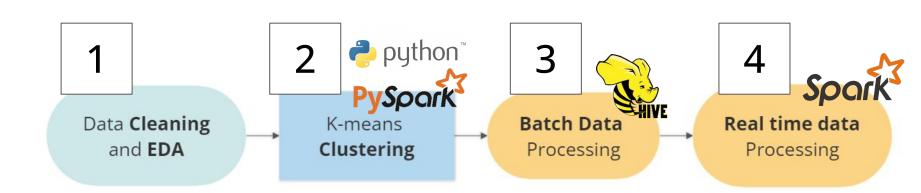
Nasdaq Stock Market Analysis

Module 6: Final Presentation

ALY6110, 2023 June 30th

Presented to Professor Behzad Ahmadi Prepared by Shyamala Venkatakrishnan, Heejae Roh

Work Flow & Index





Nasdaq

1 Data Cleaning & EDA

Stock Market Analysis

	Α	В	С	D	E	F	G	Н
1	Date	Open	High	Low	Close	Adj Close	Volume	Symbol
2	1980-12-12	0.128348	0.128906	0.128348	0.128348	0.099584	469033600	AAPL
3	1980-12-15	0.12221	0.12221	0.121652	0.121652	0.094388	175884800	AAPL
4	1980-12-16	0.113281	0.113281	0.112723	0.112723	0.087461	105728000	AAPL
5	1980-12-17	0.115513	0.116071	0.115513	0.115513	0.089625	86441600	AAPL
6	1980-12-18	0.118862	0.11942	0.118862	0.118862	0.092224	73449600	AAPL
201602	2023-06-02	98.309998	99.660004	98.190002	99.5	99.5	6220300	SBUX
201603	2023-06-05	100.04	100.87	99.5	99.93	99.93	6536500	SBUX
201604	2023-06-06	99.739998	100.33	97.529999	98.220001	98.220001	8657100	SBUX
201605	2023-06-07	98.550003	98.860001	97.589996	97.919998	97.919998	6285600	SBUX
201606	2023-06-08	97.650002	99.18	96.919998	99.150002	99.150002	6877800	SBUX

Stock data of Top 24 Nasdaq companies (by Market Cap, June 8th)

Num rows: 201,606 and Num columns: 8

Dataset describe: from Yahoo Finance, Top 24 Market Cap, Max period of each stock

Fields explanation: Date/ Prices (open, high, low, close, adj close)/ Volume/ Symbol (added)

Cleaning the Data

e of Data & i	Siluli			
#	Column	Non-Null Count	Dtype	isnull
0	Date	201611 non-null	object	0
1	Open	201611 non-null	float64	0
2	High	201611 non-null	float64	0
3	Low	201611 non-null	float64	0
4	Close	201611 non-null	float64	0
5	Adj Close	201611 non-null	float64	0
6	Volume	201611 non-null	int64	0
7	Symbol	201611 non-null	object	0

Drop: Adj Close (Adjusted Close), Checking NA values

After Drop: 201,606 Num of rows and 7 Columns

Descriptive Analysis of 24 stocks

	Open	High	Low	Close	Volume	Year
count	201611	201611	201611	201611	201611	201611
mean	47.75	48.44	47.23	47.86	4.27e+07	2003.7
std	81.85	82.80	80.74 2	81.80	1.15e+08	13.08
min	0.00	0.05	0.05	0.05	0.00	1962.00
25%	3.84	4.34	4.18	4.26	3.14	1994.00
50%	18.95	19.25	18.67	18.95	1.01	2006.00
75%	51.70	52.31	51.00	51.70	3.82	2015.00
max	835.12	921.78	798.80	812.73	7.42	2023.00

Min year: 1962 (HON) Mean price: 47.86

Focused Analysis on AAPL

	Open	High	Low	Close	Volume	Year
count	10718	10718	10718	10718	10718	10718
mean	17.99	18.20	17.80	18.00	3.24e+08	2001.68
std	37.90	38.36	37.49	37.94	3.37e+08	12.28
min	0.05	0.05	0.05	0.05	0.00e+00	1980
25%	0.29	0.30	0.28	0.29	1.18e+08	1991
50%	0.50	0.50	0.49	0.50	2.12e+08	2002
75%	17.80	17.94	17.61	17.80	4.04e+08	2012
max	186.73	186.99	194.27	186.01	7.42e+09	2023
					·	

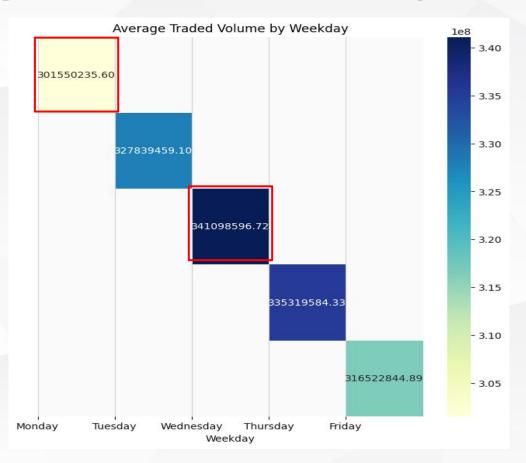
Min year: 1980(APPL) Mean price: 18.00 Max year: 2023 Max Price: 186.73 Volume min: 0.00

Volume max: 7.42e+09

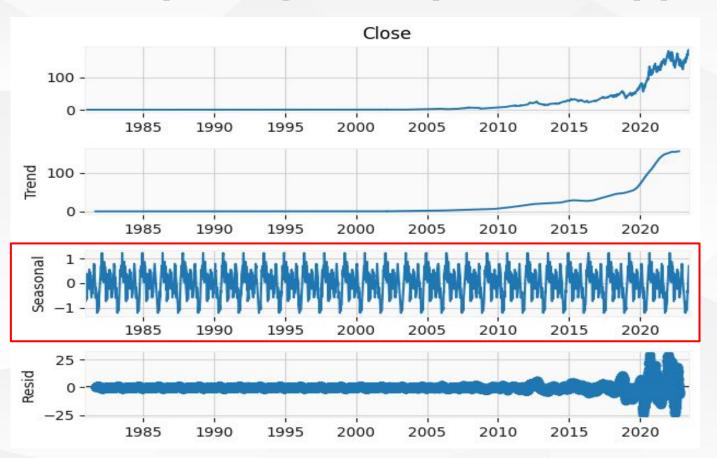
Candle chart of AAPL for one month



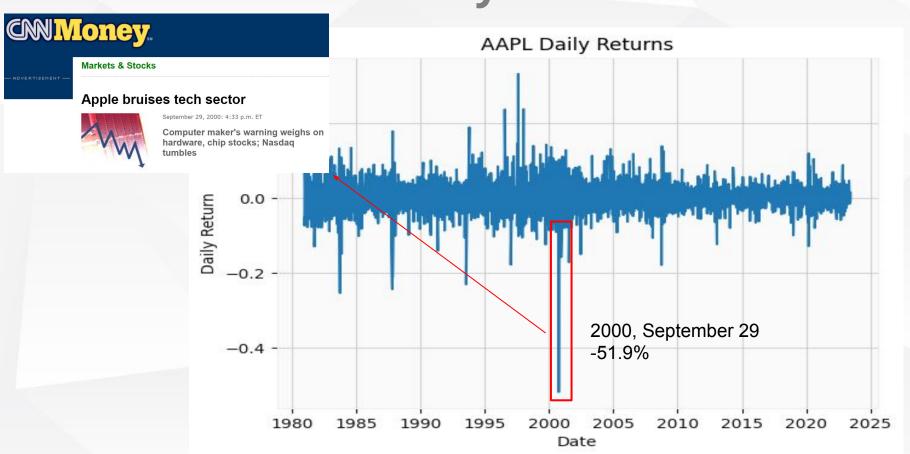
Average Traded Volume by Weekday



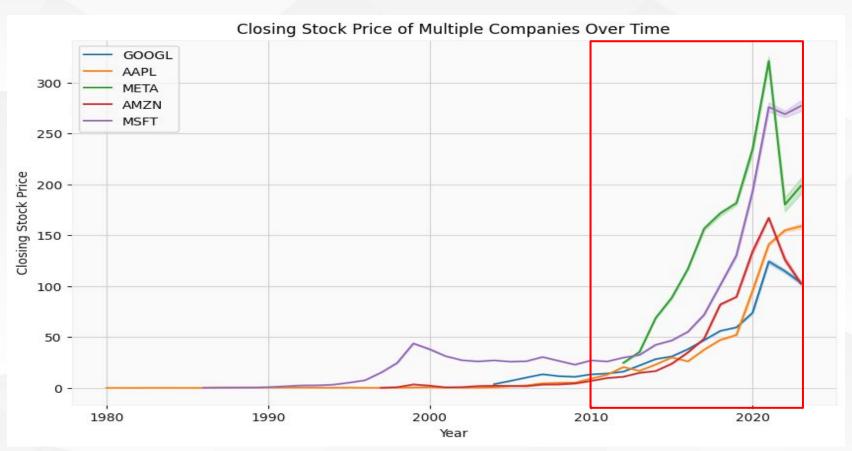
Decomposing close price of Apple



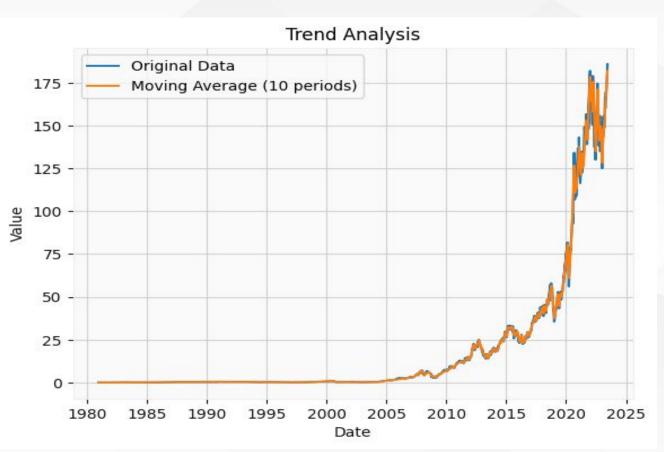
AAPL Daily Returns



Closing Stock Price of 5 Companies

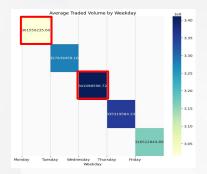


Moving Average (10 periods)



Summary of insights from EDA





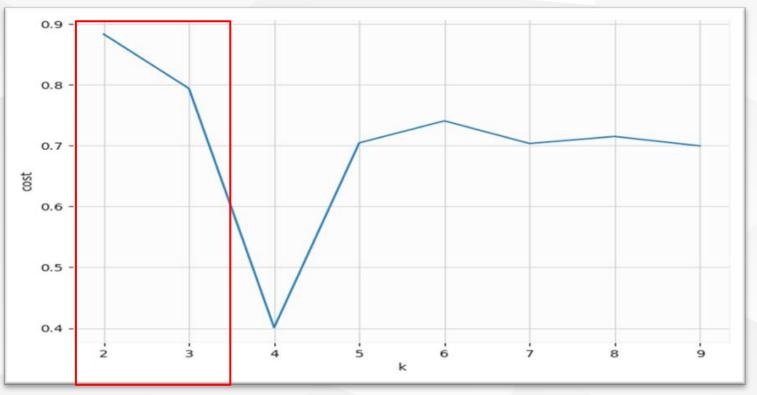


- Close Price over time & one year & one month
 - 1. AAPL have risen sharply since 2010 and did not exceed \$20 until 2010.
 - 2. Within one year, affected monetary policy and interest rates
 - 3. Within one month, affected big issue which is Apple Vision pro announcement
- Average Trade Volume by Weekday
 - Lowest in Monday and Highest in Wednesday according to News and Events
- Decomposing close price trend, seasonal, and residual
 - a change or variations is uniform for seasonal component
- 5 Companies in one graph
 - Confirm when the company started to rise
- AAPL Daily Returns
 - Checking lowest return of Day in 2000, September
- Correlation between Stock Close price
 - Most Companies generally have positive correlations, because they are most successful companies by market cap
 - AAPL & GOOGL, and GOOGL & AMZN have highest correlations



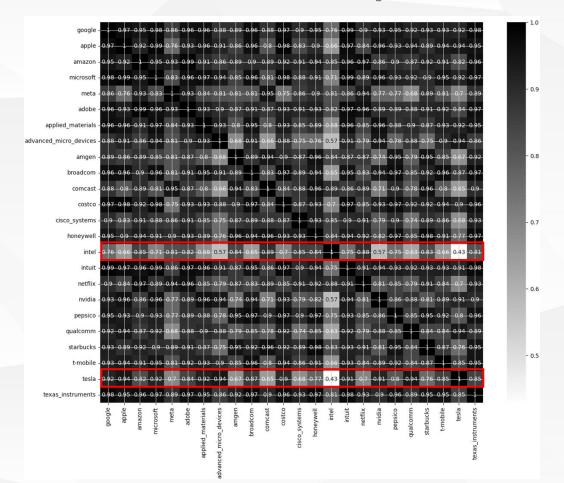
Diversified Portfoliowith K-means Clustering

Silhouette Score

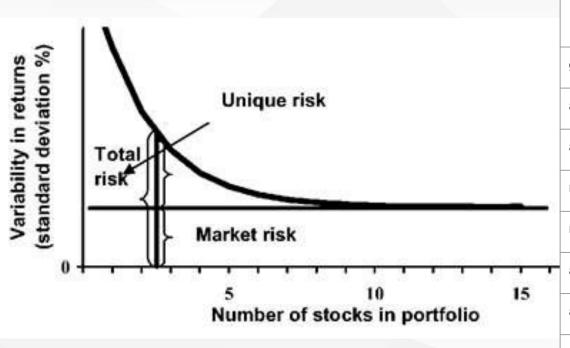


Silhouette Coefficient or silhouette score is close to 1, meaning that the cluster is well divided. This is extracted using pyspark.

Correlation Matrix of 24 companies' close price

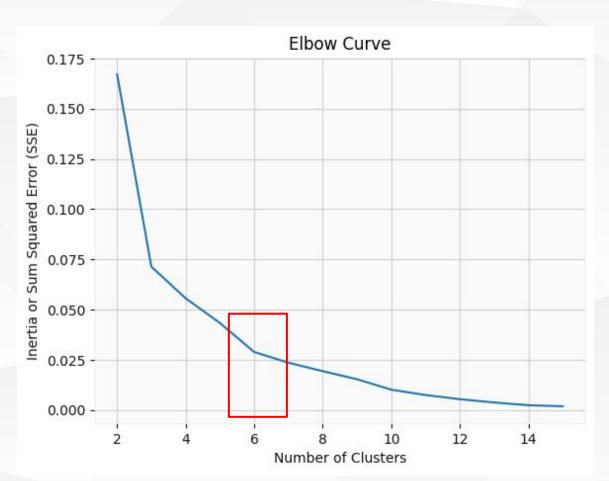


Return and Variance of Return

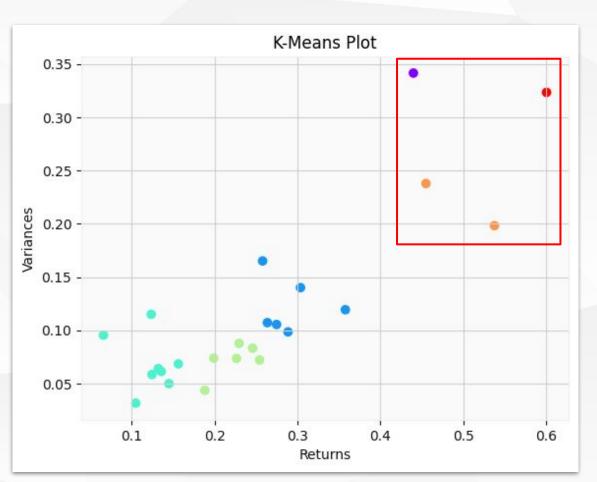


Company_name	Variances	Returns
google	0.0736	0.2265
apple	0.0831	0.2460
amazon	0.1054	0.2749
microsoft	0.0721	0.2543
meta	0.1650	0.2580
adobe	0.0984	0.2886
applied_materials	0.1400	0.3037
amd	0.3415	0.4399

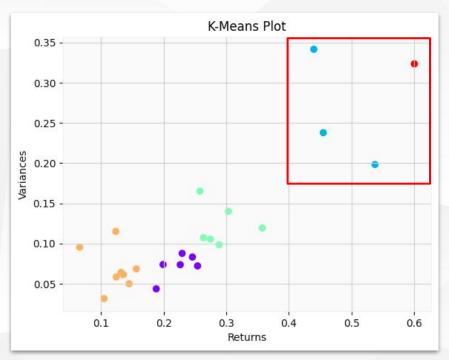
Elbow Curve for Clustering

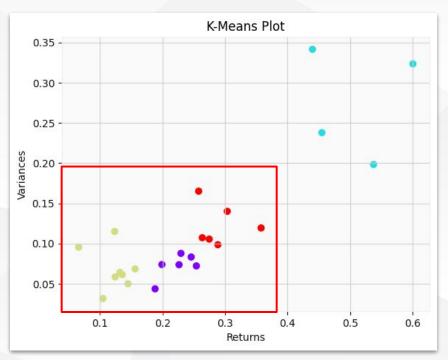


Cluster number 6



Cluster number 5 & 4

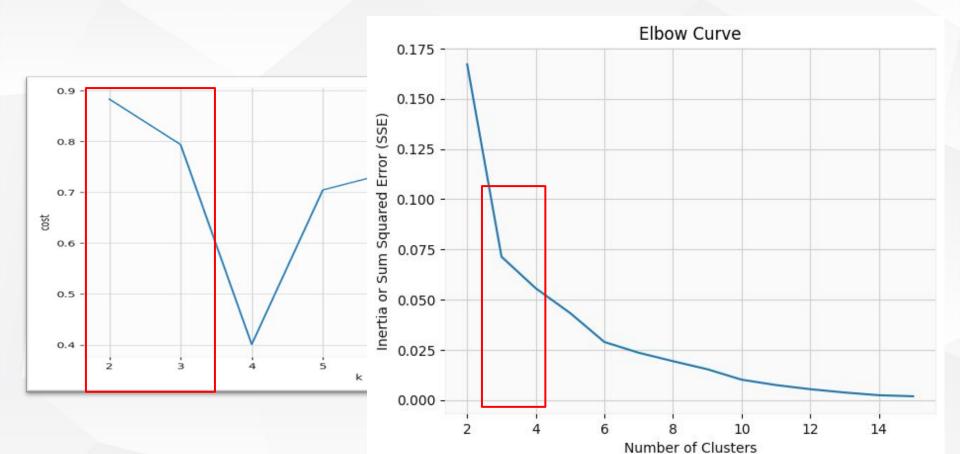




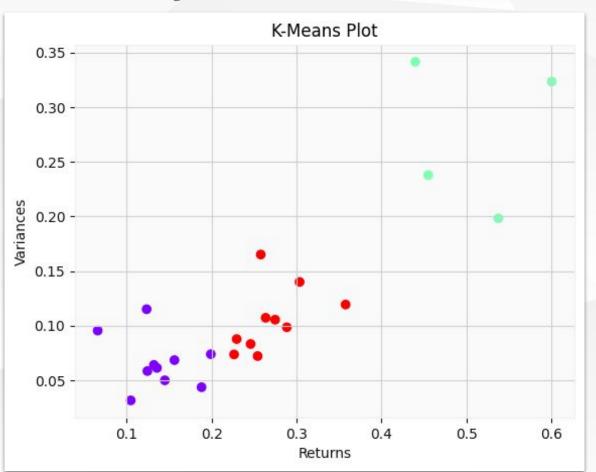
N = 5

N = 4

Elbow Curve and Silhouette Score



Finally, Cluster number 3

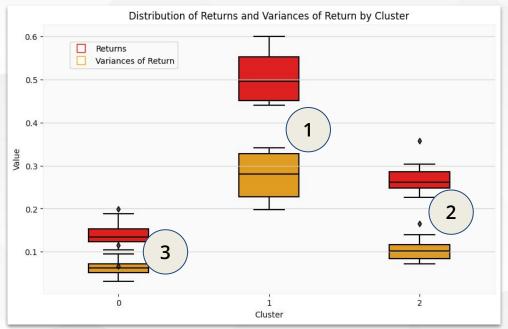


Finally, Cluster number 3

Company_name	Variances	Returns	Cluster
google	0.0736	0.2265	2
apple	0.0831	0.2460	2
amazon	0.1054	0.2749	2
microsoft	0.0721	0.2543	2
meta	0.1650	0.2580	2
adobe	0.0984	0.2886	2
applied_materials	0.1400	0.3037	2
amd	0.3415	0.4399	1
amgen	0.0612	0.1357	0

Company_name	Variances	Returns	Cluster
intuit	0.0876	0.2296	2
netfilx	0.2378	0.4551	1
nvidia	0.1982	0.5377	1
pepsico	0.0315	0.1049	0
qualcomm	0.1150	0.1236	0
starbucks	0.0684	0.1563	0
t-mobile	0.1071	0.2637	2
tesla	0.3234	0.6006	1
texas_instrument	0.0738	0.1992	0

Visualization of Cluster



Avg_Variance	0.0661	0.2752	0.1052
Avg_Return	0.1375	0.5083	0.2703
Cluster	0	1	2

Diversified Portfolio with Cluster number = 3

index	Company_name	Cluster_Labels	
8	amgen	0	AMGEN
0	advanced_micro_devices	1	AMD
3	google	2	Google

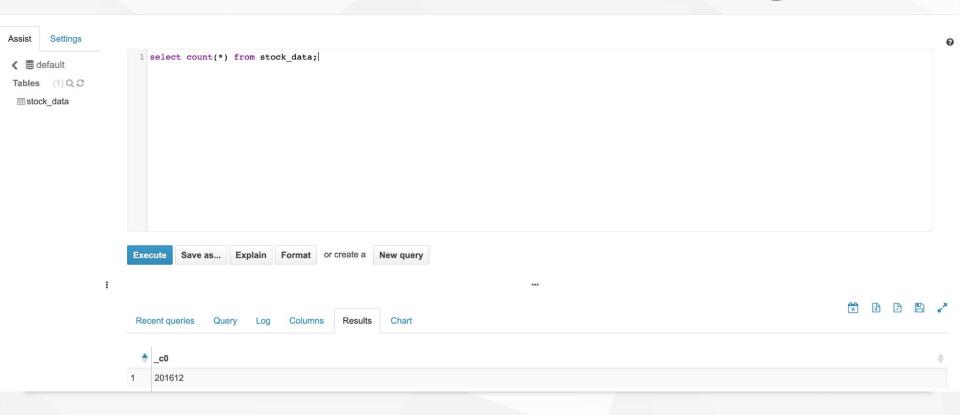
Batch Data Processing

Creation of HIVE table and loading data



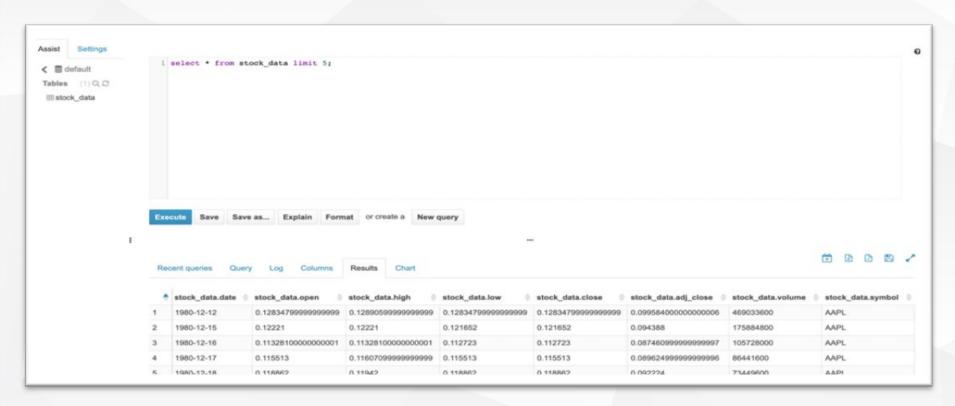
Initially, the stock market csv file was imported into HDFS from the local machine using Hive File browser. An external HIVE table was created using the file location and there are a total of 201,612 records in the table.

Creation of HIVE table and loading data



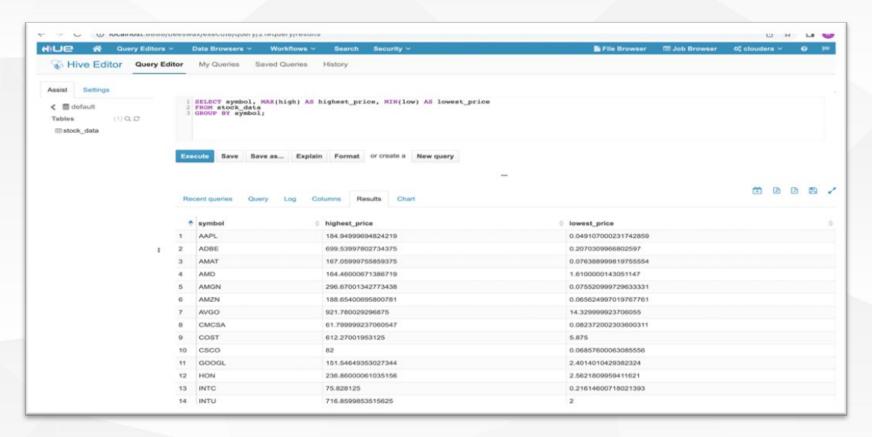
Data is Uploaded

Creation of HIVE table and loading data



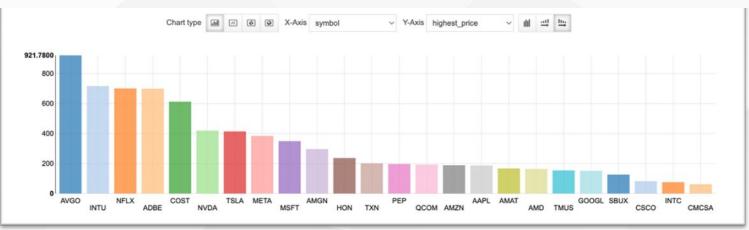
Data is Uploaded

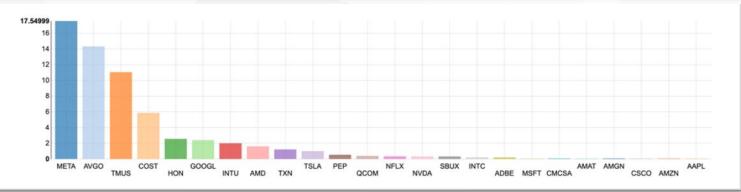
1. What is the highest and lowest price for each of the stock symbols?



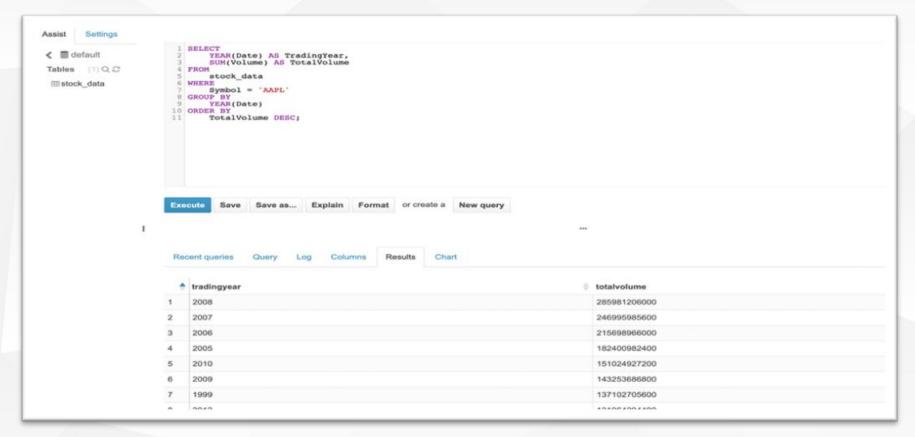
1. What is the highest and lowest price for each of the stock symbols?



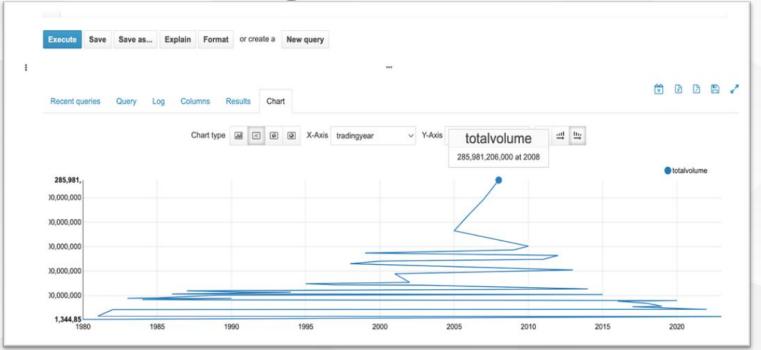




2. Which year has the greatest volume of trading for AAPL stock?

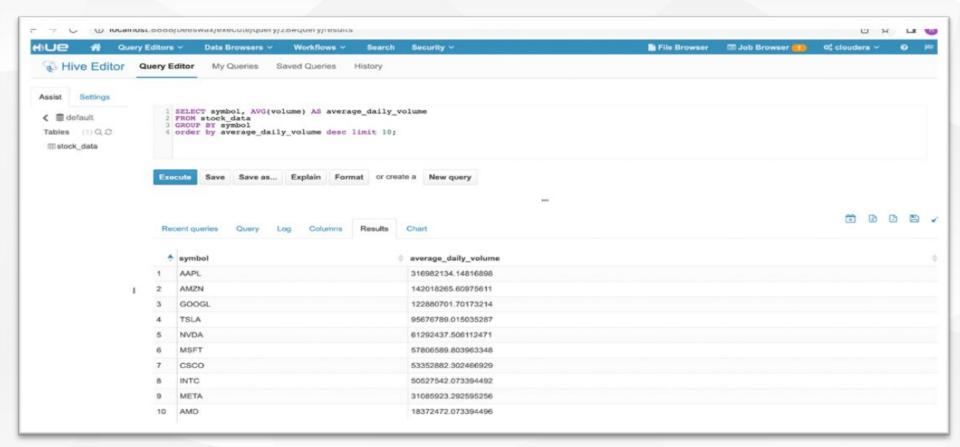


2. Which year has the greatest volume of trading for AAPL stock?

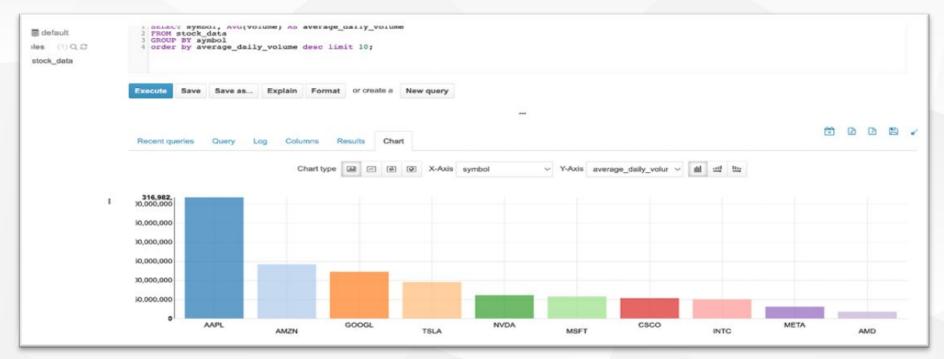


A line chart is used to visualize the total traded volume for AAPL stock every year and based on the above results, 2008 is the year in which maximum trading of stock units happened for AAPL stock

3. What are the top 10 stocks with high average daily volume of stocks traded?

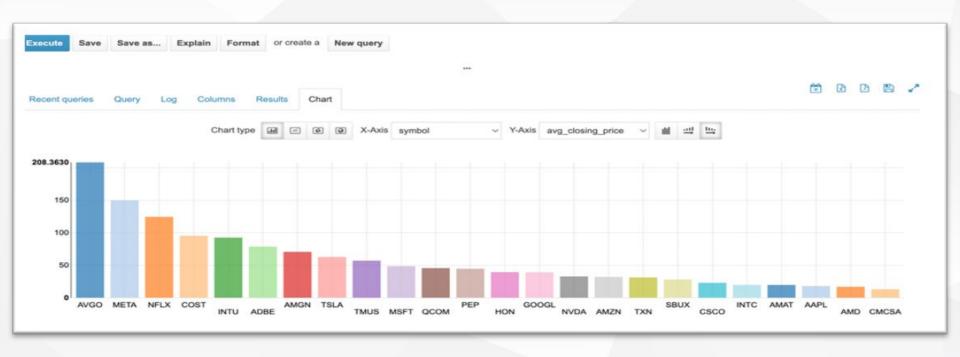


3. What are the top 10 stocks with high average daily volume of stocks traded?



The average of the volume of stocks traded is calculated for each company and the top 10 stocks with highest average volume traded are displayed and a bar chart is used to visualize the result.

4. What is the average closing stock price of the companies?





Real Time Data
Processing

Data Source - Alpaca Markets



- U.S based stock brokerage firm that provides commission-free trading services.
- It offers a platform and API that allows developers and traders to build trading algorithms, access real-time market data, and manage their investment portfolios.

Accessing real time market data API

- Creating an account
- free or paid subscription
- Generating API keys
- API KEY and API SECRET KEY
- Communication protocol Websocket
- WebSocket endpoint wss://stream.data.alpaca.markets/v2/iex

Your API Keys 1 Hide Endpoint https://paper-api.alpaca.markets API Keu ID Keys have not been generated. Generate New Keu

Client to Server

Input and output:

Server to Client

```
$ wscat -c wss://stream.data.alpaca.markets/v2/sip
connected (press CTRL+C to quit)
       [ {"T":"success", "msg":"connected"}]
          {"action": "auth", "key": "****", "secret": "****"}
           [{"T":"success","msg":"authenticated"}]
         {"action": "subscribe", "trades": ["AAPL"], "quotes": ["AMD", "CLDR"], "bars":
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< [{"T":"t", "S":"AAPL", "i":628, "x":"K", "p":162.92, "s":3, "c":["@", "F", "T", "I"], "z"
        [{"T":"q", "S":"AMD", "bx":"P", "bp":91.9, "bs":1, "ax":"Q", "ap":91.98, "as":1, "c":["F
```

Spark Structured Streaming

- Spark 2.x release onwards
- Built on the Spark SQL library
- Streaming based on Dataframe and Dataset APIs
- Implementation using PySpark library Python
- Response Json:

```
{
    "T": "b",
    "S": "SPY",
    "o": 388.985,
    "h": 389.13,
    "1": 388.975,
    "c": 389.12,
    "v": 49378,
    "t": "2021-02-22T19:15:00Z"
}
```

Defining Schema:

```
schema = StructType([
    StructField("T", StringType(), True),
    StructField("S", StringType(), True),
    StructField("o", DoubleType(), True),
    StructField("h", DoubleType(), True),
    StructField("l", DoubleType(), True),
    StructField("c", DoubleType(), True),
    StructField("v", LongType(), True),
    StructField("t", StringType(), True)
])
```

1. Identify the top 5 stocks with the highest closing price in the incoming batch of stock data.

received a message

```
[{"T":"b", "S":"CPRI", "o":35.45, "h":35.465, "l":35.465, "c":35.465, "v":442, "t":"2023-06-28T18:21:002", "n":8, "vw":35.46235
3},{"T":"b","S":"PNTG","o":12.23,"h":12.23,"l":12.23,"c":12.23,"v":234,"t":"2023-06-28T18:21:002","n":1,"vw":12.23},
{"T": "b", "S": "IAUX", "o": 2.08, "h": 2.08, "l": 2.08, "c": 2.08, "v": 100, "t": "2023-06-28T18: 21: 00Z", "n": 1, "vw": 2.08},
{"T":"b", "S":"USB", "o":32.22, "h":32.22, "1":32.215, "c":32.22, "v":400, "t":"2023-06-28T18:21:00Z", "n":4, "vw":32.21875},
{"T":"b","S":"CR","o":85.63,"h":85.63,"1":85.63,"c":85.63,"v":100,"t":"2023-06-28T18:21:00Z","n":1,"vw":85.63},
{"T":"b","S":"BMRN","o":92.5,"h":92.5,"l":92.47,"c":92.47,"v":200,"t":"2023-06-28T18:21:00Z","n":2,"vw":92.485},
{"T":"b", "S":"NCLH", "o":20.9, "h":20.945, "1":20.9, "c":20.945, "v":1124, "t":"2023-06-28T18:21:00Z", "n":16, "vw":20.91764
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31}, {"T":"b", "S":"APO", "o":75.02, "h":75.02, "l":74.995, "c":74.995, "v":310, "t":"2023-06-28T18:21:00Z", "n":4, "vw":75.003
71},{"T":"b","S":"TGI","o":12.33,"h":12.33,"l":12.33,"c":12.33,"v":945,"t":"2023-06-28T18:21:002","n":2,"vw":12.33},
{"T": "b", "S": "VEEV", "o": 196.275, "h": 196.275, "1": 196.275, "c": 196.275, "v": 100, "t": "2023-06-28T18: 21: 002", "n": 1, "vw": 19
6.275}, {"T": "b", "S": "SPOT", "o": 160.46, "h": 160.46, "l": 160.46, "c": 160.46, "v": 102, "t": "2023-06-28T18: 21: 002", "n": 2, "vw":
160.459804}, {"T": "b", "S": "URBN", "o": 32.31, "h": 32.31, "l": 32.31, "c": 32.31, "v": 516, "t": "2023-06-28T18: 21: 00Z", "n": 7, "v
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8,"vw":1.888321},{"T":"b","S":"CRK","o":11.07,"h":11.075,"1":11.07,"c":11.075,"v":400,"t":"2023-06-28T18:21:00Z","n":
5,"vw":11.07375},{"T":"b","S":"HLVX","o":16.64,"h":16.64,"l":16.64,"c":16.64,"v":100,"t":"2023-06-28T18:21:002","n":
1,"vw":16.64},{"T":"b","S":"MEOH","o":40.05,"h":40.05,"l":40.05,"c":40.05,"v":110,"t":"2023-06-28T18:21:00Z","n":2,"v
```

Output

```
result

+---+---+

| S|max(c)|

+---+---+

|COKE|646.39|

|LMT|447.83|

|NVDA|414.65|

|IGV|341.83|

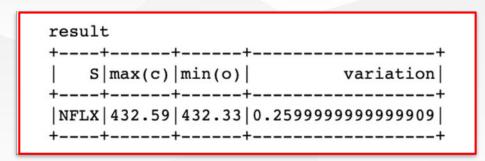
|META|285.73|

+---+-----+
```

2. Find the stock with largest price variation in the incoming batch data.

VV	n	t	v	c	1	h	٥	s	т
27.424911	6	2023-06-28T18:35:00Z	503	27.46	27.41	27.46	27.42	CVNA	b
390.738339	7	2023-06-28T18:35:00Z	277	390.74	390.73	390.74	390.73	SPGI	b
34.858696	2	2023-06-28T18:35:00Z	115	34.86	34.86	34.86	34.86	UPST	b
4.615	3	2023-06-28T18:35:00Z	321	4.615	4.615	4.615	4.615	TAST	b
3.84	2	2023-06-28T18:35:00Z	200	3.84	3.84	3.84	3.84	ADMA	b
0.2545	2	2023-06-28T18:35:00Z	200	0.2545	0.2545	0.2545	0.2545	EVLO	b
41.99	5	2023-06-28T18:35:00Z	315	41.99	41.99	41.99	41.99	CMA	b
43.949519	3	2023-06-28T18:35:00Z	104	43.95	43.95	43.95	43.95	PLAY	b
18.6	1	2023-06-28T18:35:00Z	100	18.62	18.62	18.62	18.62	ARCC	b
110.52944	8	2023-06-28T18:35:00Z	1274	110.55	110.525	110.55	110.525	AMD	b
13.74	1	2023-06-28T18:35:00Z	200	13.745	13.745	13.745	13.745	PEB	b
7.8	14	2023-06-28T18:35:00Z	1395	7.88	7.88	7.88	7.88	NU	b
42.47461	5	2023-06-28T18:35:00Z	475	42.475	42.475	42.475	42.475	MGM	b
110.02548	5	2023-06-28T18:35:00Z	364	110.025	110.025	110.025	110.025	RYAAY	b
14.84	3	2023-06-28T18:35:00Z	3700	14.845	14.845	14.845	14.845	PHYS	b
432.45678	6	2023-06-28T18:35:00Z	454	432.59	432.33	432.59	432.33	NFLX	b
63.752	3	2023-06-28T18:35:00Z	400	63.75	63.75	63.76	63.75	BERY	b

Output



Conclusions

EDA

- Close Price over time & one year & one month: AAPL did not exceed \$20 until 2010.
- Average Trade Volume by Weekday: Lowest in Monday and Highest in Wednesday
- Decomposing close price trend: a change or variations is uniform for seasonal component
- AAPL Daily Returns: Checking lowest return of Day in 2000, September
- Correlation between Stock Close price: Most Companies generally have positive correlations

K-means Clustering

- Silhouette Score shows number of cluster 3 and Elbow Curve shows 6 is suitable
- After checking the visualization by applying cluster numbers from 6 to 3, 3 is decided
- Confirm that return and variance are in descending order in the order of cluster 1,2,0 through Visualization of Cluster
- The Simplest Diversified Portfolio is amgen from cluster 0, AMD from cluster 1, and Google from 2.

Hive and Spark

- AVGO is highest price stock. The lowest price of stocks, META was the highest among stocks at 17.
- Apple stock had the highest trading volume in 2008.
- Average daily volume was in order of Apple, Amazon, Google, Tesla, and NVDA.

Recommendations

- Analyze more Nasdaq stocks
 - Extract meaningful information by analyzing not only 24 companies but also companies of a wider spectrum
- Composition of a specific portfolio
 - Composing a specific portfolio including the proportion of investment by each company
 - Reflect the seasonal component in the investment period
- Establishment of clustering targeting more companies and earning specific criteria
 - o Recorded the criterion how stocks are divided into cluster and stored historically.
- Dashboard composition through API data
 - 1. Real-time trade price tracking
 - 2. Find anomalies in volume
 - 3. Real-time Trend analysis
- Building an automated trading system
 - Establishment of a model for automatic trading based on API data and standards

Thanks! Do you have any questions?

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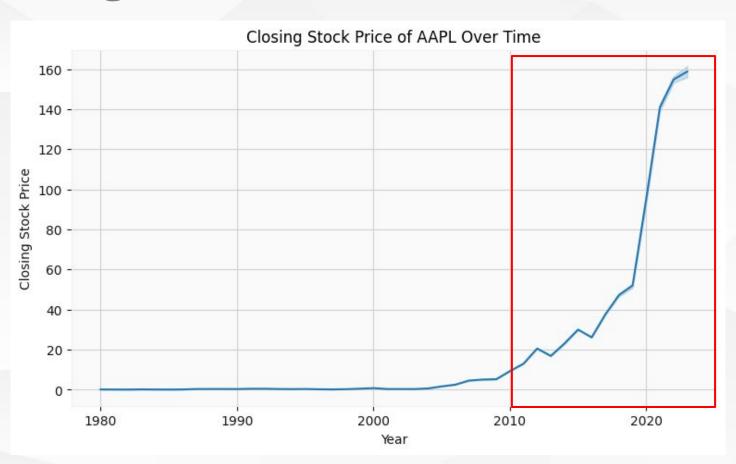
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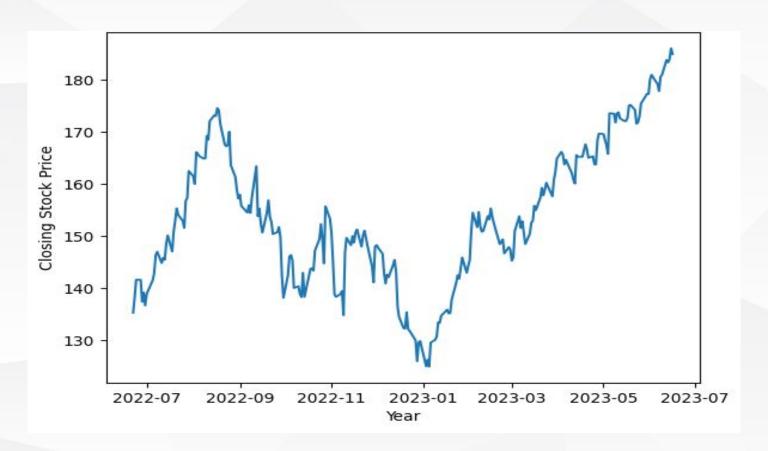
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APPENDIX EDA

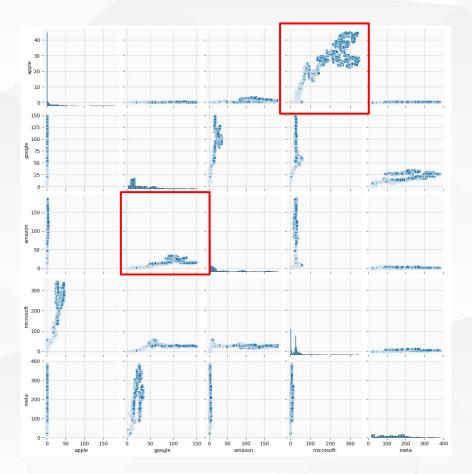
Closing Stock Price of AAPL over time



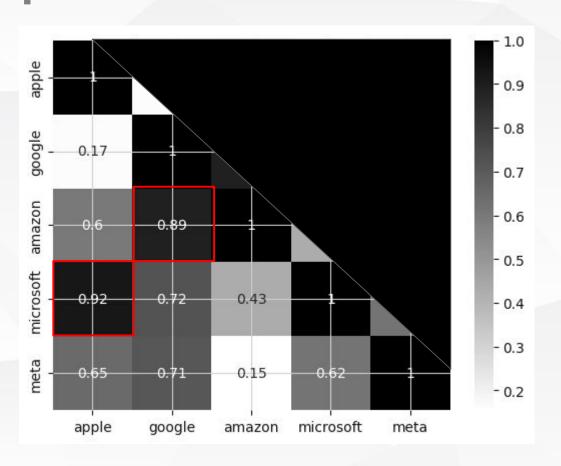
Closing Stock Price of AAPL for one year



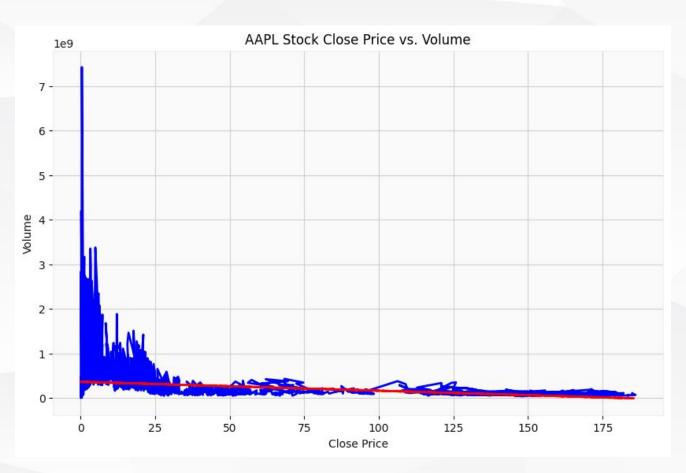
Pair Plot of 5 stocks(AAPL, GOOGL, AMZN, MSFT, META)



Heatmap of 5 stocks(AAPL, GOOGL, AMZN, MSFT, META)



AAPL Stock Close Price vs Volume



OHLC Price Chart of GOOGL

OHLC Price Chart of GOOGL May 2023 - June 2023



AAPL/ GOOGL/ AMZN/ MSFT date vs volume

