ML with Spark

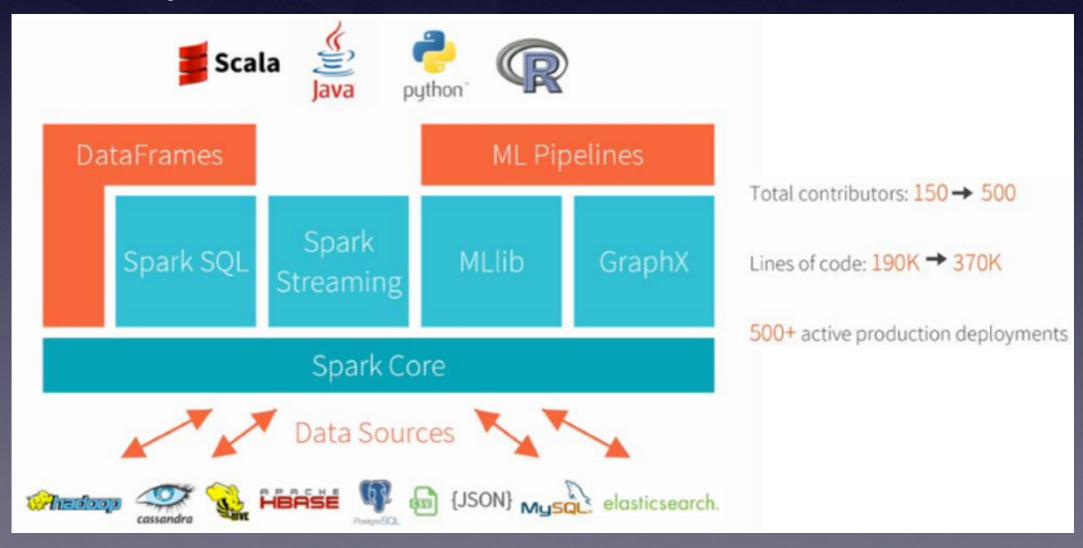
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Spark overview

- General-purpose cluster computing system
- Written in Scala, also have high-level API's for Java, Python and R(pre-alpha)
- Environment Requirements:
- Java 7+, Python 2.6+, R 3.1+, Scala 2.10



MLLib in Spark

Maven dependencies

```
<dependency>
  <groupId>org.apache.spark</groupId>
  <artifactId>spark-mllib_2.11</artifactId>
  <version>1.6.1</version>
</dependency>
```

Two set of libraries:

- 1. org.apache.spark.ml
- 2. org.apache.spark.mllib

Use 2 only if the functionality doesn't exist in 1 !!!

- Classification
 - Logistic regression
 - Decision tree classifier
 - Random forest classifier
 - Gradient-boosted tree classifier
 - Multilayer perceptron classifier
 - One-vs-Rest classifier (a.k.a. One-vs-All)
- Regression

ml

- Linear regression
- · Decision tree regression
- Random forest regression
- Gradient-boosted tree regression
- Survival regression
- Decision trees
 - · Inputs and Outputs
 - Input Columns
 - Output Columns
- · Tree Ensembles
 - Random Forests
 - Inputs and Outputs
 - Input Columns
 - Output Columns (Predictions)
 - · Gradient-Boosted Trees (GBTs)

Problem Type	Supported Methods
Binary Classification	linear SVMs, logistic regression, decision trees, random forests, gradient-boosted trees, naive Bayes
Multiclass Classification	logistic regression, decision trees, random forests, naive Bayes
Regression	linear least squares, Lasso, ridge regression, decision trees, random forests, gradient-boosted trees, isotonic regression

MLLib in Spark

Spark Concept Recall:

- SparkConf
- JavaSparkContext

Data Types:

- DataFrame
- JavaRDD<T> Typical implementation for T is LabeledPoint or Vectors

SQLContext

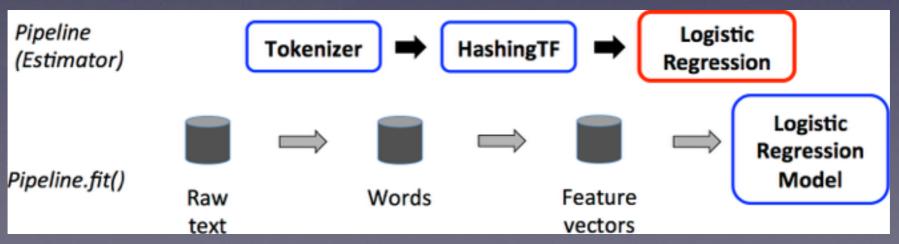
- · Create DF.
- Execute SQL-format query over DF.

