

gold-price-prediction

June 26, 2024

#GOLD PRICE PREDICTION

```
[1]: # Importing the pandas library for data manipulation and analysis
import pandas as pd

# Importing the numpy library for numerical operations
import numpy as np

# Importing the matplotlib library for plotting and visualization
import matplotlib.pyplot as plt

# Importing the seaborn library for statistical data visualization
import seaborn as sns
```

```
[2]: # Load the dataset (assuming it's stored in a CSV file)
gold_data = pd.read_csv('/content/goldstock.csv')
```

```
[3]: # Display the dimensions (number of rows and columns) of the DataFrame
↳ 'gold_data'
gold_data.shape
```

[3]: (2511, 7)

```
[4]: # Display the first five rows of the DataFrame 'gold_data'
gold_data.head()
```

```
[4]:
```

	no	Date	Close	Volume	Open	High	Low
0	0	1/19/2024	2029.3	166078	2027.4	2041.9	2022.2
1	1	1/18/2024	2021.6	167013	2009.1	2025.6	2007.7
2	2	1/17/2024	2006.5	245194	2031.7	2036.1	2004.6
3	3	1/16/2024	2030.2	277995	2053.4	2062.8	2027.6
4	4	1/12/2024	2051.6	250946	2033.2	2067.3	2033.1

```
[5]: # Display the data types of each column in the DataFrame 'gold_data'
gold_data.dtypes
```

```
[5]: no          int64
Date          object
Close        float64
Volume       int64
Open         float64
High         float64
Low          float64
dtype: object
```

```
[6]: # Convert the 'Date' column to datetime format using pandas
gold_data['Date'] = pd.to_datetime(gold_data['Date'])

# Set the 'Date' column as the index of the DataFrame 'gold_data'
gold_data.set_index('Date', inplace=True)

# Display the data types of each column in the modified DataFrame 'gold_data'
gold_data.dtypes
```

```
[6]: no          int64
Close        float64
Volume       int64
Open         float64
High         float64
Low          float64
dtype: object
```

```
[7]: # Rename columns 'Open', 'High', 'Low', and 'Close' to lowercase
gold_data.rename(columns={'Open': 'open', 'High': 'high', 'Low': 'low', 'Close':
↪ 'close'}, inplace=True)

# Display the first five rows of the modified DataFrame 'gold_data'
gold_data.head()
```

```
[7]:
```

		no	close	Volume	open	high	low
Date							
2024-01-19	0	2029.3	166078	2027.4	2041.9	2022.2	
2024-01-18	1	2021.6	167013	2009.1	2025.6	2007.7	
2024-01-17	2	2006.5	245194	2031.7	2036.1	2004.6	
2024-01-16	3	2030.2	277995	2053.4	2062.8	2027.6	
2024-01-12	4	2051.6	250946	2033.2	2067.3	2033.1	

```
[8]: # Display concise summary of the DataFrame 'gold_data', including data types
↪ and memory usage
gold_data.info()
```

```

<class 'pandas.core.frame.DataFrame'>
DatetimeIndex: 2511 entries, 2024-01-19 to 2014-01-22
Data columns (total 6 columns):
 #   Column  Non-Null Count  Dtype
---  -
 0    no      2511 non-null    int64
 1   close    2511 non-null    float64
 2  Volume    2511 non-null    int64
 3   open    2511 non-null    float64
 4   high    2511 non-null    float64
 5    low     2511 non-null    float64
dtypes: float64(4), int64(2)
memory usage: 137.3 KB

```

```

[9]: # Count the number of missing values (NaN) in each column of the DataFrame
      ↪ 'gold_data'
      gold_data.isna().sum()

```

```

[9]: no      0
      close   0
      Volume  0
      open    0
      high    0
      low     0
      dtype: int64

