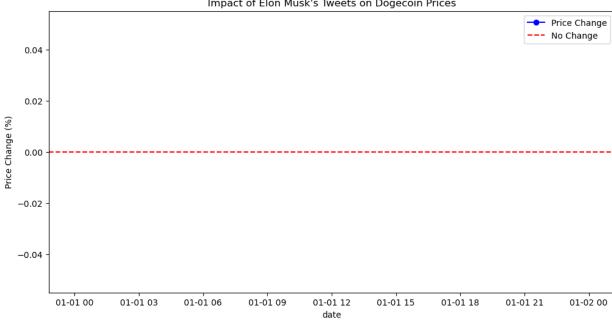
```
In [28]:
          import pandas as pd
          import numpy as np
          import matplotlib.pyplot as plt
          from datetime import datetime
In [29]:
          dogecoin_prices = pd.read_csv('DOGE-USD.csv')
          dogecoin prices
Out[29]:
                      Date
                               Open
                                          High
                                                    Low
                                                            Close
                                                                  Adj Close
                                                                                 Volume
             0
                 2017-11-09
                            0.001207
                                      0.001415
                                                0.001181
                                                         0.001415
                                                                   0.001415
                                                                               6259550.0
                 2017-11-10
                            0.001421
                                      0.001431
                                                0.001125
                                                         0.001163
                                                                   0.001163
              1
                                                                              4246520.0
             2
                  2017-11-11
                            0.001146
                                      0.001257
                                                0.001141
                                                         0.001201
                                                                   0.001201
                                                                               2231080.0
                 2017-11-12
                            0.001189
                                      0.001210
                                               0.001002
                                                         0.001038
                                                                   0.001038
                                                                              3288960.0
             4
                 2017-11-13
                            0.001046
                                      0.001212
                                                0.001019
                                                          0.001211
                                                                    0.001211
                                                                               2481270.0
          1756
                2022-08-31 0.061534
                                      0.063333
                                               0.061058
                                                         0.061330
                                                                   0.061330
                                                                            309748693.0
          1757
                2022-09-01 0.061336
                                      0.062479
                                               0.060194
                                                         0.062372
                                                                   0.062372
                                                                             328765413.0
          1758
                2022-09-02 0.062372
                                      0.062712
                                               0.060947
                                                         0.061635
                                                                   0.061635
                                                                             273453013.0
          1759
                2022-09-03
                                NaN
                                          NaN
                                                    NaN
                                                             NaN
                                                                       NaN
                                                                                    NaN
          1760 2022-09-04 0.062682 0.062744 0.062667 0.062696 0.062696
                                                                             297513408.0
         1761 rows × 7 columns
          dogecoin prices['date'] = pd.to datetime(dogecoin prices['Date'])
In [30]:
In [31]:
          elon tweets = pd.read csv('TweetsElonMusk.csv')
          elon tweets
```

3, 10.37 1 WI		110ject for Timeen								
Out[31]:		id	conversation_id	created_at	date	time	timezone			
	0	1381273474400800773	1381002894032347138	2021-04-11 18:50:33 EEST	2021- 04-11	18:50:33	300	4		
	1	1381273076709478403	1372444955050971142	2021-04-11 18:48:58 EEST	2021- 04-11	18:48:58	300	4		
	2	1381258144916008964	1381230136918433792	2021-04-11 17:49:38 EEST	2021- 04-11	17:49:38	300	4		
	3	1381221447322935303	1381221447322935303	2021-04-11 15:23:49 EEST	2021- 04-11	15:23:49	300	4		
	4	1381129584435818496	1381079981485252611	2021-04-11 09:18:47 EEST	2021- 04-11	09:18:47	300	4		
	•••									
	12557	1382255613665701894	1382189497694121990	2021-04- 14 11:53:14 EEST	2021- 04- 14	11:53:14	300	4		
	12558	1382239892445401089	1382189497694121990	2021-04- 14 10:50:45 EEST	2021- 04- 14	10:50:45	300	4		
	12559	1382239304097824768	1382189497694121990	2021-04- 14 10:48:25 EEST	2021- 04- 14	10:48:25	300	4		
	12560	1382131928619495429	1382046129450258434	2021-04- 14 03:41:45 EEST	2021- 04- 14	03:41:45	300	4		
	12561	1382052264802721792	1382046129450258434	2021-04- 13 22:25:11 EEST	2021- 04- 13	22:25:11	300	4		
	12562 rows × 36 columns									
In [32]:	[32]: elon_tweets['date'] = pd.to_datetime(elon_tweets['date'])									
In [33]:	elon_tweets['tweet']									

```
@vincent13031925 For now. Costs are decreasing...
Out[33]:
                                            Love this beautiful shot
         2
                  @agnostoxxx @CathieDWood @ARKInvest Trust the ...
         3
                                  The art In Cyberpunk is incredible
         4
                                                     @itsALLrisky 🥯 🤡
         12557
                  @eugenelee3 @PPathole @SpaceX @Tesla Yeah, not...
         12558
                  @PPathole @SpaceX @Tesla That was my night job...
         12559
                  @PPathole @SpaceX @Tesla True. Ancient times .....
         12560
                                   @Erdayastronaut @Tesla Absolutely
         12561
                  @Erdayastronaut @Tesla Tesla is building up co...
         Name: tweet, Length: 12562, dtype: object
 In [ ]:
In [34]:
         dogecoin tweets = elon tweets[elon tweets['tweet'].str.contains('Dogecoin')]
In [35]:
         time window = 6
In [62]: import pandas as pd
         import numpy as np
         import matplotlib.pyplot as plt
         from datetime import datetime
         # Filter tweets that mention Dogecoin
         dogecoin tweets = elon tweets[elon tweets['tweet'].str.contains('DOGE')]
         # Define the time window to analyze price changes after a tweet (in hours)
         time window = 6
         # Create a new column to store the price change after each tweet
         dogecoin tweets['Price Change'] = np.nan
         # Iterate through each tweet
         for index, tweet in dogecoin tweets.iterrows():
             tweet date = tweet['date']
             tweet close = dogecoin prices.loc[dogecoin prices['date'] == tweet date, '(
             future closes = dogecoin prices.loc[(dogecoin prices['date'] > tweet date)
                                                  (dogecoin prices['date'] <= tweet_date</pre>
                                                   'Close'1
              if not future closes.empty:
                 price_change = (future_closes.iloc[-1] - tweet_close) / tweet_close * 1
                 dogecoin tweets.loc[index, 'Price Change'] = price change
         # Plot the price change after each tweet
         plt.figure(figsize=(12, 6))
         plt.plot(dogecoin_tweets['date'], dogecoin_tweets['Price_Change'], 'bo-', label
         plt.axhline(0, color='r', linestyle='--', label='No Change')
         plt.xlabel('date')
         plt.ylabel('Price Change (%)')
         plt.title("Impact of Elon Musk's Tweets on Dogecoin Prices")
         plt.legend()
         plt.show()
```