PipeLine

- · Aggregation of stages, stage can be either learning algorithm or any annotat
- Pipeline pipeline = new Pipeline().setStages(new PipelineStage[]{Ir});

Estimator

· Can be either a single specific learning alg. or Pipeline



MLLib in Spark

Pros:

- 1. Works well with large dataset.
- 2. Several good optimization libraries implemented.
- 3. Implemented full pipelines concept instead of discrete classifiers.
- 4. Have all training, prediction and cv pipelines.

Cons:

- 1. Pure ML library means you need to write adapter for application.
- 2. Only some of the core ML algorithms implemented currently.
- 3. Learning Curves for Developer.

Caveat:

Parallel data processing is ok.

Parallel core training process won't work for iterative ml algorithms

Problem Definition

Dataset: Student with 2 exam scores ranging from [0 - 100]

Predict: If grant admission for a potential student.

Dataset Size: 100

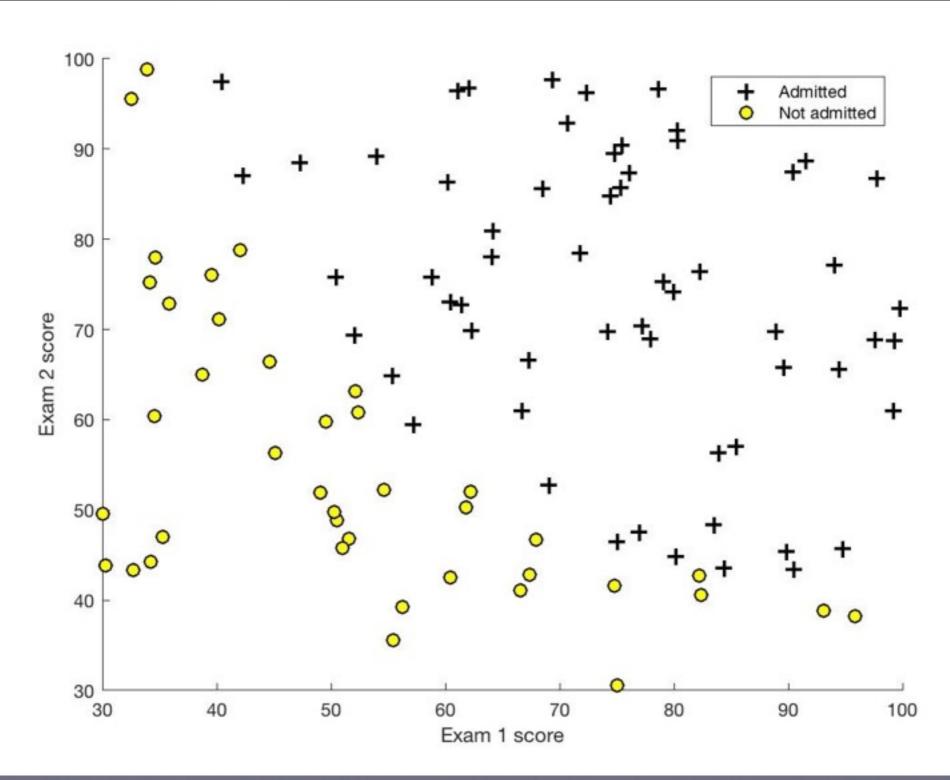
Binary Category: 0 - Decline, 1 - Admit

Examples:

