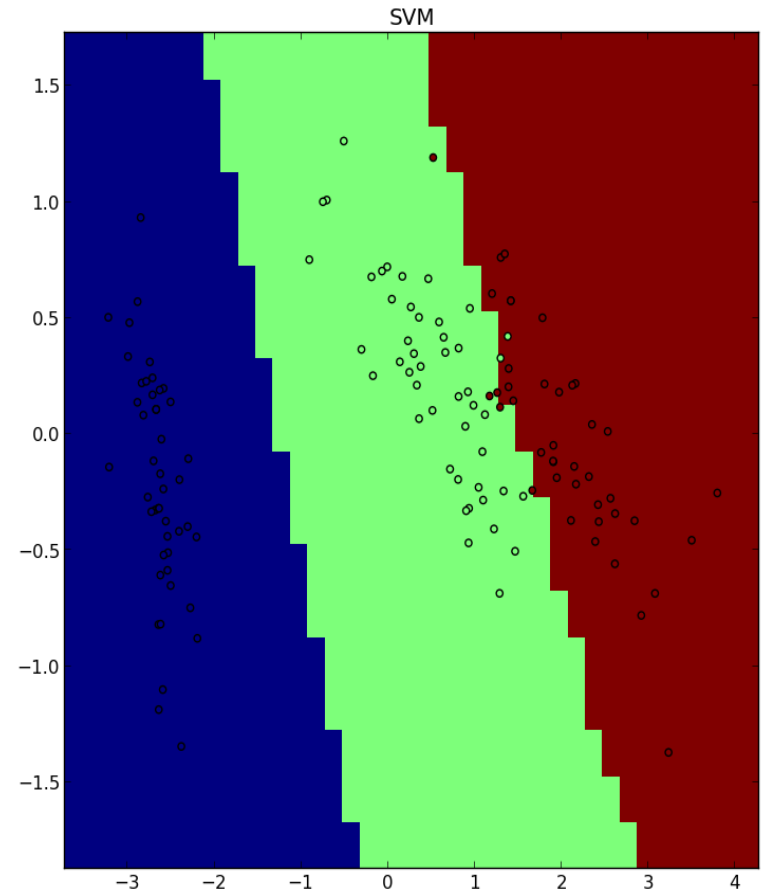
Logistic Regression

- Optimize parameters for each class label
- Choose class with highest probability
- Can be very powerful, especially after PCA



Support Vector Machine (SVM)

- Margin parameter is a configurable "allowed error" to account for class overlap
- Boundaries use a semi-arbitrary "kernel" function
- Linear, polynomial, wavelet, sigmoid



Data

- **from sklearn import datasets**
- Iris, Digits are excellent for classification
- Boston for regression
- Any classification dataset (sans labels) for clustering
- Very good for generating data



Resources

- Scikit-learn documentation and examples
 - [The infamous cheat sheet](#)
 - Coursera courses
 - [Andrew Ng's Machine Learning](#)
 - Pattern Recognition and Machine Learning
 - Christopher M. Bishop
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Final Comments

- Machine learning is a spectrum
 - Data preprocessing is vital
 - Prefer simple models to complex ones
 - Use **sklearn**
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Questions?

Code on GitHub:

<https://github.com/kastnerkyle/SciPy2013>

Bonus: Trends in Machine Learning

- Deep networks
- Generative models
- Unsupervised data from Youtube
- Text-to-speech
- Image object recognition
- Google+ untagged image search