Impact of Elon Musk's Tweets on Dogecoin Prices



```
In [72]: filtered_prices = dogecoin_prices[dogecoin_prices['Date'] < pd.to_datetime('201</pre>
         filtered prices['daily returns'] = filtered prices['Close'].pct change()
         /var/folders/xh/qqlnwhcn7c7cv5c7s2s5d59h0000gn/T/ipykernel_21919/686570824.py:
         2: SettingWithCopyWarning:
         A value is trying to be set on a copy of a slice from a DataFrame.
         Try using .loc[row indexer,col indexer] = value instead
         See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/st
         able/user_guide/indexing.html#returning-a-view-versus-a-copy
           filtered prices['daily returns'] = filtered prices['Close'].pct change()
```

```
In [73]: import numpy as np
         # Parameters for the simulation
         num simulations = 1000
         num days = 365 # Number of days to simulate into the future
         # Calculate mean and standard deviation of historical daily returns
         mean return = filtered prices['daily returns'].mean()
         std return = filtered prices['daily returns'].std()
         # Generate random samples for each simulation
         simulations = np.random.normal(loc=mean return, scale=std return, size=(num day
         # Calculate the cumulative sum of returns for each simulation
         cumulative returns = (simulations + 1).cumprod(axis=0)
```

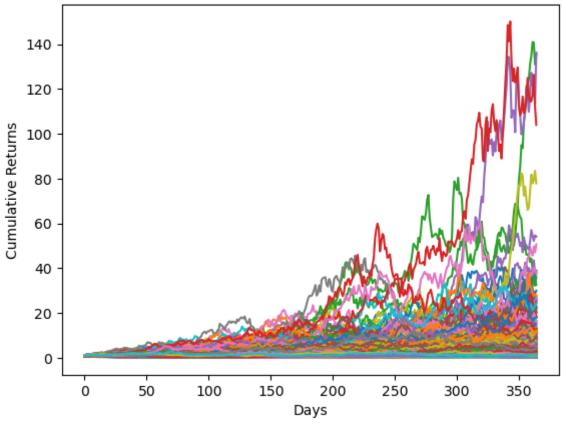
```
In [74]: import numpy as np
         # Parameters for the simulation
         num simulations = 1000
         num days = 365  # Number of days to simulate into the future
         # Calculate mean and standard deviation of historical daily returns
         mean return = filtered prices['daily returns'].mean()
         std return = filtered prices['daily returns'].std()
```

```
# Generate random samples for each simulation
simulations = np.random.normal(loc=mean_return, scale=std_return, size=(num_day)
# Calculate the cumulative sum of returns for each simulation
cumulative_returns = (simulations + 1).cumprod(axis=0)
```

```
In [75]: import matplotlib.pyplot as plt

# Plot the cumulative returns for each simulation
plt.plot(cumulative_returns)
plt.xlabel('Days')
plt.ylabel('Cumulative Returns')
plt.title('Monte Carlo Simulation - Dogecoin')
plt.show()
```

Monte Carlo Simulation - Dogecoin



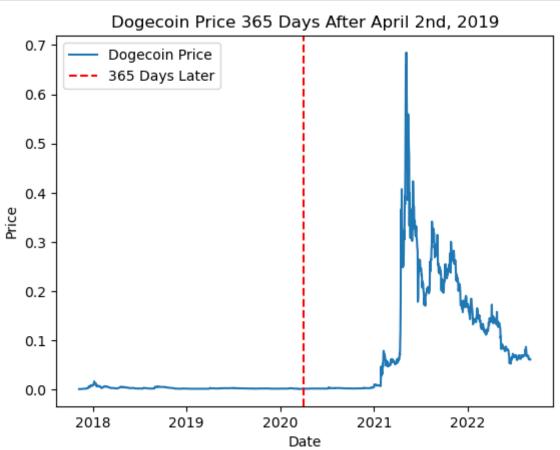
```
In [41]: # Calculate the average price for each simulation day
    average_prices = cumulative_returns.mean(axis=1)

# Calculate the standard deviation of prices for each simulation day
    std_prices = cumulative_returns.std(axis=1)
```

```
In [42]: average_prices.mean()
Out[42]: 2.123764687918773
In [43]: std_prices.std()
```