```

```

[10]: # Create a figure with a specific size
      plt.figure(figsize=(12, 6))

      # Create a line plot using seaborn to visualize 'Low Price' over time
      sns.lineplot(data=gold_data, x='Date', y='low', label='Low Price')

      # Set the title of the plot
      plt.title('Low and High Prices')

      # Label the x-axis
      plt.xlabel('Date')

      # Label the y-axis
      plt.ylabel('Price (USD)')

      # Display the legend on the plot
      plt.legend()

      # Show the plot
      plt.show()

```



```
[11]: # Display the column names of the DataFrame 'gold_data'
gold_data.columns
```

```
[11]: Index(['no', 'close', 'Volume', 'open', 'high', 'low'], dtype='object')
```

```
[12]: # Plot closing prices for initial visualization
plt.figure(figsize=(12, 6))

# Plot the 'close' column from gold_data
plt.plot(gold_data['close'], label='Gold Closing Price')

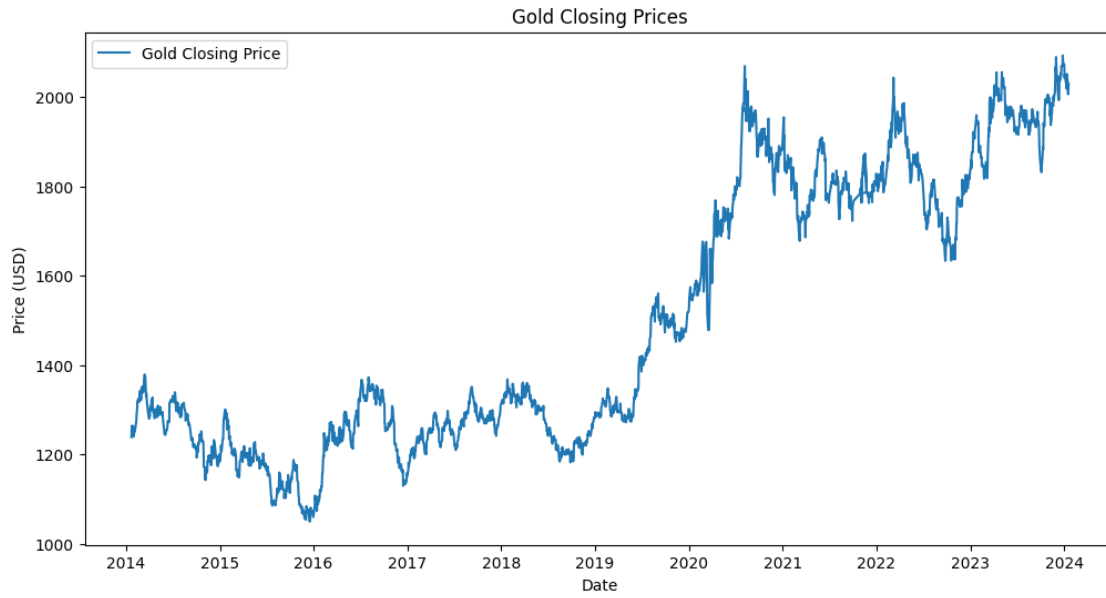
# Set the title of the plot
plt.title('Gold Closing Prices')

# Label the x-axis
plt.xlabel('Date')

# Label the y-axis
plt.ylabel('Price (USD)')

# Display the legend on the plot
plt.legend()

# Show the plot
plt.show()
```



```
[13]: # Calculate 30-day moving average for trend identification
gold_data['MA_30'] = gold_data['close'].rolling(window=30).mean()

# Plot original closing prices and moving average
plt.figure(figsize=(12, 6))

# Plot the original closing prices
plt.plot(gold_data['close'], label='Gold Closing Price')

# Plot the 30-day moving average with dashed line style
plt.plot(gold_data['MA_30'], label='30-Day Moving Average', linestyle='--')

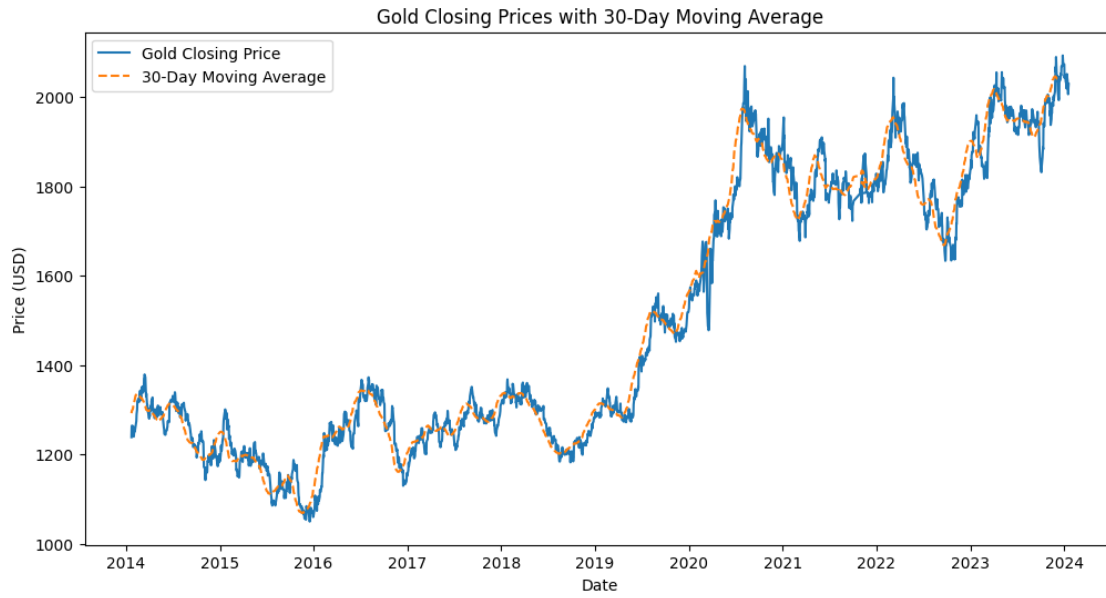
# Set the title of the plot
plt.title('Gold Closing Prices with 30-Day Moving Average')

# Label the x-axis
plt.xlabel('Date')

# Label the y-axis
plt.ylabel('Price (USD)')

# Display the legend on the plot
plt.legend()

# Show the plot
plt.show()
```



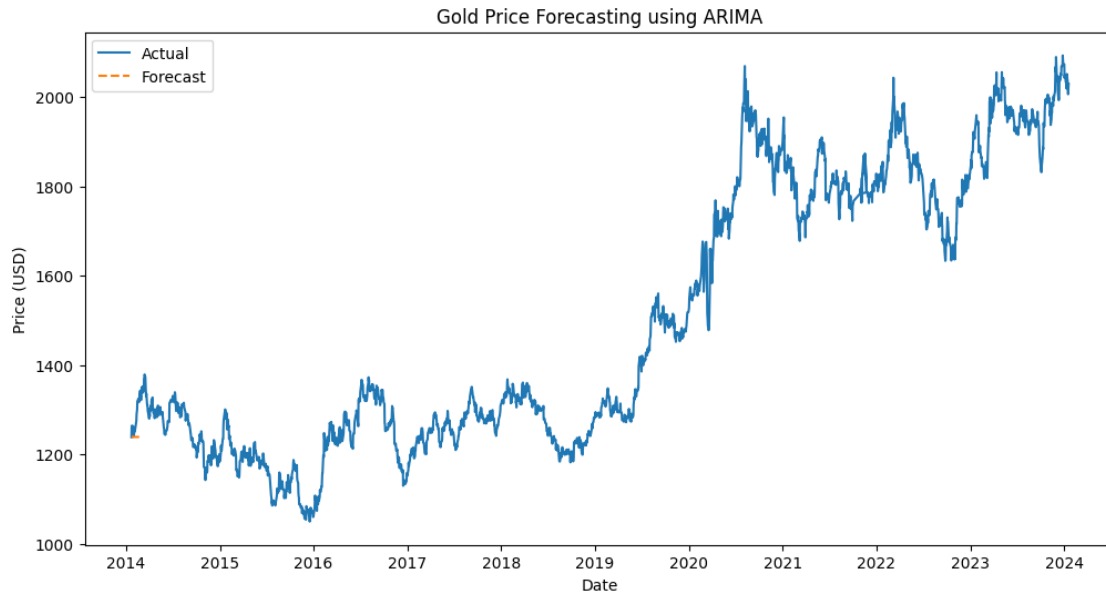
```
[14]: from statsmodels.tsa.arima.model import ARIMA
from sklearn.metrics import mean_squared_error

# Fit ARIMA model
model = ARIMA(gold_data['close'], order=(5, 1, 0)) # Example order, tune as needed
model_fit = model.fit()

# Forecast future prices
forecast_steps = 30 # Example: forecast next 30 days
forecast = model_fit.forecast(steps=forecast_steps)

# Plot original data and forecast
plt.figure(figsize=(12, 6))
plt.plot(gold_data.index, gold_data['close'], label='Actual') # Plot actual prices
plt.plot(pd.date_range(start=gold_data.index[-1], periods=forecast_steps + 1, freq='D')[1:], forecast, label='Forecast', linestyle='--') # Plot forecasted prices
plt.title('Gold Price Forecasting using ARIMA')
plt.xlabel('Date')
plt.ylabel('Price (USD)')
plt.legend()
plt.show()

# Evaluate model performance
```

Mean Squared Error (MSE): 4161.579773310184

```
[18]: # Implement a simple moving average crossover strategy
gold_data['MA_50'] = gold_data['close'].rolling(window=50).mean() # Calculate 50-day moving average
gold_data['MA_200'] = gold_data['close'].rolling(window=200).mean() # Calculate 200-day moving average

# Generate buy/sell signals based on crossover
gold_data['Signal'] = 0 # Initialize signal column
gold_data['Signal'][50:] = np.where(gold_data['MA_50'][50:] > gold_data['MA_200'][50:], 1, -1) # Generate signals

# Calculate daily returns
gold_data['Daily_Return'] = gold_data['close'].pct_change() # Calculate daily percentage change in closing prices

# Calculate strategy returns
gold_data['Strategy_Return'] = gold_data['Daily_Return'] * gold_data['Signal'].shift(1) # Calculate strategy returns

# Calculate cumulative returns
gold_data['Cumulative_Return'] = (1 + gold_data['Strategy_Return']).cumprod() # Calculate cumulative returns

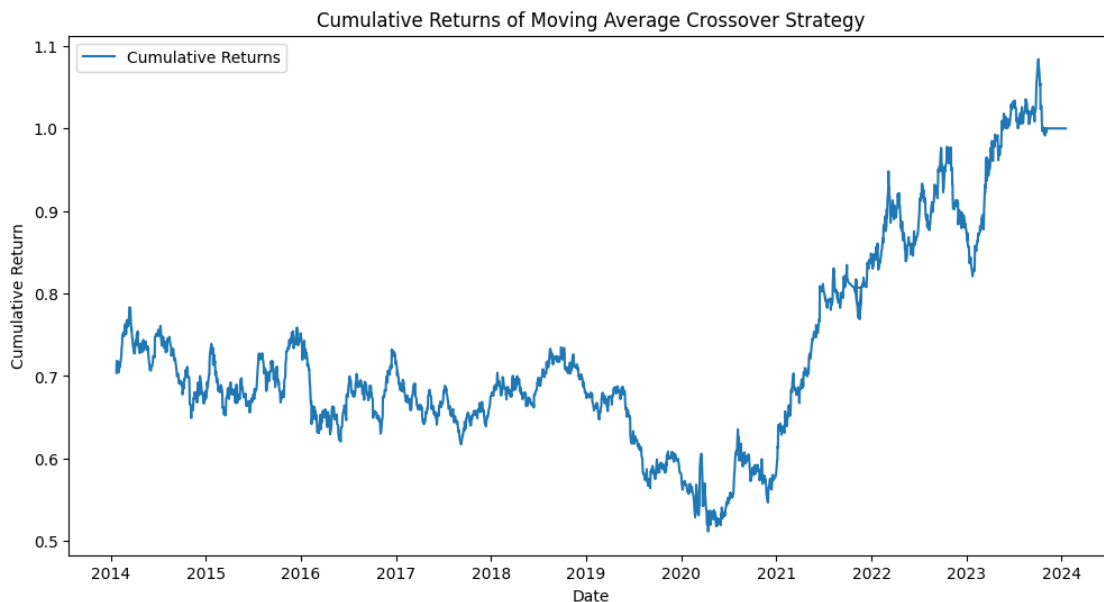
# Plot cumulative returns
plt.figure(figsize=(12, 6))
```



```
plt.plot(gold_data.index, gold_data['Cumulative_Return'], label='Cumulative_Returns') # Plot cumulative returns over time
plt.title('Cumulative Returns of Moving Average Crossover Strategy')
plt.xlabel('Date')
plt.ylabel('Cumulative Return')
plt.legend()
plt.show()
```

<ipython-input-18-7faf9f0b0115>:7: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
`gold_data['Signal'][50:] = np.where(gold_data['MA_50'][50:] >
gold_data['MA_200'][50:], 1, -1) # Generate signals`



```
[21]: gold_data.head()
```

```
[21]:
```

	no	close	Volume	open	high	low	MA_30	MA_50	MA_200	\
Date										
2024-01-19	0	2029.3	166078	2027.4	2041.9	2022.2	NaN	NaN	NaN	
2024-01-18	1	2021.6	167013	2009.1	2025.6	2007.7	NaN	NaN	NaN	
2024-01-17	2	2006.5	245194	2031.7	2036.1	2004.6	NaN	NaN	NaN	
2024-01-16	3	2030.2	277995	2053.4	2062.8	2027.6	NaN	NaN	NaN	
2024-01-12	4	2051.6	250946	2033.2	2067.3	2033.1	NaN	NaN	NaN	

Signal	Daily_Return	Strategy_Return	Cumulative_Return
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Date				
2024-01-19	0	NaN	NaN	NaN
2024-01-18	0	-0.003794	-0.0	1.0
2024-01-17	0	-0.007469	-0.0	1.0
2024-01-16	0	0.011812	0.0	1.0
2024-01-12	0	0.010541	0.0	1.0

```
[24]: # Define events and their impact dates
events = {
    'US Fed Interest Rate Decision': ['2014-01-29', '2014-03-19', '2014-06-18'],
    'Brexit Referendum': ['2016-06-23']
    # Add more events as needed
}

# Extract impact dates from dataset and analyze price movements
for event, dates in events.items():
    try:
        event_dates = pd.to_datetime(dates)
        for date in event_dates:
            # Extract price data around event date (e.g., before and after)
            price_before = gold_data.loc[date - pd.DateOffset(days=5): date -
pd.DateOffset(days=1), 'close']
            price_after = gold_data.loc[date + pd.DateOffset(days=1): date + pd.
DateOffset(days=5), 'close']

            # Compare prices and analyze impact
            if not price_before.empty and not price_after.empty:
                price_change = price_after.mean() - price_before.mean()
                print(f'Event: {event}, Date: {date.date()}, Average Price
Change: {price_change:.2f}')
            else:
                print(f'Event: {event}, Date: {date.date()}, Not enough data to
calculate price change')
        except Exception as e:
            print(f"Error processing event {event} with dates {dates}: {e}")
```

```
Event: US Fed Interest Rate Decision, Date: 2014-01-29, Not enough data to
calculate price change
Event: US Fed Interest Rate Decision, Date: 2014-03-19, Not enough data to
calculate price change
Event: US Fed Interest Rate Decision, Date: 2014-06-18, Not enough data to
calculate price change
Error processing event Brexit Referendum with dates ['2016-06-23']:
Timestamp('2016-06-18 00:00:00')
```