

Currency Exchange Rate Prediction

Group 1





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Why do we chose this topic?

We chose **Currency Exchange Rate Prediction** due to its real-world impact and the complexity of the problem. It's crucial for **businesses** and **individuals** in **international trade** or **travel**. The project provides an opportunity to apply data science skills to a multifaceted problem influenced by various factors. The potential outcome could be a tool aiding in **financial planning** and **strategy**. Beside that we also want to know the **Trend** and the **Behavior** on the rate between the 2 currency.



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Data Collection



Three step to get the Dataset



Yahoo Finance

Go to Yahoo Finance and Search "USD/KHR(KHR=X)" or click this link.



Navigation

After landing on Yahoo Finance, navigate to Historical Data.



Download

Select the criteria you want then click on the download button.



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Data Preprocessing

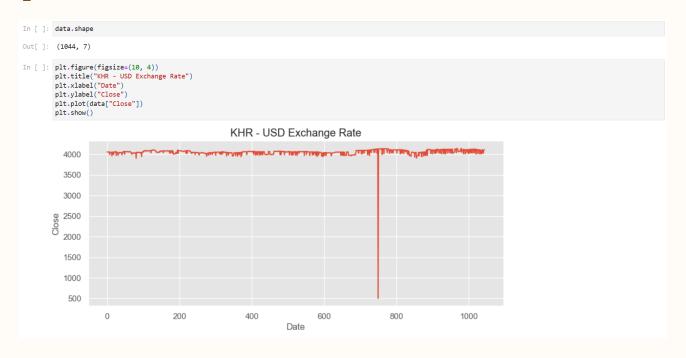




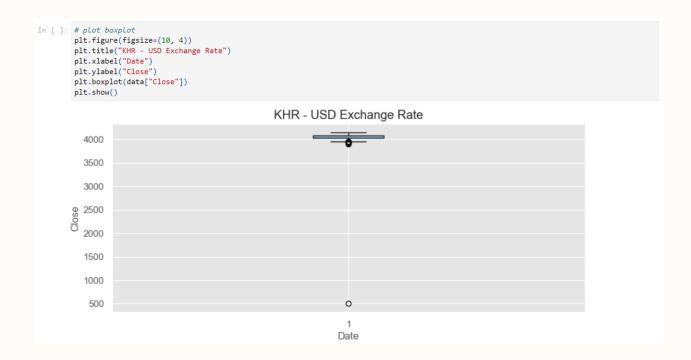
Import Data and Necessary Library

```
In [ ]: import pandas as pd
        import numpy as np
        import matplotlib.pyplot as plt
        import seaborn as sns
        sns.set() # setting seaborn default for plots
        plt.style.use('ggplot')
        data = pd.read csv("https://raw.githubusercontent.com/PLSeng/MyPage/main/web/assets/KHR%3DX.csv")
        data.head()
                                  High
                                               Low Close Adj Close Volume
                Date
                            Open
        0 2019-12-09 3992.131348 4060.0 3992.053223
                                                               4055.0
        1 2019-12-10 3995.426758 4055.0 3995.426758 4055.0
                                                               4055.0
        2 2019-12-11 3992499023 4055.0 3989313232 4053.0
                                                               4053.0
        3 2019-12-12 3968.466309 4055.0 3968.466309 4055.0
                                                               4055.0
        4 2019-12-13 3962.562256 4050.0 3962.562256 4051.0
                                                               4051.0
```

Import Data and Necessary Library



Plot Boxplot to see outliers



Drop Empty Column

```
In [ ]: # preprocessing
        # drop Volume column
        data = data.drop(['Volume'], axis=1)
        data.head()
Out[ ]:
                            Open High
                                                      Close Adj Close
                 Date
        0 2019-12-09 3992.131348 4060.0 3992.053223
                                                               4055.0
        1 2019-12-10 3995.426758 4055.0 3995.426758 4055.0
                                                                4055.0
        2 2019-12-11 3992.499023 4055.0 3989.313232 4053.0
                                                               4053.0
         3 2019-12-12 3968.466309 4055.0 3968.466309 4055.0
                                                                4055.0
         4 2019-12-13 3962.562256 4050.0 3962.562256 4051.0
                                                                4051.0
```