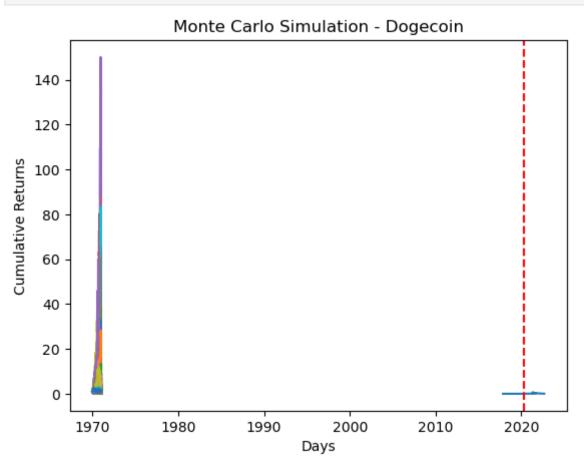
Out[43]: 2.2965862451047667

```
In [67]:
         import pandas as pd
         import matplotlib.pyplot as plt
         # Load the Dogecoin price data
         dogecoin_prices = pd.read_csv('DOGE-USD.csv')
         # Convert the date column to datetime format
         dogecoin_prices['Date'] = pd.to_datetime(dogecoin_prices['Date'])
         # Filter the dataset to include data from April 2nd, 2019
         filtered_prices = dogecoin_prices[dogecoin_prices['Date'] == pd.to_datetime('20
         # Calculate the target date after 365 days
         target_date = filtered_prices['Date'] + pd.DateOffset(days=365)
         # Get the price exactly 365 days after April 2nd, 2019
         target price = dogecoin prices[dogecoin prices['Date'] == target date.iloc[0]][
         # Plot the price
         plt.plot(dogecoin_prices['Date'], dogecoin_prices['Close'], label='Dogecoin_Pri
         plt.axvline(target_date.iloc[0], color='r', linestyle='--', label='365 Days Lat
         # Set the plot title and labels
         plt.title("Dogecoin Price 365 Days After April 2nd, 2019")
         plt.xlabel("Date")
         plt.ylabel("Price")
         # Display the legend and show the plot
         plt.legend()
         plt.show()
```



```
In [76]: import matplotlib.pyplot as plt

# Plot the cumulative returns for each simulation
plt.plot(dogecoin_prices['Date'], dogecoin_prices['Close'], label='Dogecoin Pri
plt.axvline(target_date.iloc[0], color='r', linestyle='--', label='365 Days Lat
plt.plot(cumulative_returns)
plt.xlabel('Days')
plt.ylabel('Cumulative Returns')
plt.title('Monte Carlo Simulation - Dogecoin')
plt.show()
```



```
In [66]: import pandas as pd
import matplotlib.pyplot as plt

# Load the Dogecoin price data
#dogecoin_prices = pd.read_csv('dogecoin_prices.csv')

# Convert the date column to datetime format
dogecoin_prices['Date'] = pd.to_datetime(dogecoin_prices['Date'])

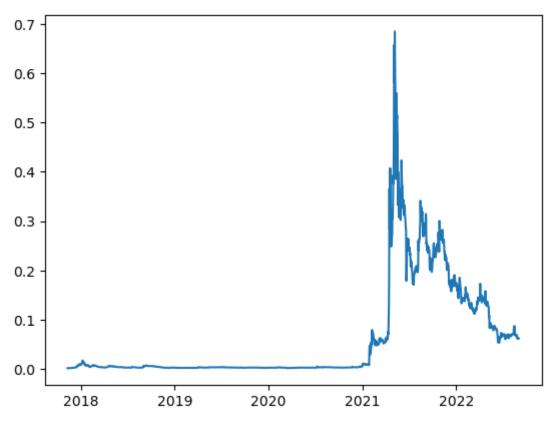
# Load the Elon Musk's tweets data
#elon_tweets = pd.read_csv('elon_tweets.csv')

# Convert the date column to datetime format
elon_tweets['date'] = pd.to_datetime(elon_tweets['date'])

# Filter Elon Musk's tweets mentioning Dogecoin
dogecoin_tweets = elon_tweets[elon_tweets['tweet'].str.contains('Dogecoin')]

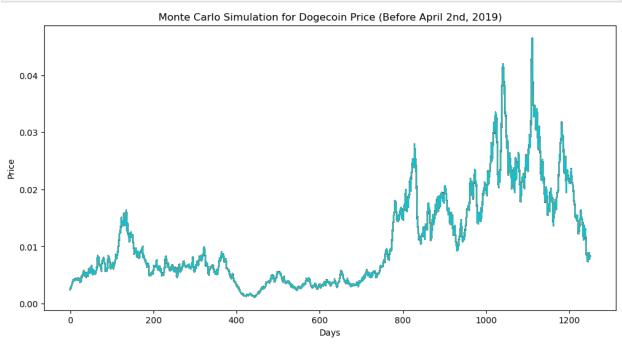
# Plot the Dogecoin price
```

```
KeyError
                                          Traceback (most recent call last)
File ~/Desktop/UCI/anaconda3/lib/python3.10/site-packages/pandas/core/indexes/
base.py:3802, in Index.get loc(self, key, method, tolerance)
   3801 try:
            return self._engine.get_loc(casted_key)
-> 3802
   3803 except KeyError as err:
File ~/Desktop/UCI/anaconda3/lib/python3.10/site-packages/pandas/ libs/index.p
yx:138, in pandas. libs.index.IndexEngine.get loc()
File ~/Desktop/UCI/anaconda3/lib/python3.10/site-packages/pandas/ libs/index.p
yx:165, in pandas. libs.index.IndexEngine.get loc()
File pandas/_libs/hashtable_class_helper.pxi:5745, in pandas._libs.hashtable.P
yObjectHashTable.get item()
File pandas/ libs/hashtable class helper.pxi:5753, in pandas. libs.hashtable.P
yObjectHashTable.get item()
KeyError: 'Date'
The above exception was the direct cause of the following exception:
KeyError
                                          Traceback (most recent call last)
Cell In[66], line 23
     20 plt.plot(dogecoin_prices['Date'], dogecoin_prices['Close'], label='Dog
ecoin Price')
     22 # Plot Elon Musk's tweets
---> 23 plt.scatter(dogecoin_tweets['Date'], dogecoin prices.loc[dogecoin twee
ts['Date']]['Close'],
                    marker='o', color='r', label="Elon Musk's Tweets")
     26 # Set the plot title and labels
     27 plt.title("Impact of Elon Musk's Tweets on Dogecoin Price")
File ~/Desktop/UCI/anaconda3/lib/python3.10/site-packages/pandas/core/frame.p
y:3807, in DataFrame. getitem (self, key)
   3805 if self.columns.nlevels > 1:
            return self. getitem multilevel(key)
-> 3807 indexer = self.columns.get loc(key)
   3808 if is integer(indexer):
   3809
            indexer = [indexer]
File ~/Desktop/UCI/anaconda3/lib/python3.10/site-packages/pandas/core/indexes/
base.py:3804, in Index.get loc(self, key, method, tolerance)
   3802
            return self. engine.get loc(casted key)
   3803 except KeyError as err:
-> 3804
            raise KeyError(key) from err
   3805 except TypeError:
            # If we have a listlike key, check indexing error will raise
   3806
   3807
            # InvalidIndexError. Otherwise we fall through and re-raise
   3808
            # the TypeError.
   3809
            self. check indexing error(key)
KeyError: 'Date'
```



