

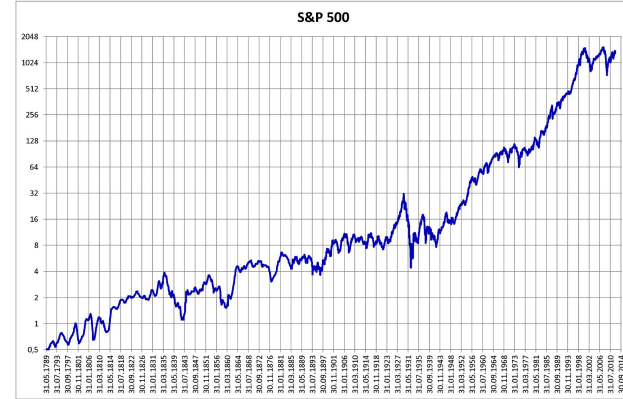


DS8003 Final Project: Stock market analysis

Adam Azoulay  
Hee Kyoung Nam  
Peter Yu

# The Problem

- Stock market → Highly researched complicated data ecosystem
- Data → constantly changing and we want to be up to date with market
- Existing tools/sites exist → Costly, doesn't meet our needs



# Work Distribution

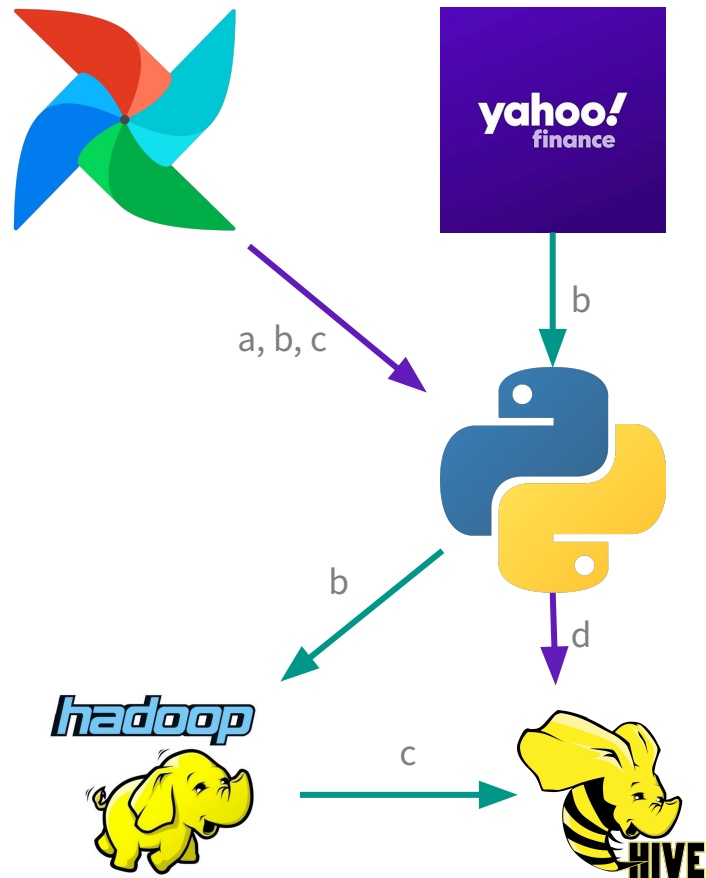
- Adam
  - Infrastructure, ETL (Airflow → Hadoop → Hive → Kibana + Spark)
- Hee
  - Data analysis using Spark and Spark SQL
- Peter
  - Data visualization with Kibana

# Data Sources

- [NASDAQ Stock Screener](#)
  - Ticker name and sector information, along with volumes and market cap data
- [Yahoo Finance API](#)
  - Chart data to track price movements (/v8/finance/chart)
  - Options information (/v7/finance/options)
  - Recent news (/v1/finance/search)
  - Sustainability report (scraped from yahoo sustainability page)
  - Recommendation information

# Processing & Tools

- The data from the endpoints goes through a few steps before it end up in our Hive DB
  - a. Airflow runs a python script to download the desired data from Yahoo
  - b. From memory we format the data and write it to our hdfs in a staging folder
  - c. We execute a command from Airflow which loads the staged file into a hive table, moving the data to the final location
  - d. Airflow executes any processing in the form of HiveQL queries to refresh our analysis tables
  - e. If desired, we can have Airflow notify us via message or email about the status and a summary of the results



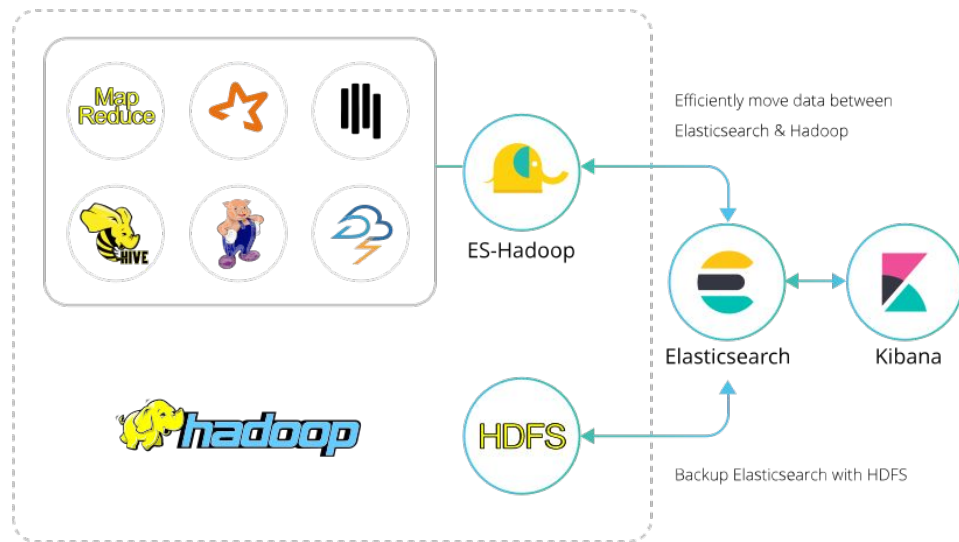
# Processing & Tools

- The data in hdfs is organized by hive, and will sit in staging if the pipeline fails which is nice to have
- When we load into Hive, we make sure to only insert new rows, so we prevent duplicates!
- At the end of the ETL we can query all of our updated data in HiveQL!

Code available [here](#)

# Processing & Tools

- Having loaded the data into Hive tables, we can create external Elasticsearch tables and insert our Hive data
- We can then run Kibana, define an index pattern and use the Kibana UI



# Processing & Tools

- **Spark**

- Used to measure correlations between stocks with Machine learning library (MLlib)
- Correlation is an important measure in stock market
  - Measure the amount of diversification among the stocks in a portfolio
  - Choosing assets with low correlation help to reduce the risk

- **Spark SQL**

- Used to examine recommendation trends
- Pulled top analysts recommendation data using “[ticker].recommendations”
- Filter to count votes from analysts

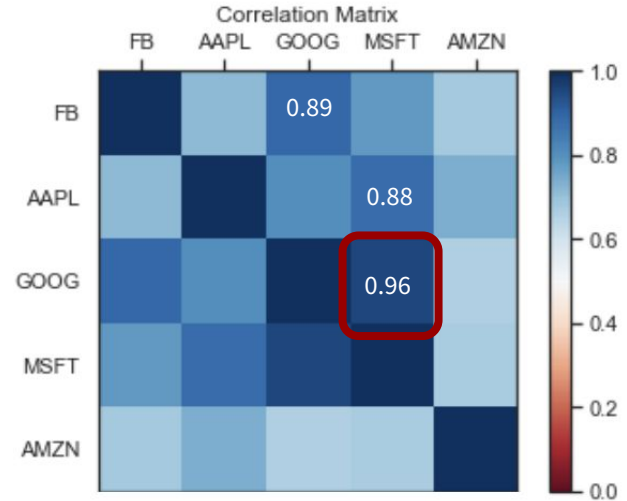


# Analysis & Insights



I want to sell one asset in my portfolio that is too closely related.

- ☐ Meta
- ☐ Apple
- ☐ Google
- ☐ Microsoft
- ☐ Amazon



- Examined last 1 year adj close prices
- **Very high positive** correlation between **Microsoft** and **google**
- High positive correlation between Meta and Google & Apple and Microsoft

# Analysis & Insights



Interested in buying. Any thoughts?

📄 Tesla

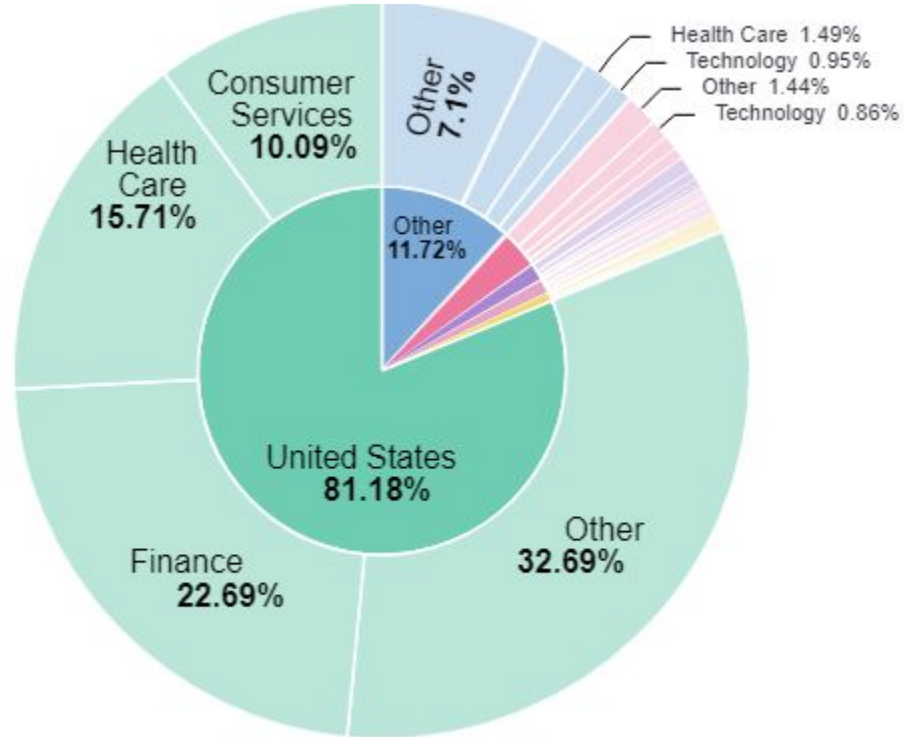
Date	Firm	To Grade	From Grade	Action
2021-09-23	Tudor Pickering	Sell		init
2021-10-04	RBC Capital	Sector Perform		main
2021-10-08	Canaccord Genuity	Buy		main
2021-10-14	Barclays	Underweight		main

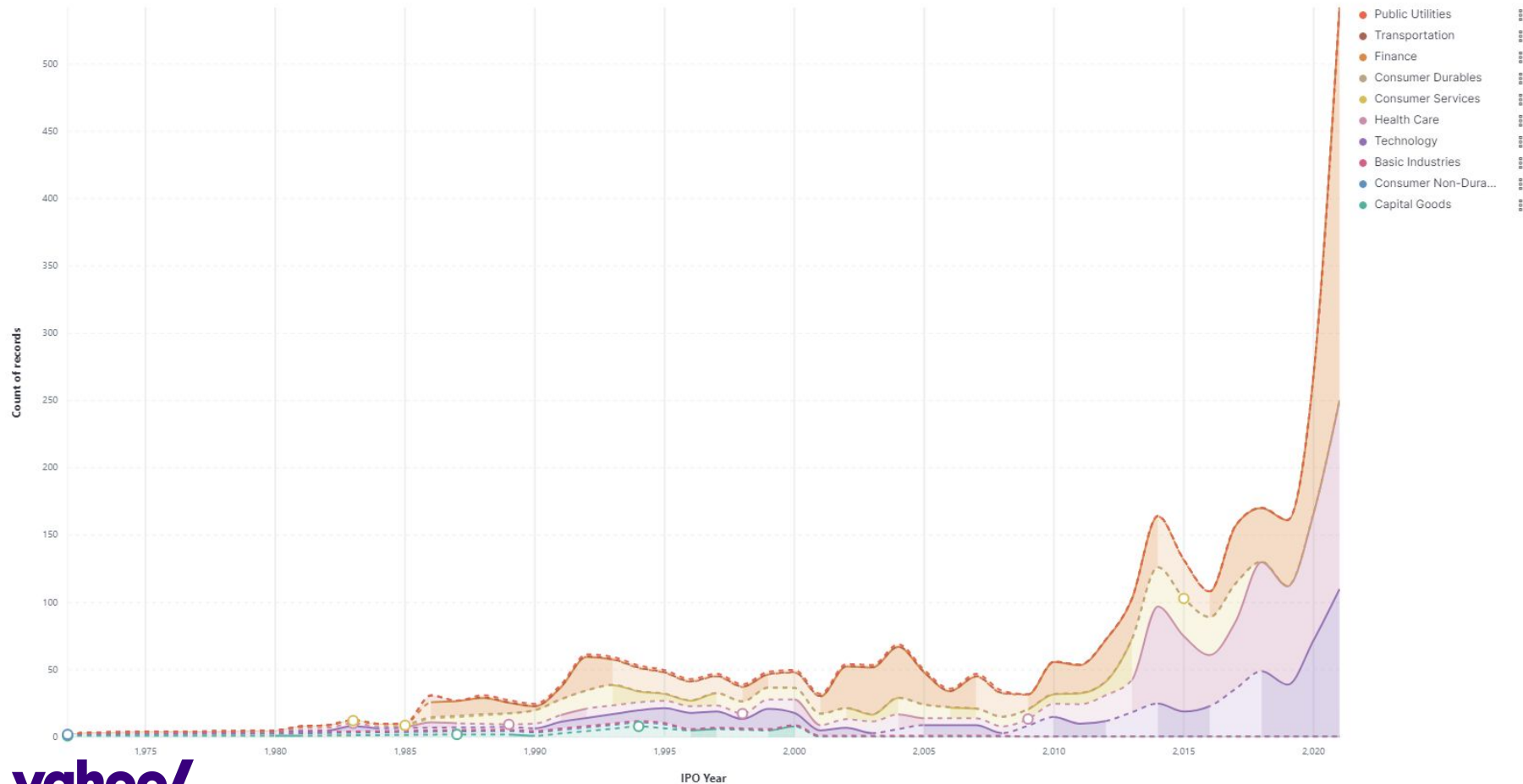


To Grade	count
Overweight	1
Outperform	1
Sell	1
Underweight	1
Buy	6
Neutral	2
Equal-Weight	1
Sector Perform	2

- Pulled last 3 months recommendations
- Most votes for "Buy" from top analysts

# Analysis & Insights





# Lessons Learned

- The infrastructure is by far the hardest part of the setup
- There is a lot of applicability in the big data tools taught in class
  - The tools we used were very powerful tools for any data-enthusiast
- ETL pipelines → Powerful but maintenance increases with additional infrastructure added