Week 2 Attempt Summary

Introduction

In this week we were advised to create a load and process data function to a dataset with multiple features, with the following functional requirements

- Should be able to specify and start and end date for the whole dataset
- Deal with Nan values in the dataset
- Split train/test data according to a specified ratio (Date)
- Should be able to download the data to the local machine and store them to future use
- Scaling in the data structure

Load_and_process_data Function

```
if save data and local path:
            data.to csv(local path)
    if nan method == 'drop':
        data = data.dropna()
    elif nan method == 'fill':
        data = data.ffill().bfill()
    if data.empty:
        raise ValueError(f"No data found for {ticker} between
{start date} and {end date}. Please check the date range and ticker
symbol.")
    if split method == 'ratio':
        train data, test data = train test split(data,
train size=train size, test size=test size, random state=random state,
shuffle=True)
    elif split method == 'date' and split date:
        train data = data[:split date]
        test data = data[split date:]
    scalers = {}
    if scale_features:
        for column in train data.columns:
            scaler = MinMaxScaler(feature_range=(0, 1))
            train data[column] =
scaler.fit transform(train data[[column]])
            if not test_data.empty:
                test data[column] =
scaler.transform(test data[[column]])
            if save scaler:
                scalers[column] = scaler
    return train data, test data, scalers
```

This function is in a different python file (load_process_data.py). I am importing the function with a import statement at the top

```
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
import yfinance as yf
import os
import tensorflow as tf
import load process data as data function
from sklearn.preprocessing import MinMaxScaler
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense, Dropout, LSTM
```

And this to instantiate it in the main python script

This week 2 code base allows the user to enter the start date and the end date manually through the terminal before training the model

```
start_date = input("Enter Start date (YYYY-MM-DD): ")
  end_date = input("Enter End date (YYYY-MM-DD): ")
```

And It handles the Nan values with predefined Nan value handlers that comes with the pandas library

Dropna:- Removes data Frames that contains Nan values

Fillna:- Fills missing data Frames with a specified value

And with with some exception handling we have this code

```
# Handle NaN values
    if nan_method == 'drop':
        data = data.dropna()
    elif nan_method == 'fill':
        data = data.ffill().bfill()

# Ensure data is not empty
    if data.empty:
        raise ValueError(f"No data found for {ticker} between
{start_date} and {end_date}. Please check the date range and ticker
symbol.")
```

it raises a Error massage when ever the data array might be empty (for debugging pruposes)

And this code facilitates the functionality to split the data into 2, Training and testing data

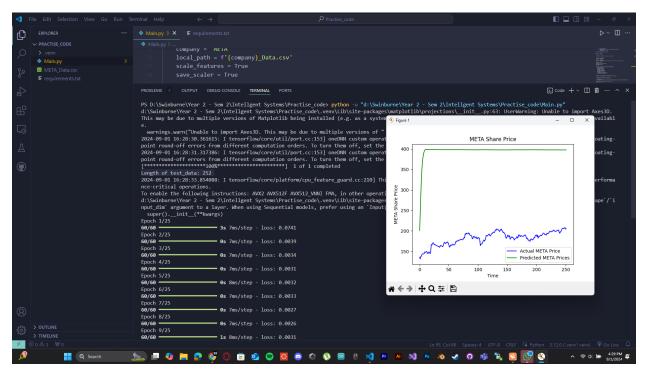
```
# Split the data
    if split_method == 'ratio':
        train_data, test_data = train_test_split(data,
train_size=train_size, test_size=test_size, random_state=random_state,
shuffle=True)
    elif split_method == 'date' and split_date:
        train_data = data[:split_date]
    test_data = data[split_date:]
```

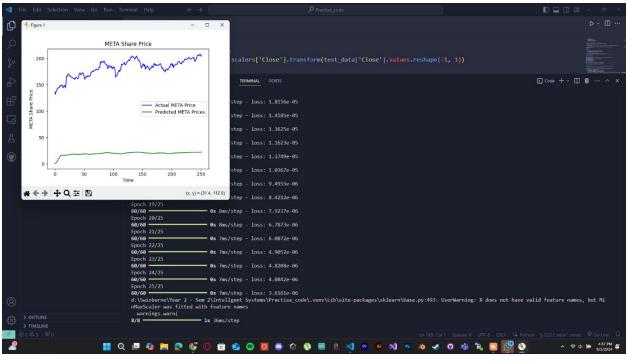
Yet the train_test_split function is still under development, for this week the date method will be used.

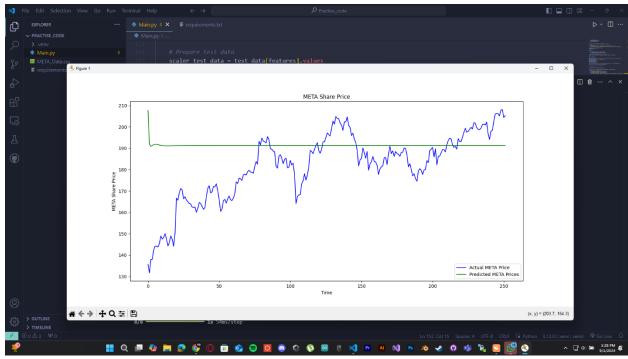
And then the last requirement was for adding scalers to the code base

```
# Scale features if required
    scalers = {}
    if scale_features:
        for column in train_data.columns:
            scaler = MinMaxScaler(feature_range=(0, 1))
            train_data[column] =
scaler.fit_transform(train_data[[column]])
            if not test_data.empty:
                test_data[column] =
scaler.transform(test_data[[column]])
            if save_scaler:
                scalers[column] = scaler
```

I had problems with using the "Adam" optimizer when compiling the model





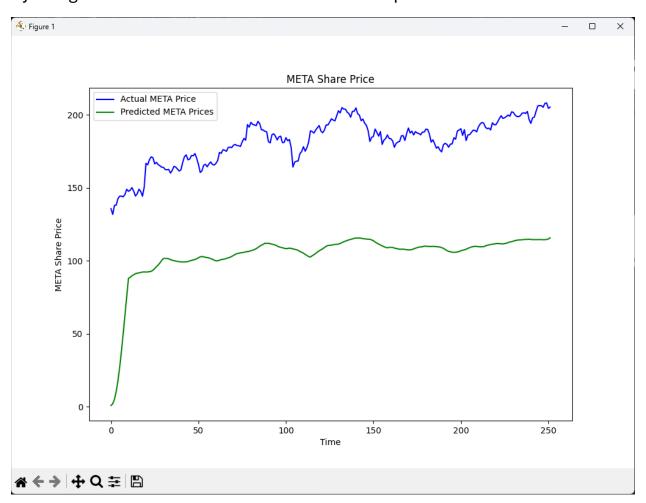


Then I changed the 'Adam' optimizer and used tensor flow keras adam optimizer instead, with 100 epochs

```
model.compile(optimizer =

tf.keras.optimizers.Adam(learning_rate=0.001),
loss='mean_squared_error')
    model.fit(x_train, y_train, epochs=50, batch_size=32)
```

By using that I was able to take a reasonable output



But this is very far from being perfect but this is still a work in progress