

# Big Data Analytics Final Report

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## I. INTRODUCTION

The Healthcare and Medical industry is one of the biggest industries in the US and even the world, with estimated net transactions of \$3.24 trillion. American consumers spend hundreds of thousands of dollars in insurance and medical costs. In such a huge industry, even small percentage increases in efficiency can result in millions of dollars in increased revenue and higher market share. It is no surprise that companies form collaborations and have huge data analytic departments to maximize their profits. From a marketing point of view, Pharmaceuticals usually give some sort of monetary benefits to healthcare providers like hospitals and doctors so that these providers can use the services of these companies. For example, the maker of a particular brand of medicinal pills may pay a doctor to use and vouch for its medicines, rather than some generic or competing brand. Benefits may also be in the form of travel reimbursements, speaking fees, research grants, or even ownership interests in the company. Such payments by drug and device companies are so commonplace, that the payments total 7.52 Billion dollars, from 12 million payments per year. We try to analyse some of this data to capture and see trends in payments due to geographic or demographic factors.

## II. SYSTEM OVERVIEW

The dataset that we are using is from openpayments. Open Payments is a federal program, required by the Affordable Care Act, that collects information about the payments drug and device companies make to physicians and teaching hospitals for things like travel, research, gifts, speaking fees, and meals. It also includes ownership interests that physicians or their immediate family members have in these companies. This data is then made available to the public each year on this website

**. For patients, consumers, and the public,** Open Payments can be used to learn about the relationships between physicians and applicable manufacturers and GPOs. We encourage patients to discuss these relationships with their healthcare providers.

**For physicians and teaching hospital representatives,** reviewing the data reported about you in the Open Payments system can ensure that this information is accurate. You can also:

- Use the information reported about you to plan for questions from patients
- Healthcare comprises 17.5% of US GDP.
- 19 major drug companies had at least one board member who also held leadership role at at least one academic medical center.

## III. ALGORITHM

MLlib fits into Spark's APIs and interoperates with NumPy in Python (as of Spark 0.9) and R libraries (as of Spark 1.5). You can use any Hadoop data source (e.g. HDFS, HBase, or local files), making it easy to plug into Hadoop workflows. ML algorithms include:

- Classification: logistic regression, naive Bayes,...
- Regression: generalized linear regression, survival regression,...
- Decision trees, random forests, and gradient-boosted trees
- Recommendation: alternating least squares (ALS)
- Clustering: K-means, Gaussian mixtures (GMMs),...
- Topic modeling: latent Dirichlet allocation (LDA)
- Frequent itemsets, association rules, and sequential pattern mining

## IV. SOFTWARE PACKAGE DESCRIPTION

Spark is a Data processing engine/framework developed by the Apache foundation. It is an improvement upon Hadoop because of its ability to process data in memory, rather than copying intermediate results to HDFS like Hadoop does. Programs can potentially be run up to 100x faster in Spark when compared to Hadoop. It allows easy use by being compatible with common programming languages like R, Python, Scala and Java. It has loads of libraries that allow us to perform operations in SQL and Machine Learning among others using MLlib.

**MLlib:** MLlib is Spark's Machine learning library. It is extremely scalable, sophisticated and versatile. It provides functions for various operations like Classification, Regressions and Clustering using multiple methods like Decision trees, Gaussian Mixtures, and Latent Dirichlet Allocations.

2. Docker: Docker is a software containerization platform, much like Virtual Machine hosts like VMware or VirtualBox. It allows for wrapping software in a complete filesystem inclusive of system tools and libraries. The end result is that Docker appears to users as its own virtual machine, when it is in fact a container running on part of a remote server.

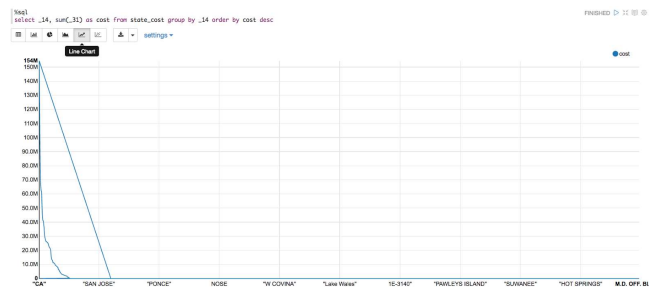
```
ssh@sshboxess:~$ sudo docker ps
CONTAINER ID        IMAGE               COMMAND             CREATED             STATUS              PORTS              NAMES
ssh@sshboxess:~$ sudo docker exec -it jolly_jupyter bash
root@jolly_jupyter:/#
Round 15 items
jupyter-viz      1 root root        15389 2018-12-07 16:18 LICENSE
jupyter-viz      1 root root        355 2018-12-07 16:18 NOTICE
jupyter-viz      1 root root        1324 2018-12-07 16:18 README.md
jupyter-viz      1 root root        4096 2018-12-15 00:15 bigdata_project
jupyter-viz      1 root root        4096 2018-12-07 16:18 bin
jupyter-viz      1 root root        4096 2018-12-15 02:17 conf
jupyter-viz      1 root root        4096 2018-12-07 16:46 interpreter
jupyter-viz      1 root root        12268 2018-12-07 16:18 lib
jupyter-viz      1 root root        4096 2018-12-07 16:18 licenses
jupyter-viz      1 root root        4096 2018-12-15 02:18 logs
jupyter-viz      1 root root        4096 2018-12-15 02:17 notebook
jupyter-viz      1 root root        4096 2018-12-15 00:45 run
jupyter-viz      1 root root        4096 2018-12-15 02:48 zeppelin-notebook
jupyter-viz      1 root root        4096 2018-12-15 02:15 zeppelin
jupyter-viz      1 root root        2427766 2018-12-07 16:13 zeppelin-web-0.7.0-SNAPSHOT-war
jupyter-viz      1 root root        1418956207 2018-12-07 16:13 zeppelin-hadoop-0.15.0-18101817
Round 4 items
jupyter-viz      1 root root        5065202714 2018-12-15 00:16 /bigdata/OF_OTL_OML_PSV2015_F06302016.csv
jupyter-viz      1 root root        1762779 2018-12-15 00:26 /bigdata/OF_OTL_OML_PSV2015_F06302016.csv
jupyter-viz      1 root root        622464231 2018-12-15 00:16 /bigdata/OF_OTL_PSV2015_F06302016.csv
jupyter-viz      1 root root        615 2018-12-15 00:16 /bigdata/OF_OTL_PSV2015_F06302016.txt
jupyter-viz      1 root root        1418956207 2018-12-07 16:13 zeppelin-hadoop-0.15.0-18101817
```

3. Zeppelin: Zeppelin is a programming notebook much like Sagemath or Jupyter. However, there are some key differences. The main difference is that Zeppelin being an Apache project supports Big Data languages like Hadoop, Pig etc. It also allows for Spark accessibility using API calls in python etc. Another useful feature is the ability to code in different languages in different sections, which is not possible in Jupyter. Having a notebook allows us to both code and visualize results on the fly in the same software bundle, rather than having to print, take screenshots, save and collate among other things.

V. EXPERIMENT RESULTS

We first did a count of the transactions, grouping by state. We saw clear signs of higher number of transactions in more populous state by demographic, and not by geographical area. For example California and New York are the highest transacted states.

state	count
CA	4464
TX	154,459,422.0
NY	154,459,384.00
IL	72,289,198.01
WA	62,162,264.60
MA	61,198,894.28
PA	47,241,778.42
NC	41,438,861.76
CO	40,543,986.80
MD	36,812,887.85
MI	35,492,240.24
OR	28,127,251.22
AK	25,487,827
HI	25,258,884.21
WY	25,162,882.82
SD	25,086,882.81
NE	24,811,412.49
ND	23,154,816.46
VT	22,287,486.44
RI	21,884,782.73
DE	20,215,385.17
WV	19,229,495.17
MS	19,442,723.84
LA	19,800,020.85
SC	11,232,116.69
UT	11,172,263.84
AL	10,879,863.39
OK	10,348,285.29
MO	9,268,249.87
IA	9,142,885.89



We can Also see that not only that are the popular states leading in terms of pure number of transactions, they also have the most expensive transactions too. This can be attributed to the fact that California and New York have financial and technological hubs like New York City and San Francisco among others which contain headquarters of the biggest pharmaceutical companies on earth.



of your algorithm. Show how did you evaluate the performance of your algorithm.

VI. CONCLUSION

Below is a rough summary of the work done as part of the project. Initial Literature review and plan formulation. Initial testing of Spark Mllib-cleaning and data organizing done in Python. Data writing to HDFS using hadoop. Analysis of data in Spark Mllib Visualization using matplotlib in Zeppelin.

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APPENDIX

## REFERENCES

- [1] <https://openpaymentsdata.cms.gov/>
- [2] <https://zeppelin.apache.org/>
- [3] <http://spark.apache.org/>.