o3tddztce

January 19, 2025

```
[33]: # 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()
```

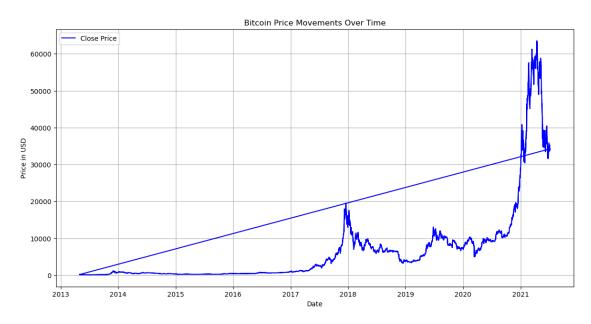
C:\Users\Nantia\anaconda3\Lib\site-packages\cryptography\x509\base.py:594: CryptographyDeprecationWarning:

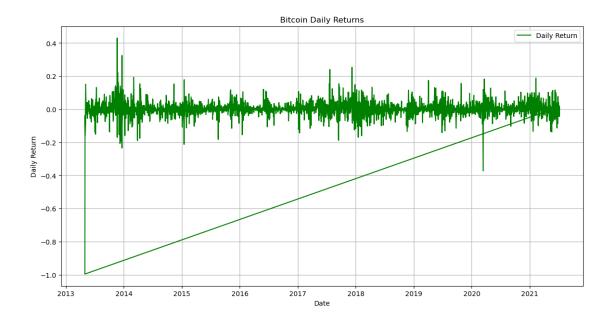
Parsed a negative serial number, which is disallowed by RFC 5280.

SNo Name Symbol Date High Low Open \

0	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



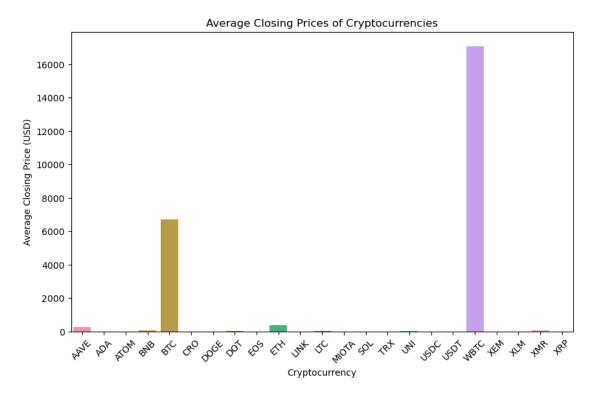


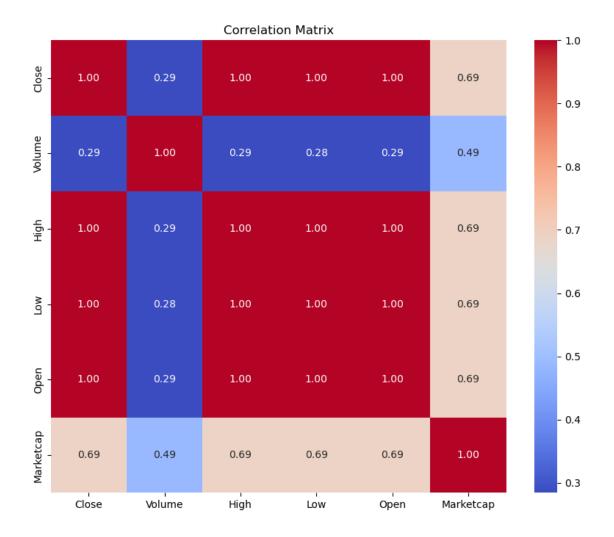
```
[7]: # Import necessary libraries
     import pandas as pd
     import numpy as np
     import matplotlib.pyplot as plt
     import seaborn as sns
     import glob
     import os
     from sklearn.model_selection import train_test_split
     from sklearn.linear_model import LinearRegression
     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 = []
     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!
[7]:
       SNo
            Name Symbol
                                        Date
                                                   High
                                                               Low
                                                                         Open \
    0
         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
                                                                    39.885262
    3
         4 Aave AAVE
                         2020-10-08 23:59:59 44.902511 36.696057
         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
[8]: # 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
    Name
                 0
    Symbol
                 0
    Date
                 0
                 0
    High
    Low
                 0
                 0
    Open
    Close
                 0
    Volume
```

```
Marketcap
    dtype: int64
    <class 'pandas.core.frame.DataFrame'>
    RangeIndex: 37082 entries, 0 to 37081
    Data columns (total 10 columns):
         Column
                    Non-Null Count Dtype
     0
         SNo
                    37082 non-null int64
     1
         Name
                    37082 non-null object
     2
                    37082 non-null object
         Symbol
                    37082 non-null datetime64[ns]
     3
         Date
     4
                    37082 non-null float64
         High
                    37082 non-null float64
     5
         Low
     6
                    37082 non-null float64
         Open
     7
                    37082 non-null float64
         Close
     8
         Volume
                    37082 non-null float64
         Marketcap 37082 non-null float64
    dtypes: datetime64[ns](1), float64(6), int64(1), object(2)
    memory usage: 2.8+ MB
[9]: # Example: Analyzing Average Closing Price for each cryptocurrency
     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
         MOTA
                   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
                  10.471388
    13
          SOL
```

```
14
      TRX
                0.032585
15
      UNI
               17.077256
16
     USDC
                1.003791
17
     USDT
                1.000696
           17086.573875
18
     WBTC
19
      XEM
                0.124662
                0.101509
20
      XLM
21
      XMR
               74.134773
22
      XRP
                0.234790
```





```
[11]: # Prepare data for regression model
    X = data[['Volume', 'High', 'Low', 'Open']]
    y = data['Close']

# Split the data into training and testing sets
    X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, \( \text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{
```

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

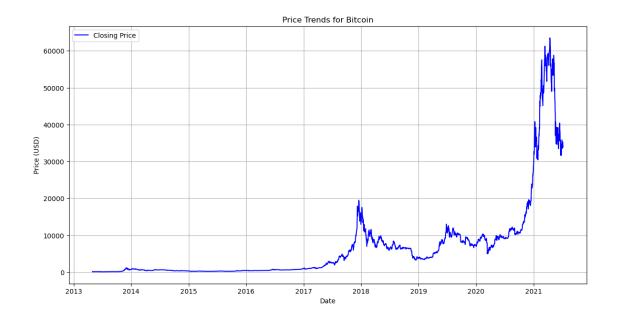
Mean Squared Error: 15899.032311721452

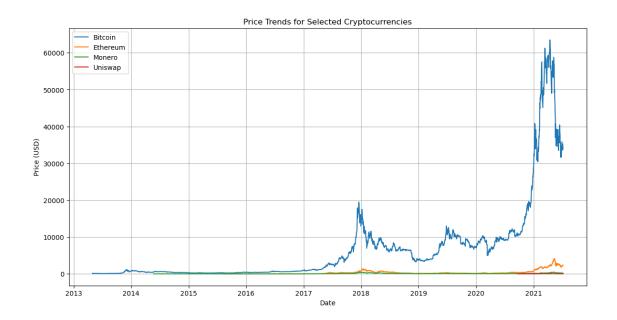
```
[28]: import pandas as pd
      import os
      # Function to load CSV files from a directory
      def load_crypto_data(folder_path):
          data = {}
          for filename in os.listdir(folder path):
              if filename.endswith('.csv'):
                  filepath = os.path.join(folder_path, filename)
                  # Load the data into a DataFrame
                  df = pd.read csv(filepath)
                  # Store DataFrame in a dictionary with the cryptocurrency name as u
       →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.

```
[27]: from itertools import cycle
      import plotly.express as px
      Monero = pd.read_csv("C:/Users/Nantia/Downloads/archive/coin_Monero.csv")
      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,__

¬font_color='black',legend_title_text='Parameters')
      fig.for_each_trace(lambda t: t.update(name = next(visualize_Monero)))
      fig.update_xaxes(showgrid=False)
      fig.update_yaxes(showgrid=False)
      fig.show()
```

```
[31]: from itertools import cycle
      import plotly.express as px
      Bitcoin = pd.read csv("C:/Users/Nantia/Downloads/archive/coin Bitcoin.csv")
      Bitcoin
      projection_Bitcoin = 5
      #creation of a new column with a name prediction
      Bitcoin['Prediction'] = Bitcoin[['Close']].shift(-projection_Bitcoin)
      Bitcoin
      visualize_Bitcoin = cycle(['Open','Close','High','Low','Prediction'])
      fig = px.line(Bitcoin, x=Bitcoin.Date, y=[Bitcoin['Open'], Bitcoin['Close'],
                                                Bitcoin['High'], Bitcoin['Low'],
       ⇔Bitcoin['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_Bitcoin)))
      fig.update xaxes(showgrid=False)
      fig.update_yaxes(showgrid=False)
```

```
fig.show()

[30]: from itertools import cycle
  import plotly.express as px

Ethereum = pd read csy("C:/Users/Nantia/Downloads/archive/coin Ethereum csy")
```

```
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)
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()
```

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

    Uniswap['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_Uniswap)))
      fig.update xaxes(showgrid=False)
      fig.update_yaxes(showgrid=False)
      fig.show()
```