



# Exploratory Data Analysis with Apache Spark

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# About Databricks

Founded by creators of Apache Spark from UC Berkeley

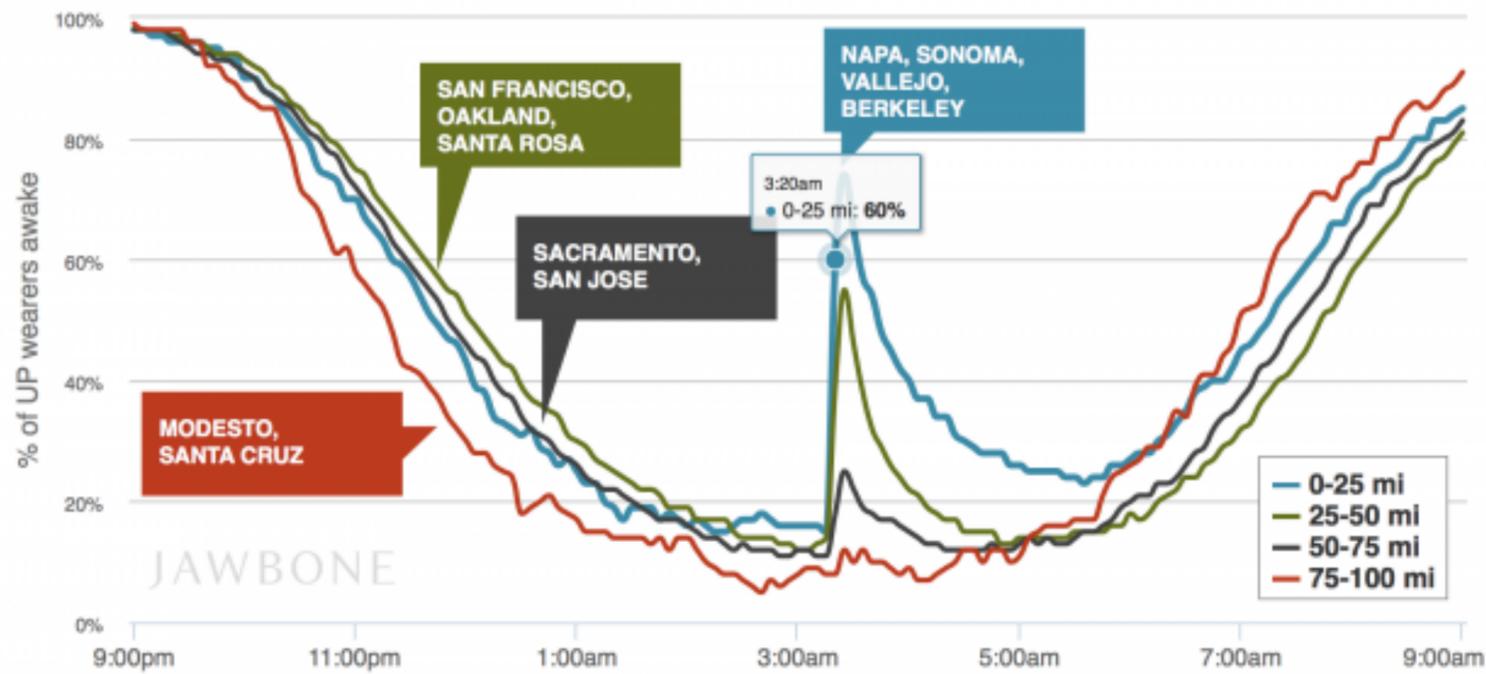
We are dedicated to open source Spark

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

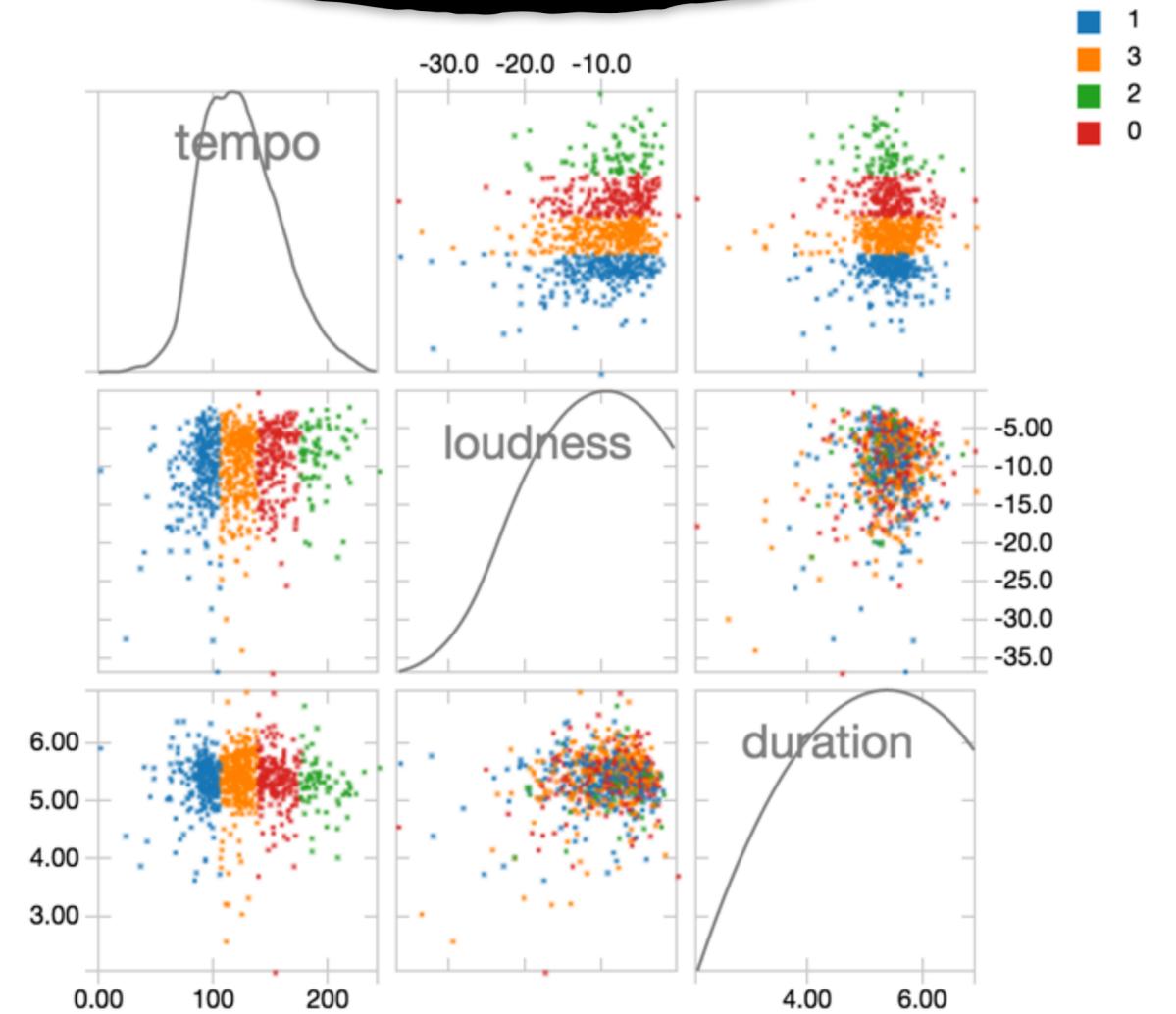
We offer Spark as a service in the cloud

# Expository vs.

# Exploratory



We wish all the people in the Bay Area who were affected by the earthquake a speedy recovery and a good night's sleep.



Large data

“Visualization is critical to data analysis.”

William S. Cleveland

But we often skip exploratory visualization with large data

# Challenges

## 1. Interactivity

with large data is challenging

## 2. Visual medium

cannot accommodate as many pixels as data points

# Solutions

## 1. Interactivity

In-memory computation

High parallelism



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: Versatile programming interface