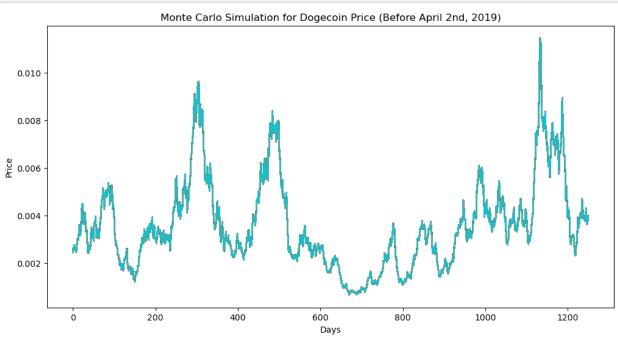
```
In [78]:
         import pandas as pd
         import numpy as np
         import matplotlib.pyplot as plt
         # Load the Dogecoin price data
         #dogecoin prices = pd.read csv('dogecoin prices.csv')
         # Filter the data before April 2nd, 2019
         start date = pd.to datetime('2019-04-02')
         filtered prices = dogecoin prices[dogecoin prices['Date'] < start date]</pre>
         # Select the 'Close' prices
         prices = filtered_prices['Close'].values
         # Calculate the daily returns
         returns = np.diff(prices) / prices[:-1]
         # Calculate the mean and standard deviation of daily returns
         mean return = np.mean(returns)
         std_return = np.std(returns)
         # Set the simulation parameters
         num simulations = 5000
         num days = 1251
         # Perform the Monte Carlo simulation
         simulations = np.zeros((num_days, num_simulations))
         simulations[0, :] = prices[-1]
         for i in range(1, num days):
             drift = 1 + mean return
             shock = std_return * np.random.normal()
              simulations[i, :] = simulations[i-1, :] * (drift + shock)
```

```
# Plot the Monte Carlo simulations
plt.figure(figsize=(12, 6))
plt.plot(simulations)
plt.title("Monte Carlo Simulation for Dogecoin Price (Before April 2nd, 2019)")
plt.xlabel("Days")
plt.ylabel("Price")
plt.show()
```



```
In [86]: import pandas as pd
         import numpy as np
         import matplotlib.pyplot as plt
         # Load the Dogecoin price data
         #dogecoin prices = pd.read csv('dogecoin prices.csv')
         # Filter the data before April 2nd, 2019
         start date = pd.to datetime('2019-04-02')
         filtered_prices = dogecoin_prices[dogecoin_prices['Date'] < start_date]</pre>
         # Select the 'Close' prices
         prices = filtered prices['Close'].values
         # Calculate the daily returns
         returns = np.diff(prices) / prices[:-1]
         # Calculate the mean and standard deviation of daily returns
         mean return = np.mean(returns)
         std return = np.std(returns)
         # Set the simulation parameters
         num simulations = 5000
         num days = 1251
         # Perform the Monte Carlo simulation
         simulations = np.zeros((num days, num simulations))
         simulations[0, :] = prices[-1]
```

```
for i in range(1, num days):
    drift = 1 + mean_return
    shock = std_return * np.random.normal()
    simulations[i, :] = simulations[i-1, :] * (drift + shock)
# Plot the Monte Carlo simulations
plt.figure(figsize=(12, 6))
plt.plot(simulations)
plt.title("Monte Carlo Simulation for Dogecoin Price (Before April 2nd, 2019)")
plt.xlabel("Days")
plt.ylabel("Price")
plt.show()
import numpy as np
# Perform the Monte Carlo simulation (assuming you already have this code)
simulations = np.zeros((num_days, num_simulations))
simulations[0, :] = prices[-1]
# ... Simulation code ...
# Calculate the mean and standard deviation of the simulated prices
```