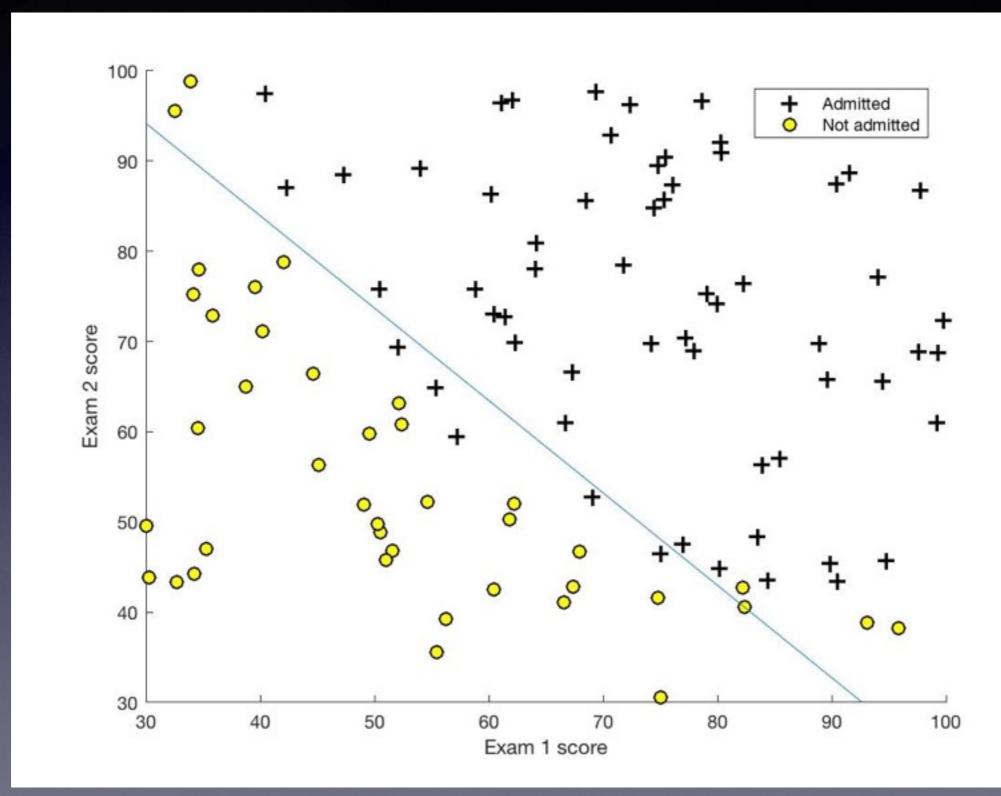
	Exam1,	Exam2,	Label
Student 1:	34.62365962451697,	78.0246928153624,	0
Student 2:	30.28671076822607,	43.89499752400101,	0
Student 3:	35.84740876993872,	72.90219802708364,	0

. . .

Visualize Dataset



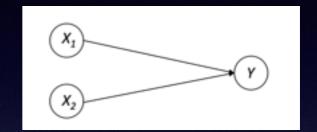
Visualize Dataset

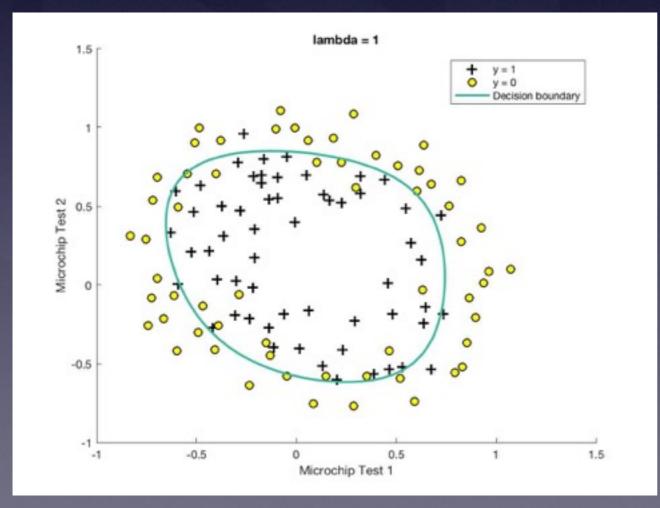


Logistic Regression (LR)

Log Linear model This is NOT a regression

- Discriminative model P(y|x)
- LR is a special kind (binomial) of MaxEnt
- Features do not necessary to scale, but encouraged to do so.
- Not necessary for linear boundary.
- The plot uses polynomial feats, x^2, x1*x2, etc.





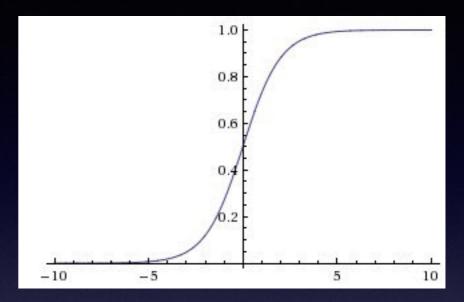
Non-linear decision boundary

Logistic Regression (LR)

Term Def:

- 1. Total Training Sample Size: m
- 2. Total features: n
- 3. i-th training sample: x_(i)
- 4. j-th dimension of i-th training sample: x_(i)_j
- 5. i-th training sample label: y_(i)
- 6. Weight vector θ (n dimensional)
- 7. $h(\theta)$ is the hypothesis function
- 8. sigmoid function == logistic function

$$sigmoid(z) = \frac{1}{1 + e^{-z}}$$
 $h(\theta) = sigmoid(\sum_{j=0}^{n} x_j^{(i)} \cdot \theta_j)$



Sigmoid Function

Training:

$$J(\theta) = \frac{1}{m} \sum_{i=1}^{m} \text{Cost}(h_{\theta}(x^{(i)}), y^{(i)})$$
$$= -\frac{1}{m} \left[\sum_{i=1}^{m} y^{(i)} \log h_{\theta}(x^{(i)}) + (1 - y^{(i)}) \log (1 - h_{\theta}(x^{(i)})) \right]$$

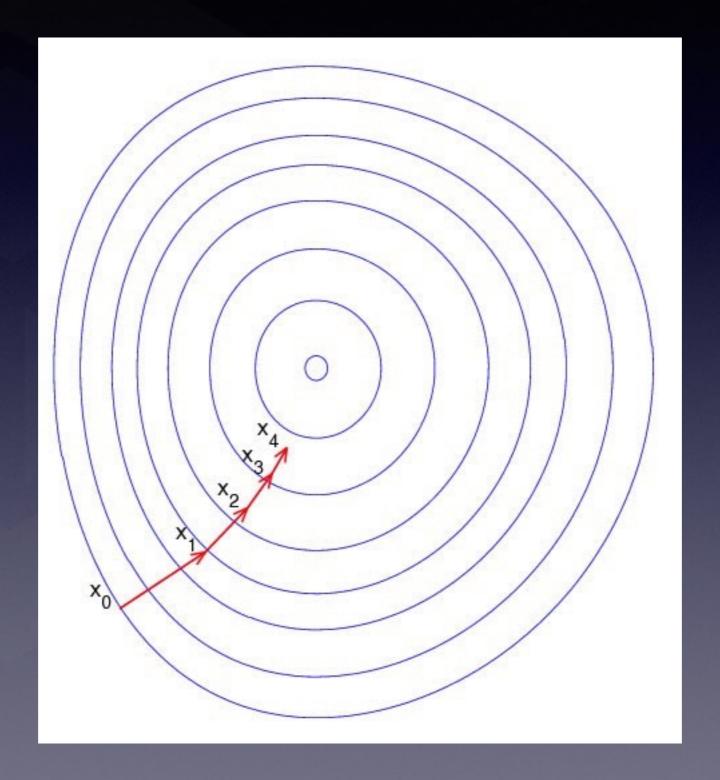
Goal: Optimize θ vector to minimize $J(\theta)$

Optimizer: Gradient Descent/SGD/LBFGS

Predict:

$$y = \begin{cases} 0 & h(\theta) < 0.5 \\ 1 & otherwise \end{cases}$$

Gradient Descent



Coding Time

Weights learned with LR+GD in Matlab by fminunc:

-24.932775, 0.204406, 0.199616 (First is intercept)

Code snippet also available at my github:

https://github.com/maochen/Playground/blob/master/playground/src/main/java/org/maochen/playground/spark/ClassificationExample.java

Dataset:

https://github.com/maochen/Playground/blob/master/playground/src/main/resources/spark/student_exam_data.txt