Time series forecasting using Prophet library

#NVIDIA Stock predictions

#Importing the important libraries

```
import pandas as pd

import numpy as np

import plotly.express as px

!pip install prophet

import matplotlib.pyplot as plt

!pip install scikit-learn
```

#Loading and prepapring the dataset

```
df = pd.read_csv(r"C:\Users\vaibh\Downloads\Time Series Analysis - NVIDIA Stocks\NVDA Historical Data.csv")

df.head()

df.info()

df.describe()

df['Date'] = pd.to_datetime(df['Date'])


dfnew = df[["Date", "Price"]]
```

#Doing the train-test split

```
train_size = int(len(dfnew) * 0.8)

train = dfnew[:train_size]  # first 80%
```

```
test = dfnew[train_size:]   # last 20%


print("Train shape:", train.shape)

print("Test shape:", test.shape)


df.columns = df.columns.str.strip().str.lower()

# Now 'Date' becomes 'date', 'Price' becomes 'price'


#Plotting the graph initially

fig = px.line(df, x='date', y='price', title='NVIDIA Stock Price Over Time')

import plotly.io as pio

pio.renderers.default = 'browser'

fig.show()


train_df = train[["Date", "Price"]].copy()

train_df.rename(columns={"Date": "ds", "Price": "y"}, inplace=True)



#Calling prophet to make predictions

from prophet import Prophet

model = Prophet(daily_seasonality=True)

model.fit(train_df)


#Making predictions for next 51 days

future = model.make_future_dataframe(periods=73,freq="D")

forecast = model.predict(future)
```

```
#Plotting actual vs predicted graph
fig = px.line(df, x='date', y='price', title='Actual vs Predicted')
fig.add_scatter(x=forecast['ds'], y=forecast['yhat'], mode='lines', name='Predicted')

fig.show()

#Again calling prophet
from prophet import Prophet

model = Prophet(
    seasonality_mode="multiplicative",
    daily_seasonality=True,
    weekly_seasonality=True,
    changepoint_prior_scale=0.1
)
model.fit(train_df)

#Making predictions for next 51 days
future = model.make_future_dataframe(periods=73,freq="D")
forecast = model.predict(future)
#Checking the general forecast with upper and lower limits
model.plot(forecast)
#Checking the components
model.plot_components(forecast)
```

```
#Doing some Time Series Accuracy tests

from sklearn.metrics import mean_absolute_error


# Step 1: Prepare actual

actual = df[["date", "price"]].copy()

actual.rename(columns={"date": "ds", "price": "y"}, inplace=True)

actual.set_index('ds', inplace=True)


# Step 2: Prepare forecast

forecast.set_index('ds', inplace=True)


# Step 3: Keep only overlapping dates

common_dates = actual.index.intersection(forecast.index)


# Step 4: Align both series

actual_y = actual.loc[common_dates]['y']

predicted_y = forecast.loc[common_dates]['yhat']


# Step 5: Calculate MAE

mae = mean_absolute_error(actual_y, predicted_y)

print(f'MAE: {mae:.2f}')


avg_price = df['price'].mean()

print(f"Average Price: {avg_price:.2f}")
```