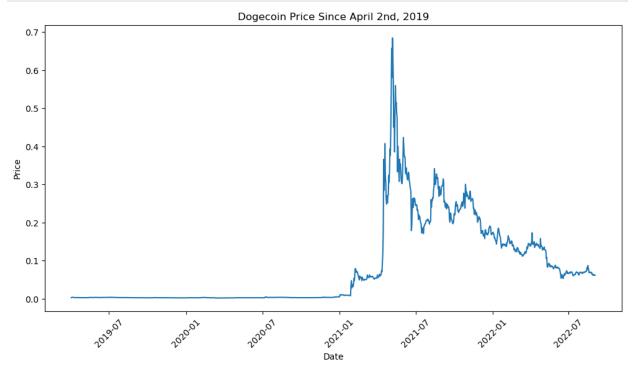
```
In [84]: import numpy as np

# Perform the Monte Carlo simulation (assuming you already have this code)
simulations = np.zeros((num_days, num_simulations))
simulations[0, :] = prices[-1]
# ... Simulation code ...

# Calculate the mean and standard deviation of the simulated prices
mean_simulation = np.mean(simulations[-1, :])
std_simulation = np.std(simulations[-1, :])
print("Mean Simulation:", mean_simulation)
print("Standard Deviation:", std_simulation)
```

Mean Simulation: 0.0 Standard Deviation: 0.0

```
In [82]:
         import pandas as pd
         import matplotlib.pyplot as plt
         # Load the Dogecoin price data
         #dogecoin_prices = pd.read_csv('dogecoin_prices.csv')
         # Convert the date column to datetime format
         dogecoin prices['Date'] = pd.to datetime(dogecoin prices['Date'])
         # Filter the data since April 2nd, 2019
         start_date = pd.to_datetime('2019-04-02')
         filtered prices = dogecoin prices[dogecoin prices['Date'] >= start date]
         # Plot the Dogecoin price
         plt.figure(figsize=(12, 6))
         plt.plot(filtered_prices['Date'], filtered_prices['Close'])
         plt.title("Dogecoin Price Since April 2nd, 2019")
         plt.xlabel("Date")
         plt.ylabel("Price")
         plt.xticks(rotation=45)
         plt.show()
```



```
In [83]: import pandas as pd
import numpy as np

# Load the Dogecoin price data
#dogecoin_prices = pd.read_csv('dogecoin_prices.csv')

# Convert the date column to datetime format
dogecoin_prices['Date'] = pd.to_datetime(dogecoin_prices['Date'])

# Filter the data since April 2nd, 2019
start_date = pd.to_datetime('2019-04-02')
filtered_prices = dogecoin_prices[dogecoin_prices['Date'] >= start_date]

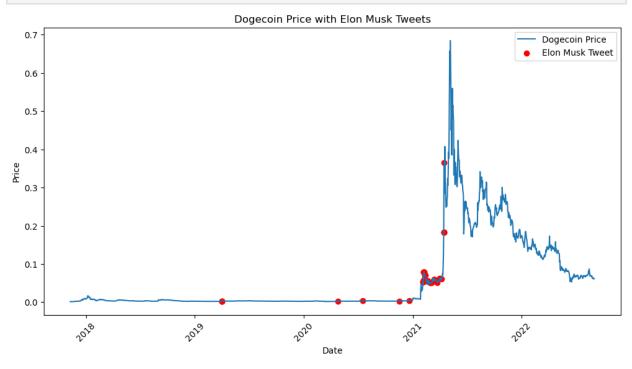
# Calculate the mean and standard deviation of Dogecoin's price
```

mean price = np.mean(filtered prices['Close'])

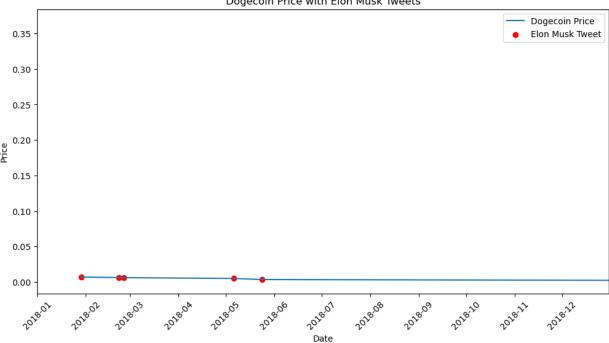
```
std price = np.std(filtered prices['Close'])
         print("Mean Price:", mean price)
         print("Standard Deviation:", std_price)
         Mean Price: 0.08229928457234212
         Standard Deviation: 0.11256315574009294
In [93]: import pandas as pd
          # Load the Dogecoin price data
          #dogecoin_prices = pd.read_csv('dogecoin_prices.csv')
          # Load Elon Musk's tweets data
          #elon tweets = pd.read csv('elon tweets.csv')
          # Convert the date columns to datetime format
         dogecoin_prices['date'] = pd.to_datetime(dogecoin_prices['Date'])
         elon_tweets['date'] = pd.to_datetime(elon_tweets['date'])
          # Filter Elon Musk's tweets mentioning Dogecoin and additional keywords
         keywords = ['Dogecoin', 'crypto', 'DOGE', 'digital currency']
          dogecoin_tweets = elon_tweets[elon_tweets['tweet'].str.contains('|'.join(keywork))
          # Merge Dogecoin price and Elon Musk's tweets based on date
         merged data = pd.merge(dogecoin prices, dogecoin tweets, on='date')
         merged data = merged data.dropna()
         correlation coefficient = merged data['Close'].corr(merged data['retweet'])
         print("Correlation Coefficient:", correlation coefficient)
         Correlation Coefficient: nan
In [102... import pandas as pd
         import matplotlib.pyplot as plt
          # Load the Dogecoin price data
          #dogecoin prices = pd.read csv('dogecoin prices.csv')
          # Convert the date column to datetime format
          dogecoin prices['date'] = pd.to datetime(dogecoin prices['Date'])
          # Load Elon Musk's tweets data
          #elon tweets = pd.read csv('elon tweets.csv')
          # Convert the date column to datetime format
         elon tweets['date'] = pd.to datetime(elon tweets['date'])
         keywords = ['Dogecoin', 'DOGE',]
         dogecoin tweets = elon tweets[elon tweets['tweet'].str.contains('|'.join(keywork))
          # Merge Dogecoin prices with Elon Musk tweets based on date
         merged data = pd.merge(dogecoin prices, dogecoin tweets, on='date', how='inner'
```

