



MLT1

An Introduction to Machine Learning with Python and scikit-learn

Speaker Name
Job Title
Company/Org Name



Agenda

- Machine Learning in 5 minutes
- Scikit-learn
- Algos & Demos: Linear Regression, Logistic Regression, Decision Trees, K-Means, Principal Component Analysis
- Scikit-learn on Amazon SageMaker
- Resources



Machine Learning in 5 minutes



Artificial Intelligence: design software applications which exhibit human-like behavior, e.g. speech, natural language processing, reasoning or intuition

Machine Learning: using statistical algorithms, teach machines to learn from featurized data without being explicitly programmed

Deep Learning: using neural networks, teach machines to learn from complex data where features cannot be explicitly expressed



Types of Machine Learning

Supervised learning

- Run an algorithm on a labeled data set.
- The model learns how to correctly predict the right answer.
- Regression and classification are examples of supervised learning.

Unsupervised learning

- Run an algorithm on an unlabeled data set.
- The model learns patterns and organizes samples accordingly.
- Clustering and topic modeling are examples of unsupervised learning.

Scikit-learn

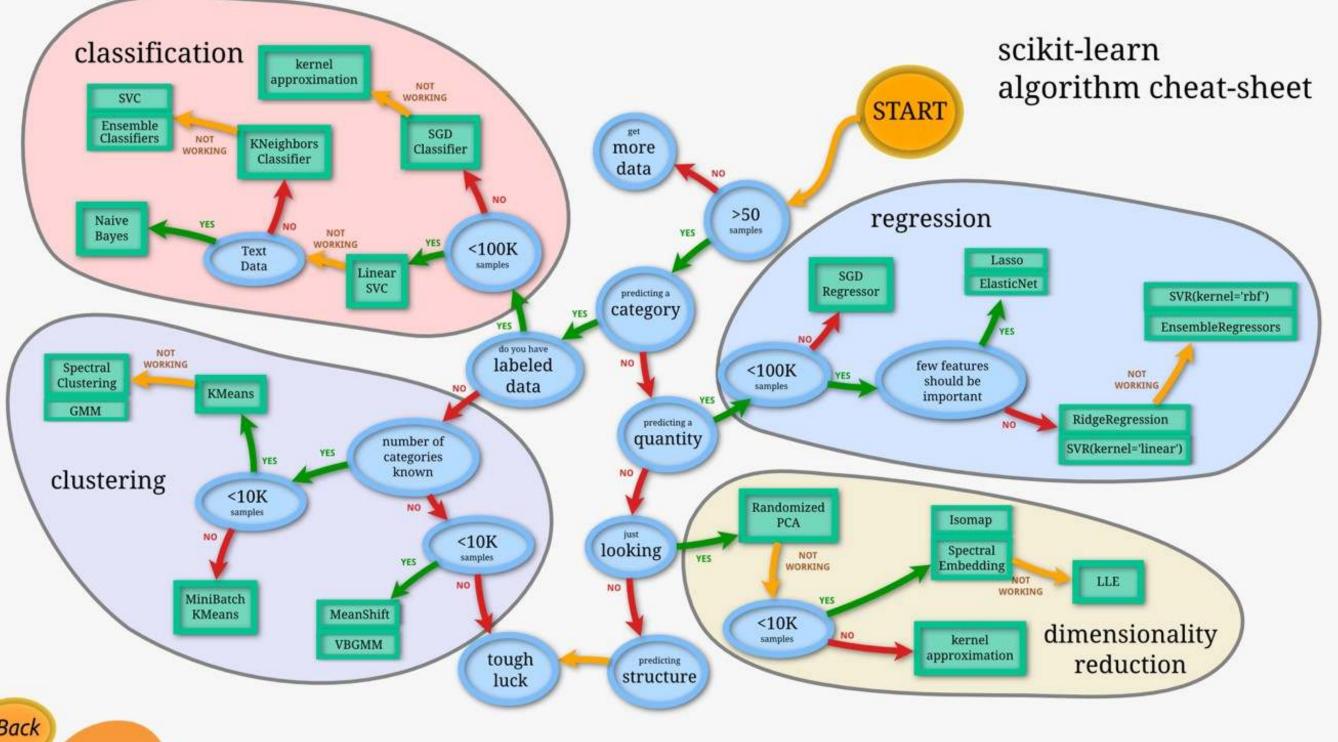


Scikit-learn



- Open Source library in Python released in February 2010
- Built on NumPy, SciPy, and matplotlib
- Simple tools for data analysis and Machine Learning
- Excellent collection of algorithms
- Very good documentation, tons of tutorials
- Limited scalability for data sets that don't fit in RAM
- Not appropriate for Deep Learning (no GPU support)



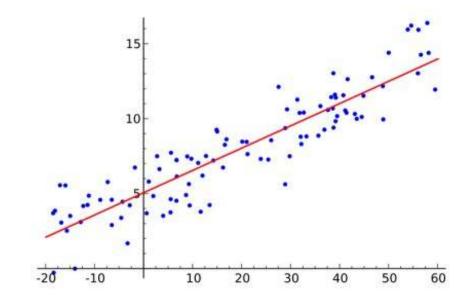


Linear Regression

https://en.wikipedia.org/wiki/Linear regression

- Supervised learning algorithm
- Goal: fit data to a linear function in order to predict numerical values
- Data set: features + target (scalar or scalar vector)
- 1 feature → line, 2 features → plane, etc.





$$y_i = \beta_0 1 + \beta_1 x_{i1} + \cdots + \beta_p x_{ip} + \varepsilon_i$$

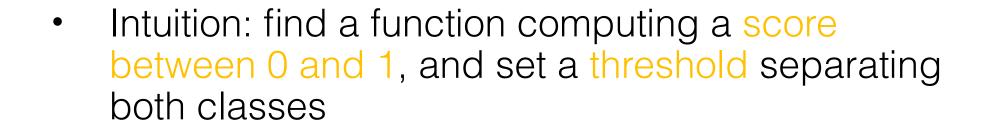
This can also be used for binary classification: a sample is either "above" or "below" the linear function

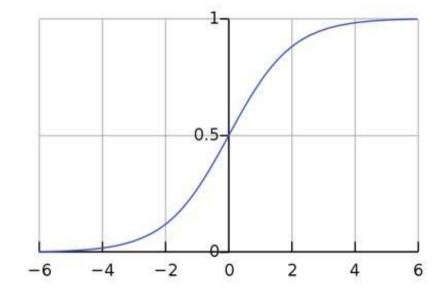


Logistic Regression (1958)

https://en.wikipedia.org/wiki/Logistic regression

- Supervised learning algorithm
- Goal: fit data to a linear function in order to predict the class of a sample
- Data set: features + binary label (yes/no, true/false, etc.)
 - This algorithm can be extended to more than two classes





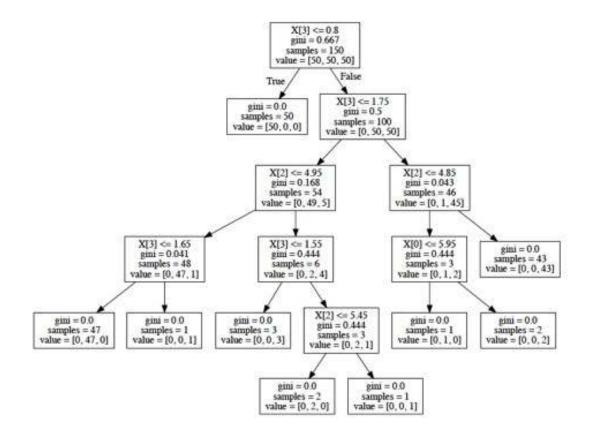
$$p(x)=rac{1}{1+e^{-(eta_0+eta_1x)}}$$



Decision Trees

https://en.wikipedia.org/wiki/Decision_tree

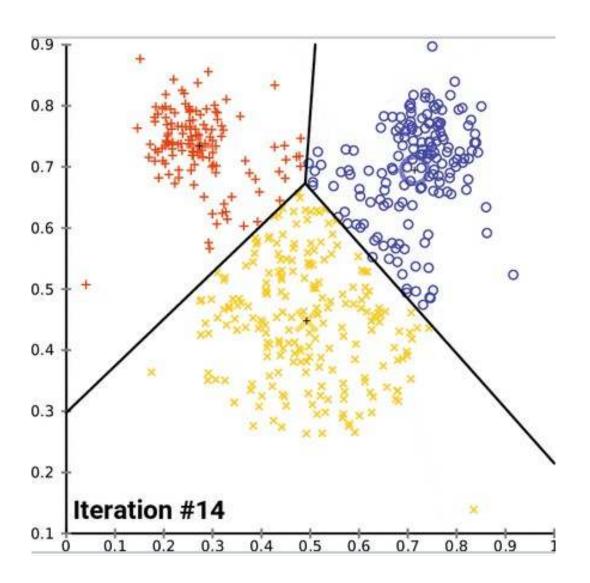
- Supervised learning algorithm
- Goal: build a decision tree for regression or classification
- Data set : features + target value / class
- Intuition: find the "best" feature thresholds to go left or right
- "Easy" to interpret, but prone to overfitting
- Plenty of variants with multiple trees: Random Forests, XGBoost (2016), etc.





K-means (1957) https://en.wikipedia.org/wiki/K-means clustering

- Unsupervised learning algorithm
- Goal: group samples in 'k' clusters
- Data set: features only
- Intuition: find 'k' cluster centers that minimize the "distance" to their respective samples
- This assumes "spherical" clusters of similar "radius": maybe, maybe not!





Principal Component Analysis aka PCA (1901!) https://en.wikipedia.org/wiki/Principal component analysis

- Unsupervised learning algorithm
- Goal: build a new data set with a smaller number of uncorrelated features (aka Dimensionality Reduction) ...keeping as much variance as the number of new features will allow
- Data set: features only
- Sample use cases:
 - Visualize high-dimension data sets in 2D or 3D
 - Remove correlation in high-dimension datasets
 - Preliminary step to building linear models



Demos

- Linear Regression
- Logistic Regression
- Decision Trees
- K-Means
- -PCA
- PCA + Logistic Regression on MNIST

https://gitlab.com/juliensimon/aws --> ML/scikit



Scaling scikit-learn

- Scikit-learn runs on a single machine, loading the full data set in memory
- Scaling options are quite limited <u>http://scikit-learn.org/stable/modules/computing.html</u>
 - Some algorithms can leverage multi-core (joblib)
 - Some algorithms support incremental training
- Amazon SageMaker can help
 - Use ML-optimized multi-core instances (C5)
 - Use pipe mode, i.e. the ability to stream data from Amazon S3



Beyond scikit-learn

Amazon SageMaker

- Train models on fully-managed infrastructure at any scale
- Built-in algorithms (17) for regression, classification, etc.
- Built-in environments for Deep Learning

Apache Spark MLLib

- Available in Amazon Elastic Map Reduce (EMR)
- Distributed processing by design
- Nice collection of Machine Learning algorithms
- Seamless integration with Amazon SageMaker (Scala / PySpark SDK)



Resources

https://ml.aws

https://aws.amazon.com/sagemaker

https://scikit-learn.org

https://www.numpy.org

https://machinelearningmastery.com

https://medium.com/@julsimon

https://gitlab.com/juliensimon/aws



Thank you!