Data visualization is 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

# Spark: Mixing SQL with Python/Scala

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

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

// Sample data and download to driver machine
sampled_words = words.sample(fraction = 0.001).collect()
```

# Reducing interaction latency with Spark

## 1. In-memory computation

- › Significantly reduces latency

## 2. High parallelism

- › Get more executors with Mesos or Yarn: a challenge in itself
- › Click a button to increase cluster size in Databricks Cloud

# Solutions

## 1. Interactivity

In-memory computation

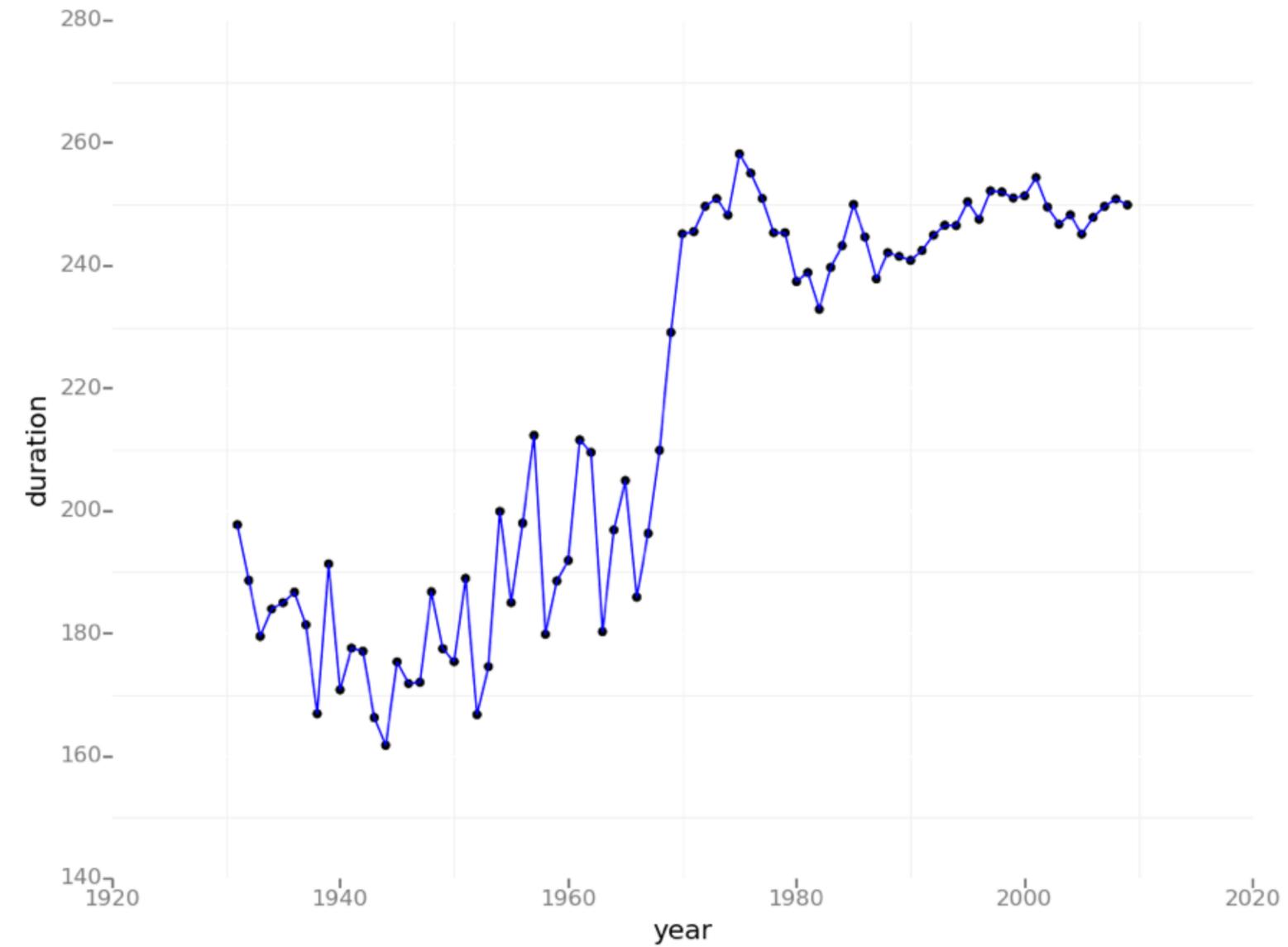
High parallelism

## 2. Visual medium

In-browser collaborative notebooks

Summarizing, Sampling and Modeling

# Summarize and visualize



# Sample and visualize

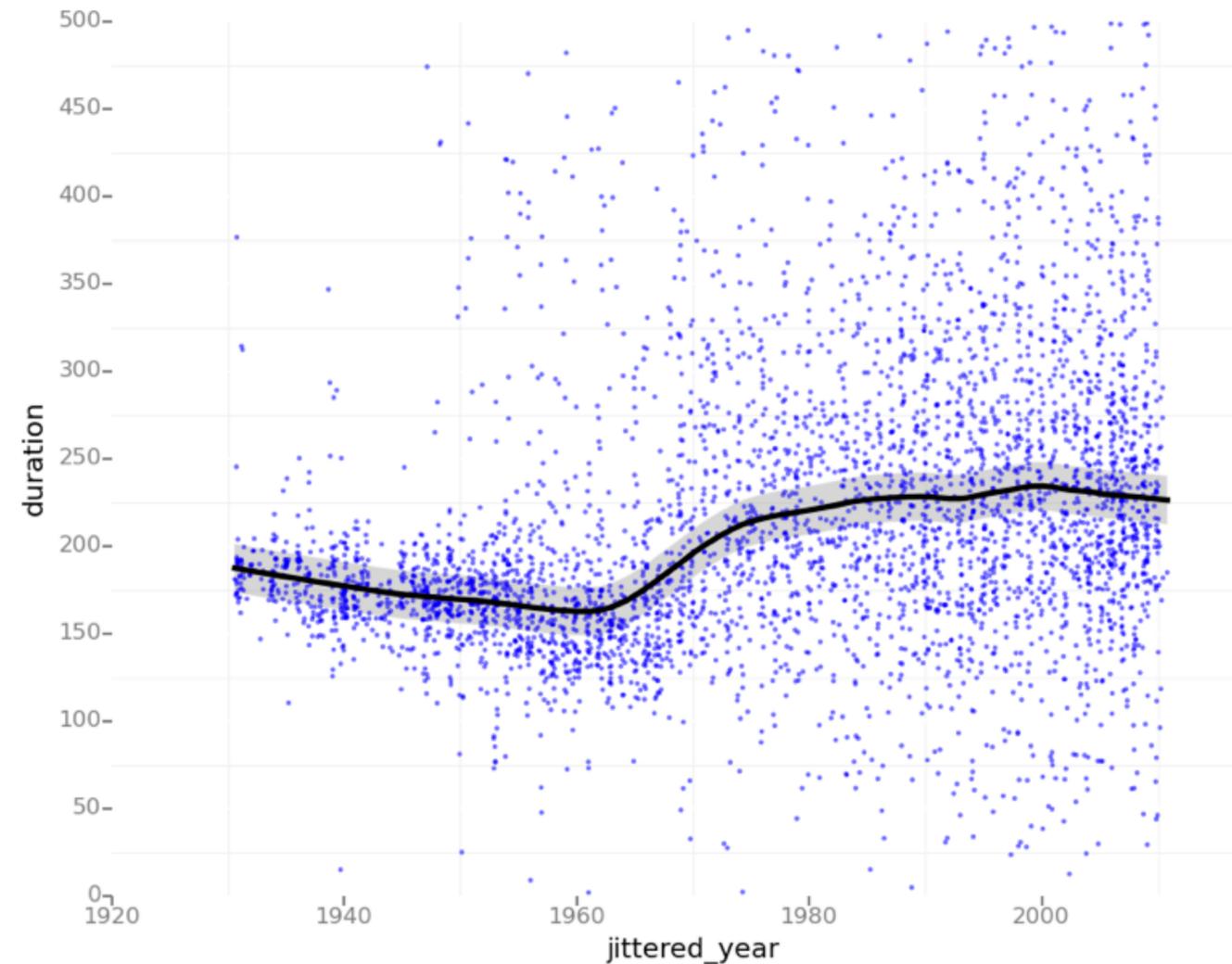
Sometimes we need to visualize (feel) individual data points

Sampling is extensively used in statistics

Spark offers native support for:

- › Approximate and exact sampling
- › Approximate and exact stratified sampling

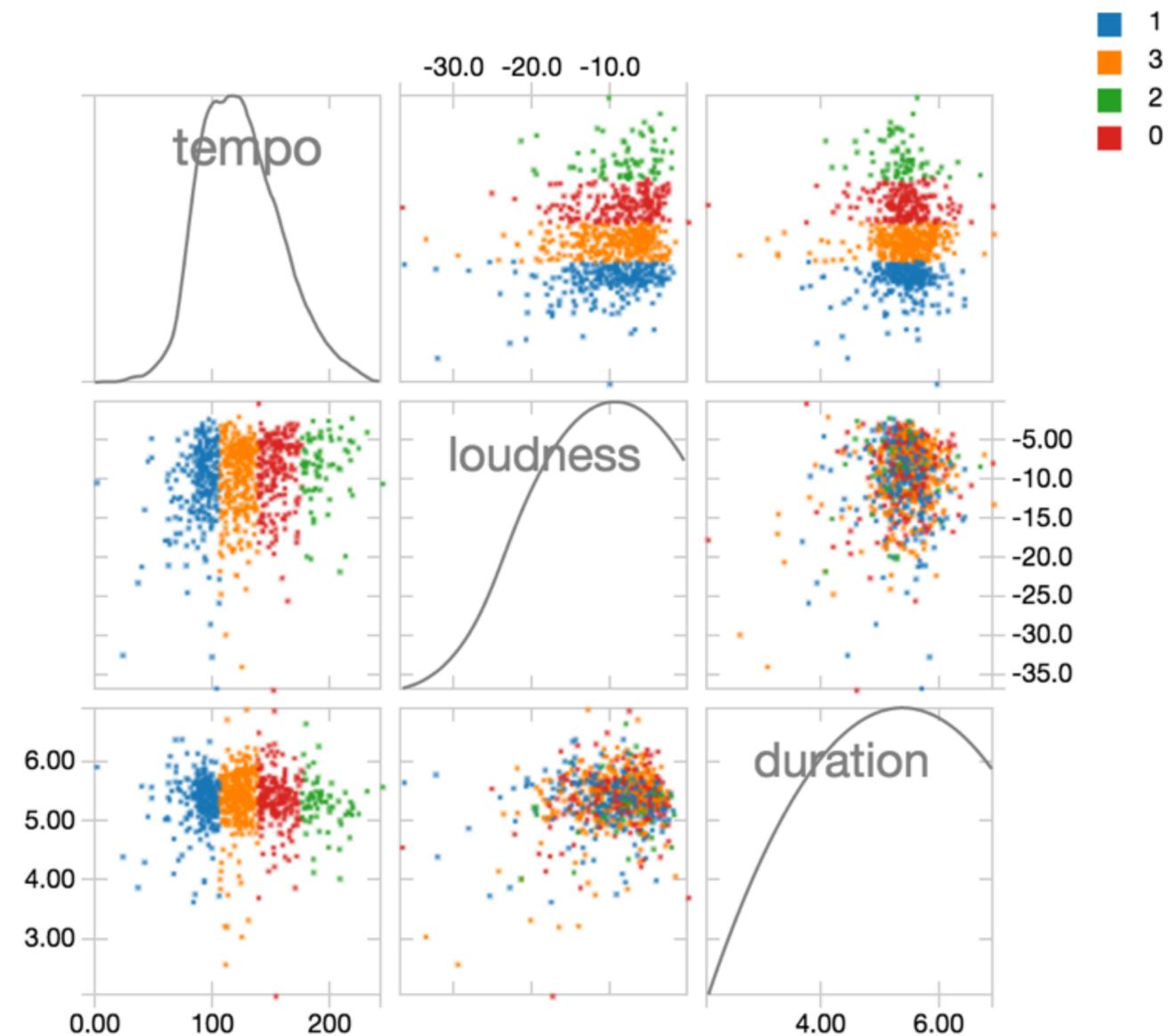
Approximate sampling is faster and is good enough in most cases



# Model and visualize

MMLib 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



# About Databricks Cloud

Databricks Workspace



Databricks Platform

- › Notebooks
- › Dashboards
- › Job launcher

- › Start clusters in seconds
- › Dynamically scale up & down

# Demo

# We saw that

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

[databricks.com](https://databricks.com)

Apache Spark

[spark.apache.org](https://spark.apache.org)

Matplotlib

[matplotlib.org](https://matplotlib.org)

Python ggplot

[ggplot.yhathq.com](https://ggplot.yhathq.com)

D3

[d3js.org](https://d3js.org)