Plot the Dogecoin price with markers for tweet dates

```
plt.figure(figsize=(12, 6))
plt.plot(dogecoin_prices['date'], dogecoin_prices['Close'], label='Dogecoin Pri
plt.scatter(merged_data['date'], merged_data['Close'], color='red', label='Elor
plt.title("Dogecoin Price with Elon Musk Tweets")
plt.xlabel("Date")
plt.ylabel("Price")
plt.legend()
plt.xticks(rotation=45)
plt.show()
```



Dogecoin Price with Elon Musk Tweets

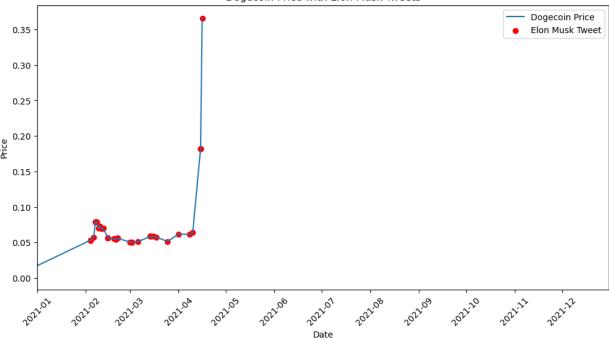


```
In [99]: plt.figure(figsize=(12, 6))
    plt.plot(merged_data['date'], merged_data['Close'], label='Dogecoin Price')
    plt.scatter(merged_data['date'], merged_data['Close'], color='red', label='Elor
    plt.title("Dogecoin Price with Elon Musk Tweets")
    plt.xlabel("Date")
    plt.ylabel("Price")
    plt.legend()
    plt.sticks(rotation=45)

# Specify the year to zoom in on
    year_to_zoom = 2021

# Set the x-axis limits to the desired time range
    plt.xlim(pd.Timestamp(year=year_to_zoom, month=1, day=1), pd.Timestamp(year=year_year)
    plt.show()
```



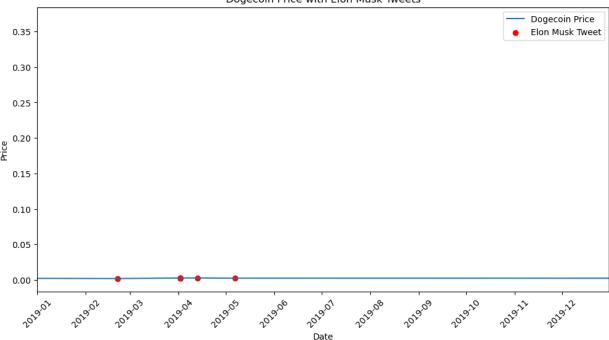


```
In [100... plt.figure(figsize=(12, 6))
    plt.plot(merged_data['date'], merged_data['Close'], label='Dogecoin Price')
    plt.scatter(merged_data['date'], merged_data['Close'], color='red', label='Elor
    plt.title("Dogecoin Price with Elon Musk Tweets")
    plt.xlabel("Date")
    plt.ylabel("Price")
    plt.legend()
    plt.sticks(rotation=45)

# Specify the year to zoom in on
    year_to_zoom = 2019

# Set the x-axis limits to the desired time range
    plt.xlim(pd.Timestamp(year=year_to_zoom, month=1, day=1), pd.Timestamp(year=year_plt.show())
```

Dogecoin Price with Elon Musk Tweets



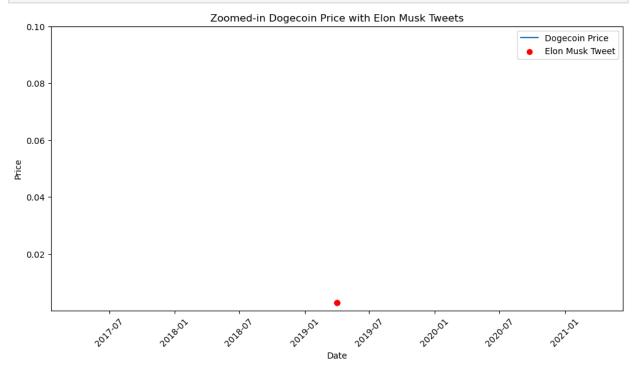
```
In [105...
         import pandas as pd
          import matplotlib.pyplot as plt
          # Load the Dogecoin price data
          #dogecoin prices = pd.read csv('dogecoin prices.csv')
          # Convert the date column to datetime format
         dogecoin prices['date'] = pd.to datetime(dogecoin prices['Date'])
          # Load Elon Musk's tweets data
          #elon tweets = pd.read csv('elon tweets.csv')
          # Convert the date column to datetime format
         elon tweets['date'] = pd.to datetime(elon tweets['date'])
          # Filter Elon Musk's tweets mentioning Dogecoin
         keywords = ['Dogecoin', 'DOGE',]
         dogecoin tweets = elon tweets[elon tweets['tweet'].str.contains('|'.join(keywork))
          # Merge Dogecoin prices with Elon Musk tweets based on date
         merged_data = pd.merge(dogecoin_prices, dogecoin_tweets, on='date', how='inner'
          # Specify the time range to zoom in on
         start date = pd.Timestamp(year=2019, month=1, day=1)
         end date = pd.Timestamp(year=2019, month=12, day=31)
          # Filter the merged data within the specified time range
          zoomed data = merged data[(merged data['date'] >= start date) & (merged data['date']
         # Plot the Dogecoin price with markers for tweet dates
         plt.figure(figsize=(12, 6))
         plt.plot(zoomed data['date'], zoomed data['Close'], label='Dogecoin Price')
         plt.scatter(zoomed_data['date'], zoomed_data['Close'], color='red', label='Elor
         plt.title("Zoomed-in Dogecoin Price with Elon Musk Tweets")
         plt.xlabel("Date")
         plt.ylabel("Price")
         plt.legend()
```

```
plt.xticks(rotation=45)

# Specify the y-axis limits for price zoom
price_lower_limit = 0.0001 # Set the lower limit of the price
price_upper_limit = 0.1 # Set the upper limit of the price

# Set the y-axis limits to the desired price range
plt.ylim(price_lower_limit, price_upper_limit)

plt.show()
```



```
In [107...
         import pandas as pd
          import numpy as np
          from sklearn.linear model import LinearRegression
          # Load the Dogecoin price data
          #dogecoin prices = pd.read csv('dogecoin prices.csv')
          # Convert the date column to datetime format
         dogecoin prices['date'] = pd.to datetime(dogecoin prices['Date'])
          # Load Elon Musk's tweets data
          #elon tweets = pd.read csv('elon tweets.csv')
          # Convert the date column to datetime format
         elon tweets['date'] = pd.to datetime(elon tweets['date'])
         keywords = ['Dogecoin', 'DOGE',]
         dogecoin_tweets = elon_tweets[elon_tweets['tweet'].str.contains('|'.join(keywork))
          # Merge Dogecoin prices with Elon Musk tweets based on date
         merged data = pd.merge(dogecoin prices, dogecoin tweets, on='date', how='inner
          # Create a feature matrix X with the tweet data
         X = merged data['sentiment score'].values.reshape(-1, 1) # Use the sentiment of
```

```
# Create a target variable y with the Dogecoin prices
y = merged_data['Close'].values
# Create a linear regression model and fit it to the data
regression_model = LinearRegression()
regression_model.fit(X, y)
# Print the coefficients of the linear regression model
print("Intercept:", regression_model.intercept_)
print("Coefficient:", regression_model.coef_[0])
# Predict Dogecoin prices based on the tweet data
predicted_prices = regression_model.predict(X)
# Plot the actual Dogecoin prices and the predicted prices
plt.figure(figsize=(12, 6))
plt.plot(merged_data['date'], y, label='Actual Prices')
plt.plot(merged_data['date'], predicted_prices, label='Predicted Prices')
plt.title("Dogecoin Prices: Actual vs. Predicted")
plt.xlabel("Date")
plt.ylabel("Price")
plt.legend()
plt.xticks(rotation=45)
plt.show()
```

```
KeyError
                                                    Traceback (most recent call last)
         File ~/Desktop/UCI/anaconda3/lib/python3.10/site-packages/pandas/core/indexes/
         base.py:3802, in Index.get loc(self, key, method, tolerance)
                     return self._engine.get_loc(casted_key)
         -> 3802
            3803 except KeyError as err:
         File ~/Desktop/UCI/anaconda3/lib/python3.10/site-packages/pandas/ libs/index.p
         yx:138, in pandas. libs.index.IndexEngine.get loc()
         File ~/Desktop/UCI/anaconda3/lib/python3.10/site-packages/pandas/ libs/index.p
         yx:165, in pandas. libs.index.IndexEngine.get loc()
         File pandas/_libs/hashtable_class_helper.pxi:5745, in pandas._libs.hashtable.P
         yObjectHashTable.get item()
         File pandas/ libs/hashtable class helper.pxi:5753, in pandas. libs.hashtable.P
         yObjectHashTable.get item()
         KeyError: 'sentiment score'
         The above exception was the direct cause of the following exception:
         KeyError
                                                    Traceback (most recent call last)
         Cell In[107], line 26
              23 merged_data = pd.merge(dogecoin_prices, dogecoin_tweets, on='date', ho
         w='inner')
              25 # Create a feature matrix X with the tweet data
         ---> 26 X = merged data['sentiment score'].values.reshape(-1, 1) # Use the se
         ntiment compound score as the feature
              28 # Create a target variable y with the Dogecoin prices
              29 y = merged_data['Close'].values
         File ~/Desktop/UCI/anaconda3/lib/python3.10/site-packages/pandas/core/frame.p
         y:3807, in DataFrame. getitem (self, key)
            3805 if self.columns.nlevels > 1:
                     return self. getitem multilevel(key)
         -> 3807 indexer = self.columns.get loc(key)
            3808 if is integer(indexer):
            3809
                     indexer = [indexer]
         File ~/Desktop/UCI/anaconda3/lib/python3.10/site-packages/pandas/core/indexes/
         base.py:3804, in Index.get loc(self, key, method, tolerance)
                     return self._engine.get_loc(casted_key)
            3802
            3803 except KeyError as err:
                     raise KeyError(key) from err
            3805 except TypeError:
                     # If we have a listlike key, check indexing error will raise
            3806
            3807
                     # InvalidIndexError. Otherwise we fall through and re-raise
            3808
                     # the TypeError.
            3809
                     self. check indexing error(key)
         KeyError: 'sentiment score'
In [108... merged data
```

Out[108]:

	Date	Open	High	Low	Close	Adj Close	Volume	date	
0	2019- 04- 02	0.002459	0.002863	0.002394	0.002795	0.002795	6.029836e+07	2019- 04- 02	111317
1	2019- 04- 02	0.002459	0.002863	0.002394	0.002795	0.002795	6.029836e+07	2019- 04- 02	111317
2	2019- 04- 02	0.002459	0.002863	0.002394	0.002795	0.002795	6.029836e+07	2019- 04- 02	111316
3	2019- 04- 02	0.002459	0.002863	0.002394	0.002795	0.002795	6.029836e+07	2019- 04- 02	111300
4	2020- 04- 25	0.002102	0.002146	0.002087	0.002142	0.002142	2.298104e+08	2020- 04- 25	1254039
5	2020- 07-18	0.003060	0.003629	0.003031	0.003473	0.003473	2.040809e+08	2020- 07-18	128429
6	2020- 11-17	0.002908	0.002976	0.002886	0.002936	0.002936	4.663146e+07	2020- 11-17	132877
7	2020- 12-20	0.003926	0.004678	0.003827	0.004625	0.004625	5.080660e+08	2020- 12-20	134059(
8	2021- 02- 04	0.037226	0.057869	0.035945	0.053289	0.053289	1.304084e+10	2021- 02- 04	135724
9	2021- 02- 04	0.037226	0.057869	0.035945	0.053289	0.053289	1.304084e+10	2021- 02- 04	135724
10	2021- 02- 04	0.037226	0.057869	0.035945	0.053289	0.053289	1.304084e+10	2021- 02- 04	135723
11	2021- 02- 06	0.046931	0.058308	0.044904	0.057595	0.057595	5.946101e+09	2021- 02- 06	135797
12	2021- 02-07	0.057502	0.084357	0.054239	0.078782	0.078782	1.426102e+10	2021- 02-07	135824
13	2021- 02-07	0.057502	0.084357	0.054239	0.078782	0.078782	1.426102e+10	2021- 02-07	135824
14	2021- 02- 08	0.078352	0.084945	0.064702	0.078825	0.078825	1.284438e+10	2021- 02- 08	135864

	Date	Open	High	Low	Close	Adj Close	Volume	date	
15	2021- 02- 08	0.078352	0.084945	0.064702	0.078825	0.078825	1.284438e+10	2021- 02- 08	135854;
16	2021- 02-10	0.070111	0.081091	0.068525	0.072896	0.072896	6.785088e+09	2021- 02-10	135951
17	2021- 02-11	0.072844	0.074301	0.068290	0.069676	0.069676	3.818557e+09	2021- 02-11	135979
18	2021- 02-12	0.069650	0.072610	0.061445	0.070069	0.070069	4.190844e+09	2021- 02-12	13600(
19	2021- 02-15	0.062568	0.063924	0.048547	0.056591	0.056591	4.944805e+09	2021- 02-15	1361096
20	2021- 02-15	0.062568	0.063924	0.048547	0.056591	0.056591	4.944805e+09	2021- 02-15	136109
21	2021- 02-20	0.055132	0.060286	0.051628	0.054384	0.054384	3.175469e+09	2021- 02-20	136306
22	2021- 02-20	0.055132	0.060286	0.051628	0.054384	0.054384	3.175469e+09	2021- 02-20	136304
23	2021- 02-21	0.054369	0.058428	0.053556	0.055980	0.055980	2.450293e+09	2021- 02-21	136360
24	2021- 03-01	0.048070	0.051479	0.048029	0.050599	0.050599	1.494427e+09	2021- 03-01	136647
25	2021- 03- 02	0.050596	0.052382	0.049299	0.050262	0.050262	1.346282e+09	2021- 03- 02	136685
26	2021- 03- 02	0.050596	0.052382	0.049299	0.050262	0.050262	1.346282e+09	2021- 03- 02	136666
27	2021- 03- 06	0.049601	0.052397	0.049383	0.050984	0.050984	1.480482e+09	2021- 03- 06	136805
28	2021- 03-14	0.062384	0.063052	0.058592	0.058592	0.058592	2.752133e+09	2021- 03-14	137088
29	2021- 03-14	0.062384	0.063052	0.058592	0.058592	0.058592	2.752133e+09	2021- 03-14	137088
30	2021- 03-14	0.062384	0.063052	0.058592	0.058592	0.058592	2.752133e+09	2021- 03-14	137088
31	2021- 03-16	0.057086	0.058921	0.055493	0.058607	0.058607	1.393772e+09	2021- 03-16	137160

	Date	Open	High	Low	Close	Adj Close	Volume	date	
32	2021- 03-18	0.057640	0.058828	0.054143	0.057383	0.057383	9.097774e+08	2021- 03-18	137265
33	2021- 03- 25	0.051699	0.052407	0.049697	0.051448	0.051448	1.084214e+09	2021- 03- 25	137498
34	2021- 04-01	0.053655	0.070111	0.053644	0.061986	0.061986	5.816047e+09	2021- 04-01	137756
35	2021- 04- 08	0.059036	0.061745	0.058817	0.061464	0.061464	1.055258e+09	2021- 04- 08	138026
36	2021- 04-15	0.121167	0.187326	0.120736	0.182207	0.182207	1.791662e+10	2021- 04-15	138255
37	2021- 04-15	0.121167	0.187326	0.120736	0.182207	0.182207	1.791662e+10	2021- 04-15	13825
38	2021- 04-15	0.121167	0.187326	0.120736	0.182207	0.182207	1.791662e+10	2021- 04-15	138246
39	2021- 04-15	0.121167	0.187326	0.120736	0.182207	0.182207	1.791662e+10	2021- 04-15	138245
40	2021- 04-16	0.181587	0.437700	0.180488	0.365870	0.365870	6.941068e+10	2021- 04-16	13828(

41 rows \times 43 columns

In []: