# Apache Pig for Data Science

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#### Introduction

- I'm a Principal Architect at Hortonworks
- I work primarily doing Data Science in the Hadoop Ecosystem
- Prior to this, I've spent my time and had a lot of fun
  - Doing data mining on medical data at Explorys using the Hadoop ecosystem
  - Doing signal processing on seismic data at Ion Geophysical using MapReduce
  - Being a graduate student in the Math department at Texas A&M in algorithmic complexity theory
- I'm going to talk about Apache Pig's role for doing scalable data science.

Hadoop is a distributed storage and processing system

• Scalable - Efficiently store and process data

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- Economical Use commodity hardware and open source software
- Not a one-trick-pony Not just MapReduce anymore

## Apache Hadoop: Who is using it?





The New York Times



amazon.com











Pig is a high level scripting language for operating on large datasets inside Hadoop

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- Familiar relational primitives available
- Extensible via User Defined Functions and Loaders for customized data processing and formats

## Apache Pig: An Familiar Example

## Understanding Data

"80% of the work in any data project is in cleaning the data."

— D.J. Patel in *Data Jujitsu* 

## Understanding Data

A core pre-requisite to analyzing data is understanding data's shape and distribution. This requires (among other things):

- · Computing distribution statistics on data
- Sampling data

An Apache Incubating project called datafu<sup>1</sup> provides some of these tooling in the form of Pig UDFs:

• Computing quantiles of data

<sup>&</sup>lt;sup>1</sup>http://datafu.incubator.apache.org/

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  - Simple Random Sample
  - Reservoir sampling
  - Weighted sampling without replacement
  - Random Sample with replacement

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## Case Study: Bootstrapping

Bootstrapping is a resampling technique which is intended to measure accuracy of sample estimates. It does this by measuring an estimator (such as mean) across a set of random samples with replacement from an original (possibly large) dataset.

## Case Study: Bootstrapping

Datafu provides two tools which can be used together to provide that random sample with replacement:

- SimpleRandomSampleWithReplacementVote Ranks multiple candidates for each position in a sample
- SimpleRandomSampleWithReplacementElect Chooses, for each position in the sample, the candidate with the lowest score

The datafu docs provide an example<sup>2</sup> of generating a boostrap of the mean estimator.

<sup>&</sup>lt;sup>2</sup>http://datafu.incubator.apache.org/docs/datafu/guide/sampling.html

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- Unsupervised Learning
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  - Market Basket Analysis
- Supervised Learning
  - Classification
  - Regression
  - Recommendation

## Building Machine Learning Models with Pig

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- Build one large model on all (or almost all) of the data
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Pig can assist in intelligently sampling down the large data into a training set. You can then use your favorite ML algorithm (which can be run on the JVM) to generate a machine learning model.

# Applying Models with Pig

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- Write a UDF in Java or another JVM language which can apply the model to a data point
- Call the UDF from a pig script to distribute the application of the model across your data in parallel

## What is Natural Language Processing?

- Natural language processing is the field of Computer Science, Linguistics & Math that covers computer understanding and manipulation of human language.
  - o Historically, linguists hand-coded rules to accomplish much analysis
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  - o Historically, linguists hand-coded rules to accomplish much analysis
  - Most modern approaches involves using Machine Learning
- Mature field with many useful libraries on the JVM
  - Apache OpenNLP
  - Stanford CoreNLP
  - MALLET

#### Natural Language Processing with Large Data

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  - Social media content analysis

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- Sometimes high-volume, complex analysis
  - Search Engines
  - Social media content analysis
- Typically many small-data problems in parallel
  - Often requires only the context of a single document
  - Ideal for encapsulating as Pig UDFs

## Natural Language Processing: Demo

- Stanford CoreNLP integrated the work of Richard Socher, et al [2] using recursive deep neural networks to predict sentiment of movie reviews.
- There is a large set of IMDB movie reviews used to analyze sentiment analysis [1].
- Let's look at how to encapsulate this into a Pig UDF and run on some movie review data.

#### Results

- Executing on a sample of size 1022 Positive and Negative documents.
- Overall Accuracy of 77.2%

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Actua	ı

Predicted

	Positive	Negative
Positive	367	114
Negative	119	422
Total	486	536

Total

#### Questions

Thanks for your attention! Questions?

- Code & scripts for this talk available on my github presentation page.<sup>3</sup>
- Find me at http://caseystella.com
- Twitter handle: @casey\_stella
- Email address: cstella@hortonworks.com

<sup>&</sup>lt;sup>3</sup>http://github.com/cestella/presentations/

## Bibliography

- [1] Andrew L. Maas, Raymond E. Daly, Peter T. Pham, Dan Huang, Andrew Y. Ng, and Christopher Potts. Learning word vectors for sentiment analysis. In *Proceedings of the 49th Annual Meeting of the Association for Computational Linguistics: Human Language Technologies*, pages 142–150, Portland, Oregon, USA, June 2011. Association for Computational Linguistics.
- [2] Richard Socher, Alex Perelygin, Jean Wu, Jason Chuang, Christopher D. Manning, Andrew Y. Ng, and Christopher Potts. Recursive deep models for semantic compositionality over a sentiment treebank. In *Proceedings of the 2013 Conference on Empirical Methods in Natural Language Processing*, pages 1631–1642, Stroudsburg, PA, October 2013. Association for Computational Linguistics.