## d6o2ixdfp

## January 20, 2025

```
[1]: # Import necessary libraries
     import pandas as pd
     import numpy as np
     import matplotlib.pyplot as plt
     from pymongo import MongoClient
     import requests
     from datetime import datetime
     import json
     # Configure MongoDB connection
     mongo_client = MongoClient("mongodb+srv://ismini6:9LogpPCZJUJEZyjl@cluster0.
     ⇔8xihe.mongodb.net/")
     db = mongo_client['Crypto']
     collection = db['Cryptocurrency']
     # Data Ingestion
     def load_data(file_path):
         df = pd.read_csv(file_path)
         return df
     # Load all datasets from the 'archive' folder
     import os
     data_folder = 'C:/Users/Nantia/Downloads/archive/'
     crypto_files = [f for f in os.listdir(data_folder) if f.endswith('.csv')]
     crypto_data = pd.DataFrame()
     for file in crypto_files:
         file_path = os.path.join(data_folder, file)
         data = load_data(file_path)
         crypto_data = pd.concat([crypto_data, data], ignore_index=True)
     print(crypto_data.head())
     # Data Cleaning and Transformation
     crypto_data['Date'] = pd.to_datetime(crypto_data['Date'])
```

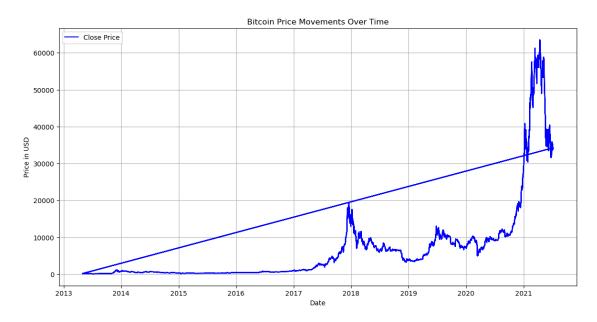
```
# Drop duplicates and handle missing values
crypto_data.drop_duplicates(inplace=True)
crypto_data.fillna(method='ffill', inplace=True)
# Store cleaned data to MongoDB
collection.insert_many(crypto_data.to_dict('records'))
# Relevant Queries for Data Extraction
# Example query: Get all data for Bitcoin
bitcoin_data = list(collection.find({"Symbol": "BTC"}))
bitcoin_df = pd.DataFrame(bitcoin_data)
# Convert 'Date' back to datetime for plotting
bitcoin_df['Date'] = pd.to_datetime(bitcoin_df['Date'])
# Analyzing Price Movements
plt.figure(figsize=(14, 7))
plt.plot(bitcoin_df['Date'], bitcoin_df['Close'], label='Close Price', __

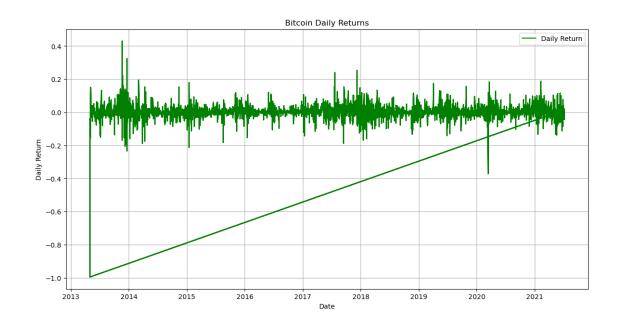
color='blue')

plt.title('Bitcoin Price Movements Over Time')
plt.xlabel('Date')
plt.ylabel('Price in USD')
plt.legend()
plt.grid()
plt.show()
# Big Data Processing Solution (Batch Analysis)
bitcoin_df['Daily Return'] = bitcoin_df['Close'].pct_change()
# Visualize Daily Returns
plt.figure(figsize=(14, 7))
plt.plot(bitcoin_df['Date'], bitcoin_df['Daily Return'], label='Daily Return', u
 ⇔color='green')
plt.title('Bitcoin Daily Returns')
plt.xlabel('Date')
plt.ylabel('Daily Return')
plt.legend()
plt.grid()
plt.show()
C:\Users\Nantia\anaconda3\Lib\site-packages\cryptography\x509\base.py:594:
CryptographyDeprecationWarning: Parsed a negative serial number, which is
disallowed by RFC 5280.
  return rust_x509.load_der_x509_certificate(data)
  SNo Name Symbol
                                                                     Open \
                                    Date
                                               High
                                                           Low
    1 Aave
             AAVE 2020-10-05 23:59:59 55.112358 49.787900 52.675035
```

1	2	Aave	AAVE	2020-10-06	23:59:59	53.402270	40.734578	53.291969
2	3	Aave	AAVE	2020-10-07	23:59:59	42.408314	35.970690	42.399947
3	4	Aave	AAVE	2020-10-08	23:59:59	44.902511	36.696057	39.885262
4	5	Aave	AAVE	2020-10-09	23:59:59	47.569533	43.291776	43.764463

	Close	Volume	Marketcap
0	53.219243	0.000000e+00	8.912813e+07
1	42.401599	5.830915e+05	7.101144e+07
2	40.083976	6.828342e+05	6.713004e+07
3	43.764463	1.658817e+06	2.202651e+08
4	46.817744	8.155377e+05	2.356322e+08





```
[3]: # Import necessary libraries
    import pandas as pd
     import numpy as np
    import matplotlib.pyplot as plt
    import seaborn as sns
    import glob
    import os
    from sklearn.model_selection import train_test_split
    from sklearn.linear model import LinearRegression
    from sklearn.metrics import mean_squared_error
    import nltk
    from nltk.sentiment.vader import SentimentIntensityAnalyzer
    # Download VADER sentiment analysis model from nltk
    nltk.download('vader_lexicon')
     # Define the path to your dataset folder
    data_folder = 'C:/Users/Nantia/Downloads/archive/'
     # Load all CSV files into a single DataFrame
    all_files = glob.glob(os.path.join(data_folder, "*.csv"))
    dfs = \Pi
    for filename in all_files:
        df = pd.read_csv(filename)
        dfs.append(df)
     # Concatenate all DataFrames into one
    data = pd.concat(dfs, ignore_index=True)
     # Display the first few rows
    data.head()
    [nltk_data] Downloading package vader_lexicon to
    [nltk_data]
                    C:\Users\Nantia\AppData\Roaming\nltk_data...
    [nltk_data]
                  Package vader_lexicon is already up-to-date!
[3]:
       SNo Name Symbol
                                        Date
                                                   High
                                                               Low
                                                                          Open \
                   AAVE
                         2020-10-05 23:59:59 55.112358 49.787900
         1 Aave
                                                                    52.675035
    1
         2 Aave
                   AAVE
                         2020-10-06 23:59:59 53.402270 40.734578
                                                                    53.291969
    2
                   AAVE 2020-10-07 23:59:59 42.408314 35.970690
         3 Aave
                                                                    42.399947
    3
         4 Aave
                   AAVE
                         2020-10-08 23:59:59 44.902511 36.696057
                                                                    39.885262
         5 Aave
                   AAVE 2020-10-09 23:59:59 47.569533 43.291776 43.764463
           Close
                        Volume
                                   Marketcap
```

```
1 42.401599 5.830915e+05 7.101144e+07
    2 40.083976 6.828342e+05 6.713004e+07
    3 43.764463 1.658817e+06 2.202651e+08
    4 46.817744 8.155377e+05 2.356322e+08
[4]: # Check for missing values
    print(data.isnull().sum())
     # Fill or drop missing values
    data.dropna(inplace=True)
     # Convert 'Date' column to datetime type
    data['Date'] = pd.to_datetime(data['Date'])
    # Reset index
    data.reset_index(drop=True, inplace=True)
    # Display cleaned data info
    data.info()
    SNo
                 0
                 0
    Name
    Symbol
                 0
                 0
    Date
                 0
    High
    Low
                 0
                 0
    Open
    Close
                 0
                 0
    Volume
    Marketcap
    dtype: int64
    <class 'pandas.core.frame.DataFrame'>
    RangeIndex: 37082 entries, 0 to 37081
    Data columns (total 10 columns):
                   Non-Null Count Dtype
     #
         Column
         -----
                    _____
                    37082 non-null int64
     0
         SNo
     1
         Name
                    37082 non-null object
     2
         Symbol
                    37082 non-null object
     3
                    37082 non-null datetime64[ns]
         Date
     4
         High
                    37082 non-null float64
     5
                    37082 non-null float64
         Low
                    37082 non-null float64
     6
         Open
     7
         Close
                    37082 non-null float64
                    37082 non-null float64
         Volume
         Marketcap 37082 non-null float64
```

0 53.219243 0.000000e+00 8.912813e+07

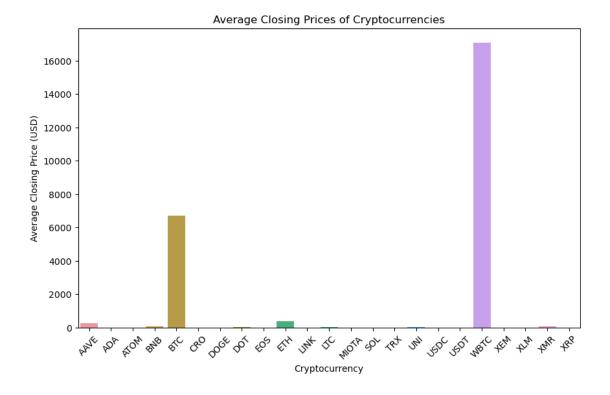
```
dtypes: datetime64[ns](1), float64(6), int64(1), object(2)
memory usage: 2.8+ MB
```

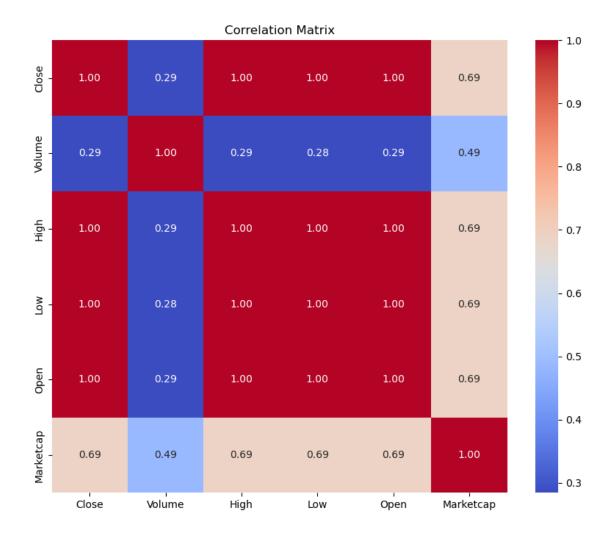
```
[5]: import pandas as pd
  import matplotlib.pyplot as plt
  import seaborn as sns

# Assuming 'data' is your DataFrame containing the cryptocurrency data
  avg_close = data.groupby('Symbol')['Close'].mean().reset_index()
  print(avg_close)

# Visualizing average closing prices
  plt.figure(figsize=(10, 6))
  sns.barplot(data=avg_close, x='Symbol', y='Close')
  plt.title('Average Closing Prices of Cryptocurrencies')
  plt.xlabel('Cryptocurrency')
  plt.ylabel('Average Closing Price (USD)')
  plt.xticks(rotation=45)
  plt.show()
```

	Symbol	Close
0	AAVE	255.525845
1	ADA	0.256313
2	ATOM	6.768099
3	BNB	52.250308
4	BTC	6711.290443
5	CRO	0.081912
6	DOGE	0.013763
7	DOT	18.143080
8	EOS	4.624088
9	ETH	383.910691
10	LINK	6.308583
11	LTC	49.279008
12	ATOIM	0.729370
13	SOL	10.471388
14	TRX	0.032585
15	UNI	17.077256
16	USDC	1.003791
17	USDT	1.000696
18	WBTC	17086.573875
19	XEM	0.124662
20	XLM	0.101509
21	XMR	74.134773
22	XRP	0.234790





```
model.fit(X_train, y_train)

# Make predictions
y_pred = model.predict(X_test)

# Evaluate the model
mse = mean_squared_error(y_test, y_pred)
print(f'Mean Squared Error: {mse}')
```

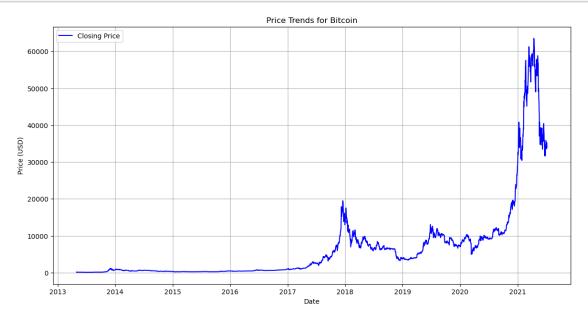
Mean Squared Error: 15899.032311721452

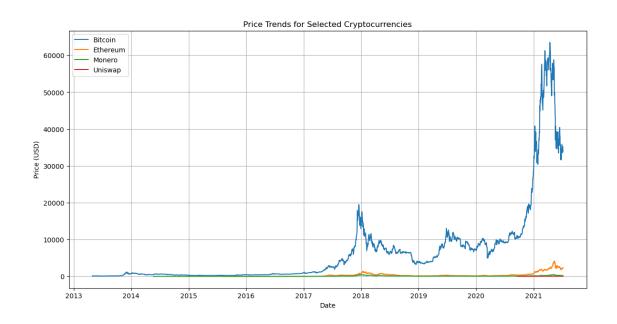
```
[8]: import pandas as pd
     import os
     # Function to load CSV files from a directory
     def load_crypto_data(folder_path):
         data = \{\}
         for filename in os.listdir(folder_path):
             if filename.endswith('.csv'):
                 filepath = os.path.join(folder_path, filename)
                 # Load the data into a DataFrame
                 df = pd.read_csv(filepath)
                 # Store DataFrame in a dictionary with the cryptocurrency name as ...
      →the key
                 crypto_name = df['Name'][0] # Assumes the first entry represents_
      →the crypto
                 data[crypto_name] = df
         return data
     # Load data from the "archive" folder
     data folder = 'C:/Users/Nantia/Downloads/archive/'
     crypto_data = load_crypto_data(data_folder)
     # Display the keys of loaded data
     crypto_data.keys()
     # Function to clean and preprocess the data
     def clean_crypto_data(df):
         # Convert 'Date' to datetime format
         df['Date'] = pd.to_datetime(df['Date'])
         # Drop unnecessary columns if needed
         df.drop(columns=['SNo', 'Name', 'Symbol'], inplace=True)
         # Rename columns for consistency
         df.rename(columns={'High': 'high', 'Low': 'low', 'Open': 'open',
```

```
'Close': 'close', 'Volume': 'volume', 'Marketcap':
 ⇔'marketcap'}, inplace=True)
    # Handle missing values
   df.fillna(method='ffill', inplace=True)
   return df
# Clean all loaded data
cleaned_crypto_data = {name: clean_crypto_data(df) for name, df in crypto_data.
 →items()}
import matplotlib.pyplot as plt
import seaborn as sns
# Function to visualize price trends
def plot_price_trends(df, crypto_name):
   plt.figure(figsize=(14, 7))
   plt.plot(df['Date'], df['close'], label='Closing Price', color='blue')
   plt.title(f'Price Trends for {crypto_name}')
   plt.xlabel('Date')
   plt.ylabel('Price (USD)')
   plt.legend()
   plt.grid()
   plt.show()
# Plotting price trends for Bitcoin as an example
if 'Bitcoin' in cleaned_crypto_data:
   plot_price_trends(cleaned_crypto_data['Bitcoin'], 'Bitcoin')
# Function to visualize price trends for multiple cryptocurrencies
def plot_multiple_price_trends(cleaned_data, crypto_names):
   plt.figure(figsize=(14, 7))
   for crypto_name in crypto_names:
        if crypto_name in cleaned_data:
            df = cleaned_data[crypto_name]
            plt.plot(df['Date'], df['close'], label=crypto_name)
   plt.title('Price Trends for Selected Cryptocurrencies')
   plt.xlabel('Date')
   plt.ylabel('Price (USD)')
   plt.legend()
   plt.grid()
   plt.show()
# List of cryptocurrencies to plot
```

```
cryptocurrencies_to_plot = ['Bitcoin', 'Ethereum', 'Monero', 'Uniswap']

# Plotting price trends for selected cryptocurrencies
plot_multiple_price_trends(cleaned_crypto_data, cryptocurrencies_to_plot)
```





Explanation: MongoDB Connection: Connects to your online MongoDB database using the connection string. Data Loading Function: Reads all CSV files from a specified directory and combines them into a single DataFrame. MongoDB Insertion: Converts the DataFrame to a dictionary format

compatible with MongoDB and inserts it into the specified collection.

Explanation: Data Cleaning: Removes rows with null values and converts the Date column into a datetime format. Feature Engineering: Adds a new column for daily returns, which quantifies price changes.

Volume: Handling large datasets may require batch loading and optimized queries. Velocity: To address real-time data influx, consider implementing a streaming solution (e.g., using Apache Kafka). Variety: Different data formats (structured and unstructured) necessitate diverse processing techniques. For instance, you could incorporate sentiment analysis from social media.

```
[9]: from itertools import cycle
    import plotly.express as px
    Monero = pd.read_csv("C:/Users/Nantia/Downloads/archive/coin_Monero.csv")
    Monero
    projection_Monero = 5
    #creation of a new column with a name prediction
    Monero['Prediction'] = Monero[['Close']].shift(-projection_Monero)
    Monero
    visualize_Monero = cycle(['Open','Close','High','Low','Prediction'])
    fig = px.line(Monero, x=Monero.Date, y=[Monero['Open'], Monero['Close'],
                                            Monero['High'], Monero['Low'],

→Monero['Prediction']],
                labels={'Date': 'Date', 'value':'Price'})
    fig.update_layout(title_text='Monero', font_size=15,__
     fig.for_each_trace(lambda t: t.update(name = next(visualize_Monero)))
    fig.update_xaxes(showgrid=False)
    fig.update_yaxes(showgrid=False)
    fig.show()
```

```
[11]: from itertools import cycle
     import plotly.express as px
     Ethereum = pd.read_csv("C:/Users/Nantia/Downloads/archive/coin_Ethereum.csv")
     Ethereum
     projection Ethereum = 5
     #creation of a new column with a name prediction
     Ethereum['Prediction'] = Ethereum[['Close']].shift(-projection_Ethereum)
     Ethereum
     visualize_Ethereum= cycle(['Open','Close','High','Low','Prediction'])
     fig = px.line(Ethereum, x=Ethereum.Date, y=[Ethereum['Open'], Ethereum['Close'],
                                                Ethereum['High'], Ethereum['Low'],

→Ethereum['Prediction']],
                  labels={'Date': 'Date', 'value': 'Price'})
     fig.update layout(title text='Bitcoin', font size=15,__

¬font_color='black',legend_title_text='Parameters')

     fig.for_each_trace(lambda t: t.update(name = next(visualize Ethereum)))
     fig.update_xaxes(showgrid=False)
     fig.update_yaxes(showgrid=False)
     fig.show()
```

```
fig.for_each_trace(lambda t: t.update(name = next(visualize_Uniswap)))
fig.update_xaxes(showgrid=False)
fig.update_yaxes(showgrid=False)
fig.show()
```

[]: