edatabricks

Interactive Visual Data Exploration with Spark in Databricks Cloud

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About Databricks

Founded by creators of Apache Spark

Offers Spark as a service in the cloud

Dedicated to open source Spark

- Largest organization contributing to Apache Spark >
- > Drive the roadmap





Databricks Cloud

Databricks Workspace



Databricks Platform

>Notebooks >Dashboards >Job launcher

>Latest version Configured / Optimized

Start clusters in seconds >Dynamically scale up & down





Fast & General distributed computing engine: batch, streaming, iterative

Capable of handling petabytes of data

Even faster by caching data in-memory

Versatile programming interfaces









Spark: Mixing SQL with Python/Scala

// Perform transformations words = rdd.flatMap(lambda r: r.text.split())

// Collect sample of data in driver machine sampled words = words.sample(fraction = 0.001)

// Query an existing table and get results back as Schema RDD rdd = hiveContext.sql("select article, text from wikipedia")





Databricks Platform

Clusters

Name	Memory	State	Nodes	Notebooks	Dashboard Cluster	Alter
Sales Analysis	102 GB	Running	Spark Master Worker 0 Worker 1 Worker 2	Sales Analysis Tom's Analysis	Make Dashboard Cluster	 Restart Remove
Default Cluster	254 GB	Running	Spark Master Worker 0 Worker 1 Worker 2 Worker 3 Worker 3 Worker 4 Worker 5 Worker 6 Worker 7 Worker 8	Sales	Current Dashboard Cluster	 Restart Remove
Marketing Pipeline	127 GB	Running	Spark Master Worker 0 Worker 1 Worker 2 Worker 3	MarketingData	 Make Dashboard Cluster 	 Restart Remove
						• Add

Start clusters in seconds

Zero-cost management

Dynamically scale up and down





Databricks Workspace

Notebooks

- > SQL
- > Python
- > Scala

Dashboards

Job Launcher





Notebooks

Exploring Diamonds

Command took 0.012s



Command took 0.002s.

Scatter plot matrix

We can use the scatter plot matrix to quickly explore correlation between all features Command took 0.003s

» select * from diamonds where rand() < 0.01</pre>



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Plot Options...

Command took 4.207s.

Supports Python, Scala, SQL

Interactive commands and plots

On-line collaboration





Dashboards

Diamonds Dashboard

Provide a quick overview of some pertinent elements of the data that we've seen so far.

Note - Data refreshed each night at midnight with the latest sales data.



Showing data results filtered by cut





Effect of cut on price controlled for color



WYSIWYG Builder

Interactive jobs

On-click publishing

Exporting from notebooks



Job Launcher

Elastic Jobs <u>New Job</u>

Job Name 🔻	Recent Runs	Active		
Market Basket Analysis	05/26/2014 06/02/2014 06/09/2014 	Today at 8:57 PM Cluster: Default Cluster Actions Image: Actions I	Minimize	Remove
Sales Dashboard ETL	Today at 5:00 PM Today at 6:00 PM Today at 7:00 PM Today at 8:00 PM Today at 9:00 PM Today at 10:00 PM Today at 11:00 PM	Tomorrow at 12:00 AM Cluster: Default Cluster	Edit	Remove
Fraud Model Training	06/09/2014 Last Tuesday at 1:00 AM Last Wednesday at 1:00 AM Last Thursday at 1:00 AM Last Friday at 1:00 AM Last Saturday at 1:00 AM Yesterday at 1:00 AM	Actions 🕩 databricks/ml/training.jar : Triggers 🚭 Daily: 1am	Minimize	Remove

Runs arbitrary Spark jobs programmatically













"Visualization is critical to data analysis."

But we often skip exploratory visualization with large data

William S. Cleveland





1. Interactivity

with large data is challenging

2. Visual medium



cannot accommodate as many pixels as data points



1. Interactivity In-memory computation High parallelism

Solutions





Reducing interaction latency with Spark

- 1. In-memory computation
 - > Significantly reduces latency
- 2. High parallelism

 - > Click a button to increase cluster size in Databricks Cloud

> Get more executors with Mesos or Yarn: a challenge in itself



Versatile programming interface

Data visualization is very much like programming.

- Point and click doesn't really cut it
- > Requires an API (grammar): ggplot, matplotlib, bokeh, etc.

Spark has SQL, Scala, Python, Java and (experimental) R API

Libraries for distributed statistics and machine learning



1. Interactivity In-memory computation High parallelism

2. Visual medium In-browser collaborative notebooks Summarizing, Sampling and Modeling

Solutions



More data points than pixels

Short answer: no

- Long answer: > Summarize & visualize

Can we visualize 200GB of multidimensional data?

Sample & visualize > Model & visualize



Summarize and visualize

Extensively used by BI tools

- > Aggregation
- > Pivoting

Most data scientists' nightly jobs summarize data



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Sample and visualize

Sometimes we need to visualize (feel) individual data points

Sampling is extensively used in statistics

Spark offers native support for:

- > Approximate and exact sampling
- > Approximate and exact stratified sampling

Approximate sampling is faster and is good enough in most cases









Model and visualize

MLLib supports a large (and growing) set of distributed algorithms

- Clustering: k-means
- Classification and regression: LM, DT, NB
- Dimensionality reduction: SVD, PCA
- Collaborative filtering: ALS
- Correlation, hypothesis testing

















Summary

With new big data tools we can resume interactive visual exploration of data

Using Spark we can manipulate large data in seconds

- > Cache data in memory
- > Increase parallelism

To visualize millions of data points we can

- > Summarize
- > Sample
- Models







Databricks Cloud Apache Spark Matplotlib Python ggplot 1)3

databricks.com

spark.apache.org

matplotlib.org

ggplot.yhathq.com

d3js.org





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