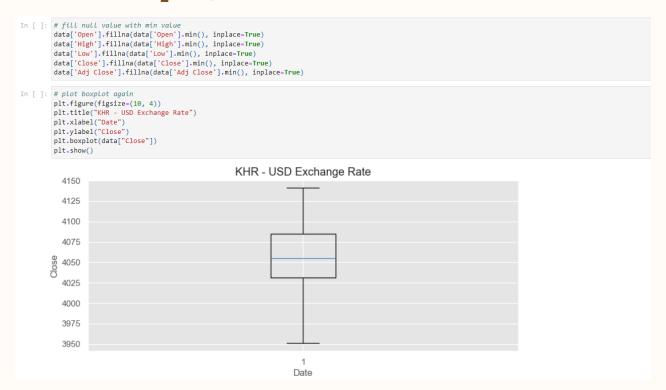
Drop Empty Column

```
# preprocessing
        # drop Volume column
        data = data.drop(['Volume'], axis=1)
        data.head()
Out[ ]:
                            Open
                                   High
                                                      Close Adj Close
                 Date
        0 2019-12-09 3992.131348 4060.0 3992.053223 4055.0
                                                               4055.0
        1 2019-12-10 3995.426758 4055.0 3995.426758 4055.0
                                                               4055.0
        2 2019-12-11 3992.499023 4055.0 3989.313232 4053.0
                                                               4053.0
        3 2019-12-12 3968.466309 4055.0 3968.466309 4055.0
                                                               4055.0
        4 2019-12-13 3962.562256 4050.0 3962.562256 4051.0
                                                               4051.0
```

Drop Outliers

```
In [ ]: # find outlier of each column except date and remove the outlier
        Q1 = data['Open'].quantile(0.25)
        Q3 = data['Open'].quantile(0.75)
        IOR = 03 - 01
        # set outlier to null
        data['Open'] = np.where(data['Open'] > (Q3 + 1.5 * IQR), None, data['Open'])
        data['Open'] = np.where(data['Open'] < (Q1 - 1.5 * IQR), None, data['Open'])
        Q1 = data['High'].quantile(0.25)
        Q3 = data['High'].quantile(0.75)
        IOR = 03 - 01
        # set outlier to null
        data['High'] = np.where(data['High'] > (Q3 + 1.5 * IQR), None, data['High'])
        data['High'] = np.where(data['High'] < (01 - 1.5 * IOR), None, data['High'])
        Q1 = data['Low'].quantile(0.25)
        Q3 = data['Low'].quantile(0.75)
        IQR = Q3 - Q1
        # set outlier to null
        data['Low'] = np.where(data['Low'] > (Q3 + 1.5 * IQR), None, data['Low'])
        data['Low'] = np.where(data['Low'] < (Q1 - 1.5 * IQR), None, data['Low'])
        Q1 = data['Close'].quantile(0.25)
        03 = data['Close'].quantile(0.75)
        IQR = Q3 - Q1
        # set outlier to null
        data['Close'] = np.where(data['Close'] > (03 + 1.5 * IOR), None, data['Close'])
        data['Close'] = np.where(data['Close'] < (Q1 - 1.5 * IQR), None, data['Close'])</pre>
        Q1 = data['Adj Close'].quantile(0.25)
        Q3 = data['Adj Close'].quantile(0.75)
        IQR = Q3 - Q1
        # set outlier to null
        data['Adj Close'] = np.where(data['Adj Close'] > (Q3 + 1.5 * IQR), None, data['Adj Close'])
        data['Adj Close'] = np.where(data['Adj Close'] < (Q1 - 1.5 * IQR), None, data['Adj Close'])
In [ ]: data.isnull().sum()
Out[]: Date
                     13
        High
                     16
        Low
        Close
                     11
        Adj Close 11
        dtype: int64
```

