11. LSTMCLosePricePredictionWithSentiment

August 30, 2023

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[]: # dataset @ https://finance.yahoo.com/quote/MSFT/history/
              # If you want the exact same dataset as the YouTube video,
               # use this link: https://drive.google.com/file/d/
                 →1WLm1AEYgU28Nk4lY4zNkGPSctdImbhJI/view?usp=sharing
[]: import pandas as pd
              import datetime
               # Stocks :- AAPL, MSFT, AMZN, NVDA, TSLA, GOOGL, UNH
               # Sector Indices :- SPINF (^SP500-45)
              method= 'TextBlob'
              ticker = "AAPL"
              df = pd.read_csv(f"SentimentAnalysis/{method}/
                 df
[]: df = df[['Date', 'Close', 'polarity']]
[]: | # # Normalize the 'Close' and 'polarity' columns using min-max scaling
               # min_close = df['Close'].min()
              # max_close = df['Close'].max()
               # min_polarity = df['polarity'].min()
               # max_polarity = df['polarity'].max()
               \# df['Close'] = (df['Close'] - min_close) / (max_close - min_close)
               \# df['polarity'] = (df['polarity'] - min_polarity) / (max_polarity - delta for a finite for a 
                  ⇔min_polarity)
[]: df["Date"] = pd.to_datetime(df['Date'], format="%Y-%m-%d")
              df.index = df.pop('Date')
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[]: import matplotlib.pyplot as plt
     import matplotlib.dates as mdates
     plt.plot(df.index, df['Close'])
     plt.xticks(rotation=90)
[]: import numpy as np
     def df to windowed df(dataframe, first_date str, last_date_str, n=3):
         first_date = pd.to_datetime(first_date_str, format="%Y-%m-%d")
         last_date = pd.to_datetime(last_date_str, format="%Y-%m-%d")
         target_date = first_date
         dates = []
         X, Y, Polarity = [], [], []
         last_time = False
         while True:
             df_subset = dataframe.loc[:target_date].tail(n+1)
             if len(df_subset) != n+1:
                 print(f'Error: Window of size {n} is too large for date⊔
      →{target_date}')
                return
            values = df_subset['Close'].to_numpy()
            x, y = values[:-1], values[-1]
             target_polarity = df_subset['polarity'].iloc[-1] # Fetch polarity for_
      →target date
             dates.append(target_date)
            X.append(x)
            Y.append(y)
            Polarity.append(target_polarity) # Store polarity
            next_week = dataframe.loc[target_date:target_date+datetime.
      →timedelta(days=7)]
            next_datetime_str = str(next_week.head(2).tail(1).index.values[0])
            next_date_str = next_datetime_str.split('T')[0]
             year_month_day = next_date_str.split('-')
            year, month, day = year_month_day
            next_date = datetime.datetime(day=int(day), month=int(month),__
      →year=int(year))
             if last_time:
                 break
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target_date = next_date
             if target_date == last_date:
                 last_time = True
         ret_df = pd.DataFrame({})
         ret_df['Target Date'] = dates
         X = np.array(X)
         for i in range(0, n):
             ret_df[f'Target-{n-i}'] = X[:, i]
         ret_df['Target'] = Y
         ret_df['Target Polarity'] = Polarity # Include polarity in returned □
      \hookrightarrow dataframe
         return ret_df
     windowed_df = df_to_windowed_df(df, '2021-03-25', '2022-03-25', n=3)
     windowed_df
[]: def windowed_df_to_date_X_y(windowed_dataframe):
         df_as_np = windowed_dataframe.to_numpy()
         dates = df_as_np[:, 0]
         # Create a matrix that contains Close values and the target polarity for
      ⇔each record
         middle_matrix = np.column_stack((df_as_np[:, 1:-2], df_as_np[:, -1][:, np.
      →newaxis]))
         X = middle_matrix.reshape((len(dates), middle_matrix.shape[1], 1))
         Y = df_as_np[:, -2] # Get the 'Target' values
         return dates, X.astype(np.float32), Y.astype(np.float32)
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[]: q_80 = int(len(dates) * .8)
q_90 = int(len(dates) * .9)
dates_train, X_train, y_train = dates[:q_80], X[:q_80], y[:q_80]
dates_val, X_val, y_val = dates[q_80:q_90], X[q_80:q_90], y[q_80:q_90]
dates_test, X_test, y_test = dates[q_90:], X[q_90:], y[q_90:]

plt.plot(dates_train, y_train)
plt.plot(dates_val, y_val)
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dates, X, y = windowed_df_to_date_X_y(windowed_df)

dates.shape, X.shape, y.shape

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plt.xticks(rotation=90)
     plt.legend(['Train', 'Validation', 'Test'])
[]: from tensorflow.keras.models import Sequential
     from tensorflow.keras.optimizers import Adam
     from tensorflow.keras import layers
     # Adjusting the input shape
     input_shape = (X.shape[1], 1) # This will now be (4, 1) - 3 for 'Close' values_
      →and 1 for the target polarity
     model = Sequential([
        layers.Input(shape=input_shape),
        layers.LSTM(64),
        layers.Dense(32, activation='relu'),
        layers.Dense(32, activation='relu'),
        layers.Dense(1)
     ])
     model.compile(loss='mse',
                   optimizer=Adam(learning_rate=0.001),
                   metrics=['mean_absolute_error'])
     model.fit(X_train, y_train, validation_data=(X_val, y_val), epochs=100)
[]: loss, mae = model.evaluate(X_test, y_test, verbose=0)
     print(f"Mean Squared Error on the test set: {loss}")
[]: train_predictions = model.predict(X_train).flatten()
     plt.plot(dates_train, train_predictions)
     plt.plot(dates_train, y_train)
     plt.legend(['Training Predictions', 'Training Observations'])
[]: val_predictions = model.predict(X_val).flatten()
     plt.plot(dates_val, val_predictions)
     plt.plot(dates_val, y_val)
     plt.legend(['Validation Predictions', 'Validation Observations'])
     plt.xticks(rotation=90)
[]: test_predictions = model.predict(X_test).flatten()
     plt.plot(dates_test, test_predictions)
     plt.plot(dates_test, y_test)
     plt.legend(['Testing Predictions', 'Testing Observations'])
     plt.xticks(rotation=90)
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plt.plot(dates_test, y_test)