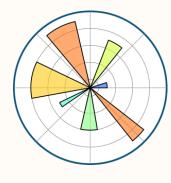
Fill Empty cell with min value





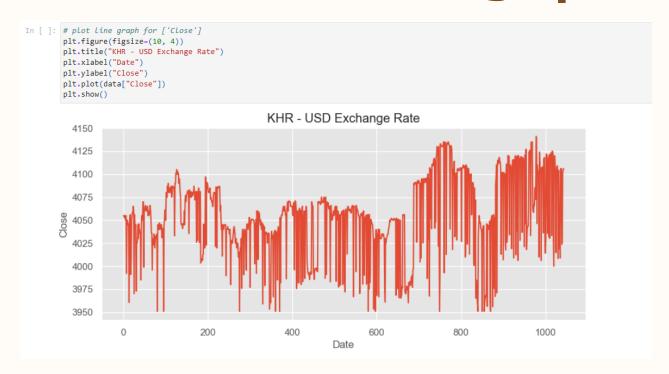


Data Visualisation





Plot the time series graph



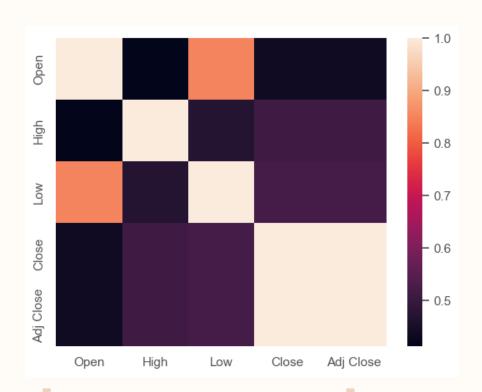
Plot a heatmap to see the correlation

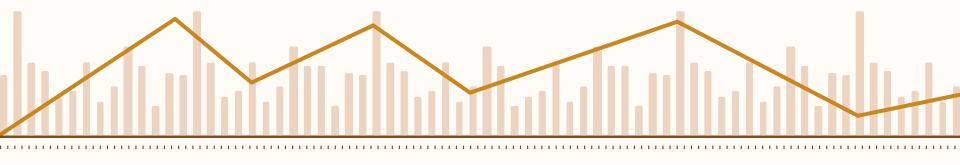
```
In []: # Select all columns except the first one
    data_for_corr = data.iloc[:, 1:]

# Calculate the correlation matrix
    corr_matrix = data_for_corr.corr()
    print(corr_matrix)

# Create a heatmap
    sns.heatmap(corr_matrix)
    plt.show()
```

Plot a heatmap to see the correlation







Data Analysis



Scikit-Learn



train_test_split

train_test_split in scikit-learn splits data into random train and test subsets. It takes arrays as inputs and you can specify the size of the test and train datasets. It also allows for reproducible output and stratified splitting.



DecisionTreeRegressor

DecisionTreeRegressor in scikit-learn is a decision tree regressor. It splits data into train and test subsets based on various parameters like criterion, splitter, max_depth, etc. It's used for creating models that predict the value of a target variable.

Regression

```
In []: x = data[["Open", "High", "Low"]]
y = data["Close"]
x = x.to_numpy()
y = y.to_numpy()
y = y.reshape(-1, 1)
```

```
In []: # Predict the rate for the next 7 days
    from sklearn.model_selection import train_test_split

# Remove rows with missing values
    data.dropna(inplace=True)

# Split the data into training and testing sets
    xtrain, xtest, ytrain, ytest = train_test_split(x, y, test_size=0.2, random_state=4

from sklearn.tree import DecisionTreeRegressor
    model = DecisionTreeRegressor()
# Train the model
    model.fit(xtrain, ytrain)
# Make predictions
ypred = model.predict(xtest)
```

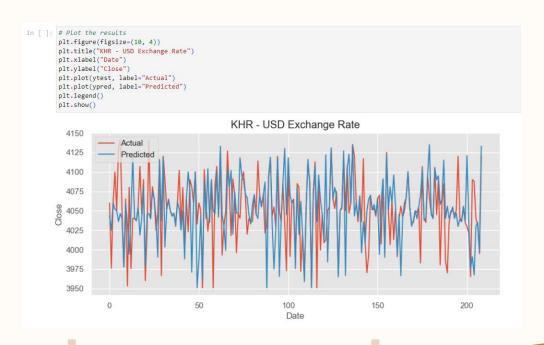
Prediction



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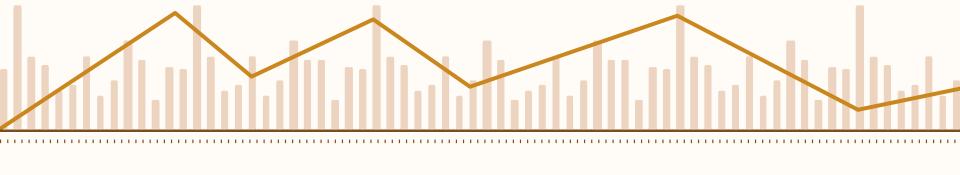
Accuracy Rate

Trendline between Actual and Predicted Value



Get the accuracy rate

```
In [ ]: # Calculate the mean absolute error
        from sklearn.metrics import mean_absolute_error
        mae = mean_absolute_error(ytest, ypred)
        print("MAE:", mae)
        # Calculate the mean squared error
        from sklearn.metrics import mean squared error
        mse = mean_squared_error(ytest, ypred)
        print("MSE:", mse)
        # Calculate the root mean squared error
        from math import sqrt
        rmse = sqrt(mse)
        print("RMSE:", rmse)
        # Calculate the mean absolute percentage error
        errors = abs(ypred - ytest)
        mape = 100 * (errors / ytest)
       MAE: 26.479466464114825
       MSE: 1891.0576261088233
      RMSE: 43.48629239322229
```



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Conclusion

Thanks!

Do you have any questions?

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