#### **Stock Market Indexes Predictions**

# **Importing Libraries**

```
In [1]: import pandas as pd
        import numpy as np
        from datetime import datetime
        from sklearn.metrics import mean absolute error
        import matplotlib.pyplot as plt
        get ipython().run line magic('matplotlib', 'inline')
        import warnings
        # !pip install pystan~=2.14
        # !pip install fbprophet
        # warnings.filterwarnings('ignore')
        # import fbprophet
        # # print version number
        # print('Prophet %s' % fbprophet. version )
        # from matplotlib import pyplot
        # from fbprophet import Prophet
        from prophet import Prophet
        import holidays
```

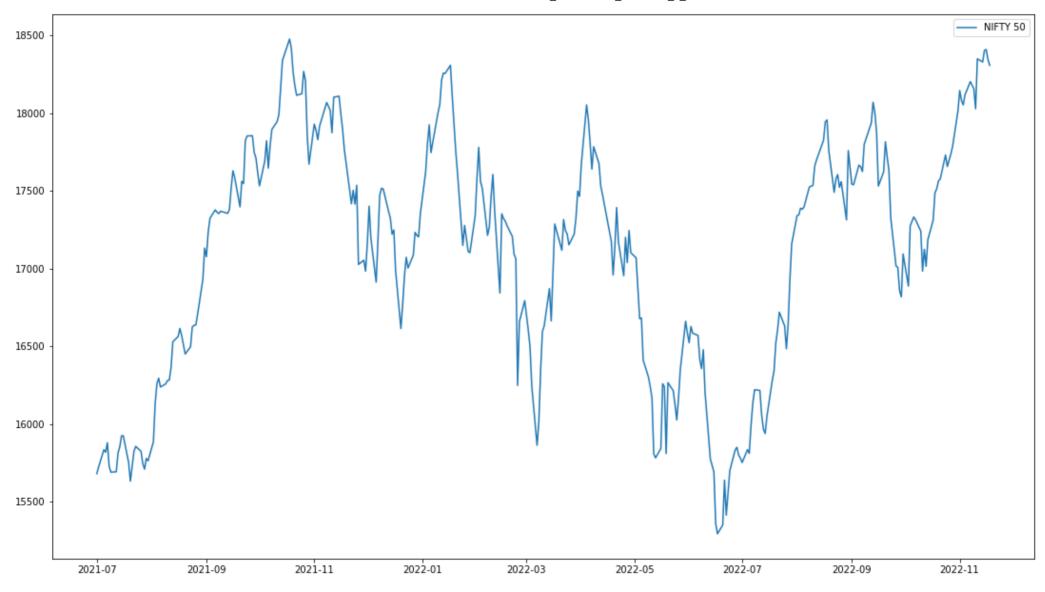
# NIfty 50

```
In [388... #Reading Nifty File
    df_Nifty = pd.read_csv('/content/NIFTY50.csv')
    df_Nifty["date_clean"] = df_Nifty["Date"].apply(lambda x: datetime.strptime(str(x), '%d-%m-%Y').date())
    df_Nifty['year']=df_Nifty['date_clean'].apply(lambda x: x.year)
    df_Nifty['Month']=df_Nifty['date_clean'].apply(lambda x: x.month)
    df_Nifty['day']=df_Nifty['date_clean'].apply(lambda x: x.day)
    #df_Nifty=df_Nifty[(df_Nifty['year']>=2020)]
    df_Nifty=df_Nifty[((df_Nifty['year']==2021)&(df_Nifty['Month']>6))|((df_Nifty['year']==2022)&(df_Nifty['Month']>=1))]
    df_Nifty.head()
```

Out[388]:		Date	Open	High	Low	Close	Adj Close	Volume	date_clean	year	Month	day
	615	01-07-2021	15755.05	15755.55	15667.05	15680.00	15680.00	224900.0	2021-07-01	2021	7	1
	616	02-07-2021	15705.85	15738.35	15635.95	15722.20	15722.20	254800.0	2021-07-02	2021	7	2
	617	05-07-2021	15793.40	15845.95	15762.05	15834.35	15834.35	207000.0	2021-07-05	2021	7	5
	618	06-07-2021	15813.75	15914.20	15801.00	15818.25	15818.25	391400.0	2021-07-06	2021	7	6
	619	07-07-2021	15819.60	15893.55	15779.70	15879.65	15879.65	329300.0	2021-07-07	2021	7	7

### EDA

```
from matplotlib import pyplot
fig = pyplot.gcf()
fig.set_size_inches(18.5, 10.5)
pyplot.plot(df_Nifty['date_clean'],df_Nifty["Close"], label='NIFTY 50')
pyplot.legend()
pyplot.show()
```



```
In [390...

def preprocessing(df):
    df=df.sort_values(by=["date_clean"],ascending=True)
    df=df.rename(columns={'date_clean':'ds','Close':'y'})
    df=df.reset_index(drop=True)
    df=df[['ds','y']]
    return df

DF_Nifty=preprocessing(df_Nifty)
```

```
def mean absolute percentage error(y true, y pred):
   v true, v pred = np.arrav(v true), np.arrav(v pred)
    return np.mean(np.abs((v true - v pred) / v true)) * 100
def ProhetModel(DF.CountryName):
 #train test split
  df2 test=DF[-5:]
  df2 train=DF[:-5]
  test val = df2 test.copv()
  train val = df2 train.copv()
  train val = train val.reset index(drop=True)
  test val = test val.reset index(drop=True)
  # Setup model with holidays
  final model = Prophet(#holidays=holiday,
                      changepoint prior scale= 0.009.
                      #holidays prior scale = 0.3,
                      n changepoints = 2.
                      seasonality mode = 'multiplicative'.
                      weekly seasonality=10.
                      daily seasonality = False.
                      #yearly seasonality = 5.
                      interval width=0.95)
  final model.add country holidays(country name=CountryName)
  #final model.add seasonality(name='auarterly'. period=30.5)
  final model.add seasonality('quarterly', period=91.25, fourier order=4, mode='additive')
  #final model.add seasonality(name='monthly', period=30.5, fourier order=5,mode='additive')
  #final model.add country holidays(country name=CountryName)
  #final model.add seasonality(name='auarterly'. period=30.5.fourier order=5)
 final model.fit(train val)
  forecast = final model.predict(test val.drop("v", axis=1))
  mae = mean absolute error(test val["y"], list(forecast["yhat"]))
  print('MAE: %.3f' % mae)
  mape = mean absolute percentage error(test val["y"], list(forecast["yhat"]))
  print('MAPE: %.3f' % mape)
  se = np.square(test val["y"] -list(forecast["yhat"]))
 mse = np.mean(se)
 rmse = np.sqrt(mse)
 return mae, mape, forecast, test val, rmse, train val
mae nifty, mape nifty, forecast nifty,test val nifty,rmse nifty,train val nifty = ProhetModel(DF Nifty,'India')
print('MAE NIFTY = ',mae nifty)
print('MAPE NIFTY = ', mape nifty)
print('RMSE NIFTY = ',rmse nifty)
print(forecast_nifty[['ds','yhat','yhat_lower','yhat_upper']])
print(test val nifty)
from matplotlib import pyplot
```

```
pyplot.plot(test val nifty["v"], label='Actual nifty')
pvplot.plot(list(forecast niftv["vhat"]), label='Predicted niftv')
pvplot.legend()
pvplot.show()
INFO:prophet:Disabling yearly seasonality. Run prophet with yearly seasonality=True to override this.
/usr/local/lib/python3.7/dist-packages/prophet/hdays.py:209: Warning: We only support Diwali and Holi holidays from 2010 to 2030
 warnings.warn(warning msg, Warning)
DEBUG:cmdstanpv:input tempfile: /tmp/tmp9puoo3fo/6phpavsx.ison
DEBUG:cmdstanpy:input tempfile: /tmp/tmp9puoo3fo/5x kpnm .json
DEBUG:cmdstanpy:idx 0
DEBUG:cmdstanpv:running CmdStan, num threads: None
DEBUG:cmdstanpy:CmdStan args: ['/usr/local/lib/python3.7/dist-packages/prophet/stan model/prophet model.bin', 'random', 'seed=2338', 'data', 'file=/tmp/tmp9puoo3fo/6php
avsx.json', 'init=/tmp/tmp9puoo3fo/5x kpnm .json', 'output', 'file=/tmp/tmp9puoo3fo/prophet modelkxfxbb4f/prophet model-20221120134107.csv', 'method=optimize', 'algorit
hm=lbfgs', 'iter=10000']
13:41:07 - cmdstanpy - INFO - Chain [1] start processing
INFO:cmdstanpv:Chain [1] start processing
13:41:07 - cmdstanpy - INFO - Chain [1] done processing
INFO:cmdstanpy:Chain [1] done processing
/usr/local/lib/python3.7/dist-packages/prophet/hdays.py:209: Warning: We only support Diwali and Holi holidays from 2010 to 2030
 warnings.warn(warning msg, Warning)
/usr/local/lib/python3.7/dist-packages/prophet/hdays.py:209: Warning: We only support Diwali and Holi holidays from 2010 to 2030
 warnings.warn(warning msg, Warning)
MAE: 35.581
MAPE: 0.194
MAE NIFTY = 35.58099846776749
MAPE NIFTY = 0.19397875309095514
RMSE NIFTY = 39.05617325875561
         ds
                     yhat yhat lower
                                          yhat upper
0 2022-11-14 18277.285732 17139.133699 19442.066715
1 2022-11-15 18377.798420 17212.634694 19471.542154
2 2022-11-16 18418.633537 17279.886436 19606.519411
3 2022-11-17 18392.694737 17209.840441 19631.196204
4 2022-11-18 18350.310871 17237.591832 19468.136990
           ds
0 2022-11-14 18329.15
1 2022-11-15 18403.40
2 2022-11-16 18409.65
3 2022-11-17 18343.90
4 2022-11-18 18307.65
```

```
18420
18400
18380
18360
18340
18320
18300
                                                  Actual nifty
18280
                                                  Predicted nifty
                    10
                           1.5
                                  2.0
                                                     3.5
              0.5
                                        2.5
```

```
def preprocessing(df):
    df=df.sort_values(by=["date_clean"],ascending=True)
    df=df.rename(columns={'date_clean':'ds','Close':'y'})
    df=df.reset_index(drop=True)
    df=df[['ds','y']]
    return df

DF_Nifty=preprocessing(df_Nifty)

def mean_absolute_percentage_error(y_true, y_pred):
    y_true, y_pred = np.array(y_true), np.array(y_pred)
    return np.mean(np.abs((y_true - y_pred) / y_true)) * 100
```

## Predicting(NIFTY 50) for 21st Nov to 25th Nov

```
import dateutil
future = []
for i in range(3,8,1):
    future.append(DF_Nifty["ds"][DF_Nifty.shape[0]-1] + dateutil.relativedelta(days=i))

test_nifty = pd.DataFrame({"ds":future})
test_nifty = test_nifty.reset_index(drop=True)
print(test_nifty)
```

```
ds
         0 2022-11-21
         1 2022-11-22
         2 2022-11-23
         3 2022-11-24
         4 2022-11-25
In [394... def ProhetModel test nifty(test,train,CountryName):
           #train test split
           test val = test.copy()
           train val = train.copy()
           train val = train val.reset index(drop=True)
           test val = test val.reset index(drop=True)
           # Setup model with holidays
           final model = Prophet(#holidays=holiday,
                                changepoint prior scale= 0.009,
                                #holidays prior scale = 0.3.
                                n changepoints = 2.
                                seasonality mode = 'multiplicative',
                                weekly seasonality=10,
                                daily seasonality = False,
                                #vearly seasonality = 5.
                               interval width=0.95)
           final model.add country holidays(country name=CountryName)
           #final model.add seasonality(name='quarterly', period=30.5)
           final model.add seasonality('quarterly', period=91.25, fourier order=4, mode='additive')
           #final model.add seasonality(name='monthly', period=30.5, fourier order=5,mode='additive')
           #final model.add country holidays(country name=CountryName)
           #final model.add seasonality(name='quarterly', period=30.5, fourier order=5)
           final model.fit(train val)
           forecast = final model.predict(test val)
           a=final model.train holiday names.to list()
           return forecast , test_val,train_val
         train nifty=DF Nifty.copy()
         final forecast nifty, final test val nifty, final train val nifty = ProhetModel test nifty (test nifty, train nifty, 'India')
```

```
INFO:prophet:Disabling yearly seasonality. Run prophet with yearly seasonality=True to override this.
/usr/local/lib/python3.7/dist-packages/prophet/hdays.py:209: Warning: We only support Diwali and Holi holidays from 2010 to 2030
  warnings.warn(warning msg. Warning)
DEBUG:cmdstanpv:input tempfile: /tmp/tmp9puoo3fo/8olh95ve.ison
DEBUG:cmdstanpv:input tempfile: /tmp/tmp9puoo3fo/9f56ttem.ison
DEBUG:cmdstanpv:idx 0
DEBUG:cmdstanpy:running CmdStan, num threads: None
DEBUG:cmdstanpy:CmdStan args: ['/usr/local/lib/python3.7/dist-packages/prophet/stan model/prophet model.bin', 'random', 'seed=65235', 'data', 'file=/tmp/tmp9puoo3fo/8ol
h95ve.ison', 'init=/tmp/tmp9puoo3fo/9f56ttem.ison', 'output', 'file=/tmp/tmp9puoo3fo/prophet modelbx0vgih /prophet model-20221120134129.csv', 'method=optimize', 'algori
thm=lbfgs', 'iter=10000'l
13:41:29 - cmdstanpy - INFO - Chain [1] start processing
INFO:cmdstanpv:Chain [1] start processing
13:41:29 - cmdstanpy - INFO - Chain [1] done processing
INFO:cmdstanpv:Chain [1] done processing
/usr/local/lib/python3.7/dist-packages/prophet/hdays.py:209: Warning: We only support Diwali and Holi holidays from 2010 to 2030
 warnings.warn(warning msg, Warning)
/usr/local/lib/python3.7/dist-packages/prophet/hdays.py:209: Warning: We only support Diwali and Holi holidays from 2010 to 2030
  warnings.warn(warning msg, Warning)
```

#### In [395... print(final forecast nifty[['ds','yhat','yhat lower','yhat upper']])

```
ds
                    yhat yhat lower yhat upper
0 2022-11-21 18299.014561 17115.637728 19518.023350
1 2022-11-22 18377.627620 17161.071690 19618.496775
2 2022-11-23 18407.555434 17166.323913 19565.507564
3 2022-11-24 18382.413941 16998.169038 19544.338790
4 2022-11-25 18334.004754 17261.725373 19544.577826
```

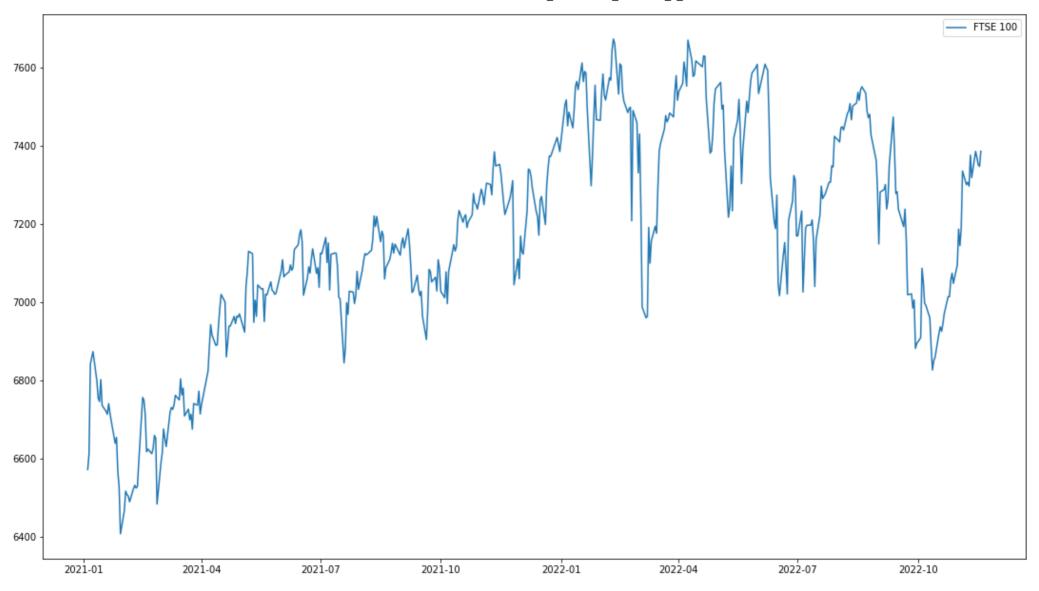
#### **FTSF 100**

```
In [376... #Reading FTSE File
        df FTSE = pd.read csv('/content/FTSE 100.csv')
        df FTSE["date clean"] = df FTSE["Date"].apply(lambda x: datetime.strptime(str(x), '%d-%m-%Y').date() )
        df FTSE['year']=df FTSE['date clean'].apply(lambda x: x.year)
        df FTSE['Month']=df FTSE['date clean'].apply(lambda x: x.month)
        df FTSE['day']=df FTSE['date clean'].apply(lambda x: x.day)
        #df FTSE=df FTSE[df FTSE['year'].isin([2021,2022])]
        df FTSE=df FTSE[((df FTSE['year']==2021)&(df FTSE['Month']>=1))]
        df FTSE.head()
```

Out[376]:		Date	Close	Open	High	Low	Volume	Chg%	date_clean	year	Month	day
	0	18-11-2022	7385.52	NaN	NaN	NaN	NaN	NaN	2022-11-18	2022	11	18
	1	17-11-2022	7346.54	NaN	NaN	NaN	NaN	NaN	2022-11-17	2022	11	17
	2	16-11-2022	7351.19	NaN	NaN	NaN	NaN	NaN	2022-11-16	2022	11	16
	3	15-11-2022	7369.44	NaN	NaN	NaN	NaN	NaN	2022-11-15	2022	11	15
	4	14-11-2022	7385 17	NaN	NaN	NaN	NaN	NaN	2022-11-14	2022	11	14

### **EDA**

```
from matplotlib import pyplot
fig = pyplot.gcf()
fig.set_size_inches(18.5, 10.5)
pyplot.plot(df_FTSE['date_clean'],df_FTSE["Close"], label='FTSE 100')
pyplot.legend()
pyplot.show()
```



```
def preprocessing(df):
    df=df.sort_values(by=["date_clean"],ascending=True)
    df=df.rename(columns={'date_clean':'ds','Close':'y'})
    df=df.reset_index(drop=True)
    df=df[['ds','y']]
    return df
DF_FTSE=preprocessing(df_FTSE)
```

```
def mean absolute percentage error(y true, y pred):
   v true, v pred = np.arrav(v true), np.arrav(v pred)
    return np.mean(np.abs((v true - v pred) / v true)) * 100
def ProhetModel(DF.CountryName):
 #train test split
  df2 test=DF[-5:]
  df2 train=DF[:-5]
  test val = df2 test.copv()
  train val = df2 train.copv()
  train val = train val.reset index(drop=True)
  test val = test val.reset index(drop=True)
  # Setup model with holidays
  final model = Prophet(#holidays=holiday,
                      changepoint prior scale= 0.03.
                      #holidays prior scale = 0.3,
                      n changepoints = 2.
                      seasonality mode = 'additive',
                      weekly seasonality=40.
                      daily seasonality = False.
                      yearly seasonality = 10,
                      #quarterly seasonality = True,
                      interval width=0.95)
 final model.add seasonality('quarterly', period=91.25, fourier_order=8, mode='additive')
  final model.add seasonality(name='monthly', period=30.5, fourier_order=5)
  #final model.add country holidays(country name=CountryName)
  #final model.add seasonality(name='quarterly', period=30.5, fourier order=5)
  final model.fit(train val)
 forecast = final model.predict(test val.drop("y", axis=1))
  mae = mean absolute error(test val["v"], list(forecast["vhat"]))
  print('MAE: %.3f' % mae)
  mape = mean absolute percentage error(test val["y"], list(forecast["yhat"]))
  print('MAPE: %.3f' % mape)
 se = np.square(test val["y"] -list(forecast["yhat"]))
  mse = np.mean(se)
 rmse = np.sqrt(mse)
 return mae, mape, forecast, test val, rmse, train val
mae ftse, mape ftse, forecast ftse, test val FTSE, rmse ftse, train val ftse = ProhetModel(DF FTSE, 'UK')
print('MAE ftse = ',mae ftse)
print('MAPE ftse = ', mape ftse)
print('RMSE ftse = ',rmse ftse)
print(forecast_ftse[['ds','yhat','yhat_lower','yhat_upper']])
print(test_val_FTSE)
from matplotlib import pyplot
pyplot.plot(test val FTSE["y"], label='Actual ftse')
```

```
Final Stockmarket 11202022 7 37
pyplot.plot(list(forecast ftse["vhat"]), label='Predicted ftse')
pvplot.legend()
pyplot.show()
DEBUG:cmdstanpy:input tempfile: /tmp/tmp9puoo3fo/cusr8rix.json
DEBUG:cmdstanpy:input tempfile: /tmp/tmp9puoo3fo/b34pi7vg.json
DEBUG:cmdstanpv:idx 0
DEBUG:cmdstanpv:running CmdStan, num threads: None
DEBUG:cmdstanpy:CmdStan args: ['/usr/local/lib/python3.7/dist-packages/prophet/stan model/prophet model.bin', 'random', 'seed=33038', 'data', 'file=/tmp/tmp9puoo3fo/cus
r8rix.json', 'init=/tmp/tmp9puoo3fo/b34pj7yq.json', 'output', 'file=/tmp/tmp9puoo3fo/prophet model-20221120133830.csv', 'method=optimize', 'algori
thm=lbfgs', 'iter=10000']
13:38:30 - cmdstanpy - INFO - Chain [1] start processing
INFO:cmdstanpv:Chain [1] start processing
13:38:30 - cmdstanpy - INFO - Chain [1] done processing
INFO:cmdstanpy:Chain [1] done processing
MAE: 25.050
MAPF: 0.340
MAE ftse = 25.050417109569025
MAPE ftse = 0.33975673492170333
RMSE ftse = 27.843986951419595
         ds
                    yhat yhat lower yhat upper
0 2022-11-14 7345.445940 7159.803447 7530.194008
1 2022-11-15 7359.222715 7165.318058 7551.775268
2 2022-11-16 7377.502255 7199.144106 7555.920267
3 2022-11-17 7358.765620 7178.117291 7546.892152
4 2022-11-18 7348.747134 7162.916099 7534.249042
          ds
0 2022-11-14 7385.17
1 2022-11-15 7369.44
2 2022-11-16 7351.19
3 2022-11-17 7346.54
4 2022-11-18 7385.52
7385
                        Actual ftse
                         Predicted ftse
7380
7375
7370
7365
7360
```

0.5 1.0 1.5

2.0 2.5 3.0 3.5

7355 7350 7345

### Predicting(FTSE) for final data 21th Nov to 25th Nov

```
In [379... import dateutil
         future = []
         for i in range(3,8,1):
           future.append(DF FTSE["ds"][DF FTSE.shape[0]-1] + dateutil.relativedelta.relativedelta(days=i))
         test ftse = pd.DataFrame({"ds":future})
         test ftse = test ftse.reset index(drop=True)
         print(test ftse)
                    ds
         0 2022-11-21
         1 2022-11-22
         2 2022-11-23
         3 2022-11-24
         4 2022-11-25
In [383... def ProhetModel test ftse(test,train,CountryName):
           #train test split
           test val = test.copy()
           train val = train.copy()
           train val = train val.reset index(drop=True)
            test val = test val.reset index(drop=True)
           # Setup model with holidays
           final model = Prophet(#holidays=holiday,
                                changepoint prior scale= 0.03,
                                #holidays prior scale = 0.3.
                                n changepoints = 2,
                                seasonality mode = 'additive',
                                weekly_seasonality=40,
                                daily seasonality = False,
                               yearly seasonality = 10,
                                #quarterly seasonality = True,
                               interval width=0.95)
           final_model.add_seasonality('quarterly', period=91.25, fourier_order=8, mode='additive')
            final model.add seasonality(name='monthly', period=30.5, fourier order=5)
           final model.fit(train val)
           forecast = final_model.predict(test_val)
           return forecast , test_val,train_val
In [386... train ftse=DF FTSE.copy()
         final forecast ftse, final test val ftse, final train val ftse = ProhetModel test ftse(test ftse, train ftse, 'UK')
```

```
DEBUG:cmdstanpv:input tempfile: /tmp/tmp9puoo3fo/8i6hs0pa.ison
         DEBUG:cmdstanpv:input tempfile: /tmp/tmp9puoo3fo/vgw9dw88.ison
         DEBUG:cmdstanpv:idx 0
         DEBUG:cmdstanpv:running CmdStan, num threads: None
         DEBUG:cmdstanpy:CmdStan args: ['/usr/local/lib/python3.7/dist-packages/prophet/stan model/prophet model.bin', 'random', 'seed=31856', 'data', 'file=/tmp/tmp9puoo3fo/8i6
         hs@pa.ison', 'init=/tmp/tmp9puoo3fo/vaw9dw88.ison', 'output', 'file=/tmp/tmp9puoo3fo/prophet modelwcxnvkhg/prophet model-20221120134024.csv', 'method=optimize', 'algori
         thm=lbfgs', 'iter=10000'l
         13:40:24 - cmdstanpy - INFO - Chain [1] start processing
         INFO:cmdstanpv:Chain [1] start processing
         13:40:24 - cmdstanpy - INFO - Chain [1] done processing
         INFO:cmdstanpy:Chain [1] done processing
        print(final forecast ftse[['ds','yhat','yhat lower','yhat upper']])
In Γ387...
                              yhat yhat lower yhat upper
                   ds
         0 2022-11-21 7332.379774 7151.228985 7513.454027
         1 2022-11-22 7338.851145 7153.357354 7515.184619
         2 2022-11-23 7345.383246 7164.901553 7526.269651
         3 2022-11-24 7317.045418 7140.575471 7506.172259
         4 2022-11-25 7304.885816 7124.510640 7501.027899
```

### Nasdaq 100

```
#Reading NASDAQ File

df_Nasdaq = pd.read_csv('/content/Nasdaq100.csv')

df_Nasdaq["date_clean"] = df_Nasdaq["Date"].apply(lambda x: datetime.strptime(str(x), '%d-%m-%Y').date())

df_Nasdaq['year']=df_Nasdaq['date_clean'].apply(lambda x: x.year)

df_Nasdaq['Nonth']=df_Nasdaq['date_clean'].apply(lambda x: x.month)

df_Nasdaq['day']=df_Nasdaq['date_clean'].apply(lambda x: x.day)

#df_Nasdaq=df_Nasdaq[df_Nasdaq['year'].isin([2022])]

#df_Nasdaq=df_Nasdaq[(df_Nasdaq['year']==2021)&(df_Nasdaq['Month']>9))((df_Nasdaq['year']==2022)&(df_Nasdaq['Month']>=1))]

#df_Nasdaq=df_Nasdaq=((df_Nasdaq['year']==2021)&(df_Nasdaq['Month']>=2))((df_Nasdaq['year']==2022)&(df_Nasdaq['Month']>=7)((df_Nasdaq['year']==2022)&(df_Nasdaq['Month']>=0))((df_Nasdaq['year']==2022)&(df_Nasdaq['Month']>=0))((df_Nasdaq['year']==2022)&(df_Nasdaq['Month']>=0))((df_Nasdaq['year']==2022)&(df_Nasdaq['Month']>=0))((df_Nasdaq['year']==2022)&(df_Nasdaq['Month']>=0))((df_Nasdaq['year']==2022)&(df_Nasdaq['Month']>=0)((df_Nasdaq['year']==2022)&(df_Nasdaq['Month']>=0)((df_Nasdaq['year']==2022)&(df_Nasdaq['Month']>=0)((df_Nasdaq['year']==2022)&(df_Nasdaq['Month']>=0)((df_Nasdaq['year']==2022)&(df_Nasdaq['Month']>=0)((df_Nasdaq['year']==2022)&(df_Nasdaq['Month']>=0)((df_Nasdaq['year']==2022)&(df_Nasdaq['Month']>=0)((df_Nasdaq['year']==2022)&(df_Nasdaq['Month']>=0)((df_Nasdaq['year']==2022)&(df_Nasdaq['Month']>=0)((df_Nasdaq['year']==2022)&(df_Nasdaq['Month']>=0)((df_Nasdaq['year']==2022)&(df_Nasdaq['Month']>=0)((df_Nasdaq['year']==2022)&(df_Nasdaq['Month']>=0)((df_Nasdaq['year']==2022)&(df_Nasdaq['Month']>=0)((df_Nasdaq['year']==2022)&(df_Nasdaq['Month']>=0)((df_Nasdaq['year']==2022)&(df_Nasdaq['Month']>=0)((df_Nasdaq['year']==2022)&(df_Nasdaq['Month']>=0)((df_Nasdaq['year']==2022)&((df_Nasdaq['year']==2022)&((df_Nasdaq['year']==2022)&((df_Nasdaq['year']==2022)&((df_Nasdaq['year']==2022)&((df_Nasdaq['year']==2022)&((df_Nasdaq['year']==2022)&((df_Nasdaq['year']==2022)&((df_Nasdaq['year']==2022)&((df_Nasdaq['year']==2022)&((df_Nasdaq['year']==2022)&((df_Nasdaq['
```

Out[396]:		Date	Close	Open	High	Low	Volume	Chg%	date_clean	year	Month	day
	0	18-11-2022	11677.02	NaN	NaN	NaN	NaN	NaN	2022-11-18	2022	11	18
	1	17-11-2022	11676.86	NaN	NaN	NaN	NaN	NaN	2022-11-17	2022	11	17
	2	16-11-2022	11699.09	NaN	NaN	NaN	NaN	NaN	2022-11-16	2022	11	16
	3	15-11-2022	11871.15	NaN	NaN	NaN	NaN	NaN	2022-11-15	2022	11	15
	4	14-11-2022	11700.94	NaN	NaN	NaN	NaN	NaN	2022-11-14	2022	11	14
	•••											
	555	08-09-2020	11068.26	11,143.91	11,400.45	11,055.16	338.09M	-4.77%	2020-09-08	2020	9	8
	556	04-09-2020	11622.13	11,686.24	11,846.18	11,145.99	364.55M	-1.27%	2020-09-04	2020	9	4
	557	03-09-2020	11771.36	12,195.75	12,235.00	11,662.10	359.31M	-5.23%	2020-09-03	2020	9	3
	558	02-09-2020	12420.54	12,417.45	12,439.48	12,176.18	273.59M	1.04%	2020-09-02	2020	9	2
	559	01-09-2020	12292.86	12,203.36	12,300.44	12,132.78	228.04M	1.50%	2020-09-01	2020	9	1

560 rows × 11 columns

```
def preprocessing(df):
    df=df.sort_values(by=["date_clean"],ascending=True)
    df=df.rename(columns={'date_clean':'ds','Close':'y'})
    df=df.reset_index(drop=True)
    df=df[['ds','y']]
    return df

DF_Nasdaq=preprocessing(df_Nasdaq)
DF_Nasdaq
```

```
        Out[332]:
        ds
        y

        0
        2020-09-01
        12292.86

        1
        2020-09-02
        12420.54

        2
        2020-09-03
        11771.36

        3
        2020-09-04
        11622.13

        4
        2020-09-08
        11068.26

        ...
        ...
        ...

        555
        2022-11-14
        11700.94

        556
        2022-11-15
        11871.15

        557
        2022-11-16
        11699.09

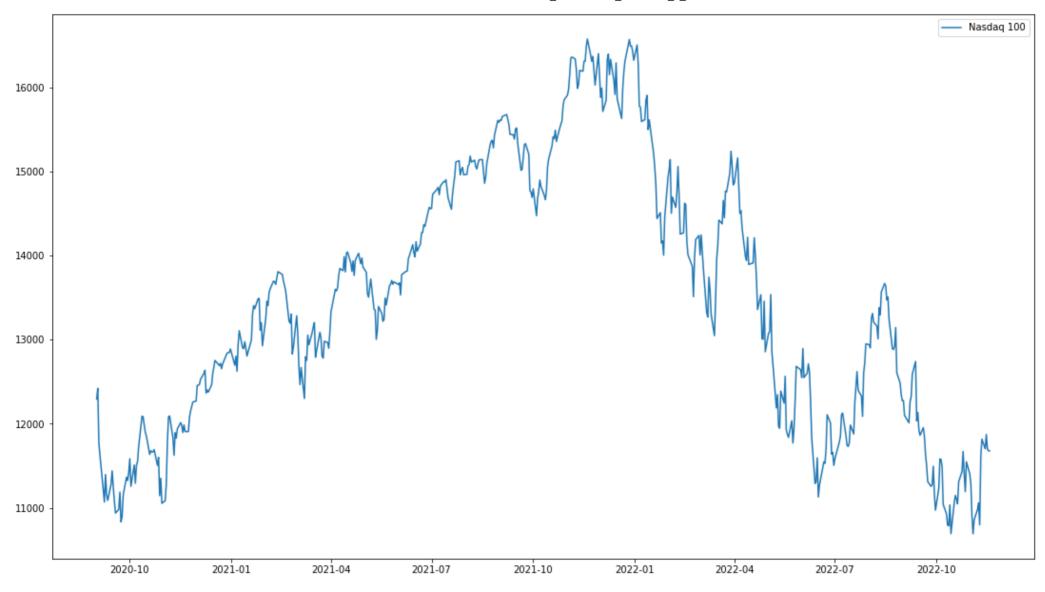
        558
        2022-11-17
        11676.86

        559
        2022-11-18
        11677.02
```

560 rows × 2 columns

#### **EDA**

```
from matplotlib import pyplot
fig = pyplot.gcf()
fig.set_size_inches(18.5, 10.5)
pyplot.plot(DF_Nasdaq['ds'],DF_Nasdaq["y"], label='Nasdaq 100')
pyplot.legend()
pyplot.show()
```



```
def ProhetModel_nasdaq(DF,CountryName):
    #train test split
    df2_test=DF[-5:]
    df2_train=DF[:-5]
    test_val = df2_test.copy()
    train_val = df2_train.copy()
    train_val = train_val.reset_index(drop=True)
    test_val = test_val.reset_index(drop=True)
```

```
# Setup model with holidays
  final model = Prophet(#holidays=holiday.
                      changepoint prior scale= 0.6.
                      #holidays prior scale = 0.3.
                      n changepoints = 5.
                      seasonality mode = 'additive',
                      weeklv seasonalitv=5.
                      daily seasonality = 10.
                      vearlv seasonalitv = True.
                      interval width=0.95)
  final model.add country holidays(country name=CountryName)
 final model.add seasonality('quarterly', period=91.25, fourier_order=8, mode='additive')
  final model.add seasonality(name='monthly', period=30.5, fourier order=5)
  #final model.add regressor('regressor', mode='additive')
  #final model.add seasonality(name='quarterly', period=30.5,fourier order=5)
  final model.fit(train val)
  #FutureDates nasdaa=final model.make future dataframe(periods=1)
  forecast = final model.predict(test val.drop("y", axis=1))
  mae = mean absolute error(test val["v"], list(forecast["vhat"]))
  print('MAE: %.3f' % mae)
  mape = mean absolute percentage error(test val["v"], list(forecast["vhat"]))
  print('MAPE: %.3f' % mape)
  se = np.square(test val["v"] -list(forecast["vhat"]))
  mse = np.mean(se)
 rmse = np.sart(mse)
  return mae, mape, forecast, test val, rmse, train val
mae nasdaq, mape nasdaq, forecast nasdaq,test val nasdaq,rmse nasdaq,train val nasdaq = ProhetModel nasdaq(DF Nasdaq,'US')
print('MAE Nasdag = ',mae nasdag)
print('MAPE Nasdag = ', mape nasdag)
print('RMSE Nasdaq = ',rmse nasdaq)
print(forecast nasdag[['ds','yhat','yhat lower','yhat upper']])
print(test val nasdag)
from matplotlib import pyplot
pyplot.plot(test val nasdaq["y"], label='Actual nasdaq')
pyplot.plot(list(forecast nasdag["yhat"]), label='Predicted nasdag')
pyplot.legend()
pyplot.show()
```

```
DEBUG:cmdstanpv:input tempfile: /tmp/tmp9puoo3fo/6slg ppw.ison
DEBUG:cmdstanpy:input tempfile: /tmp/tmp9puoo3fo/r8h3nd89.json
DEBUG:cmdstanpv:idx 0
DEBUG:cmdstanpv:running CmdStan, num threads: None
DEBUG:cmdstanpy:CmdStan args: ['/usr/local/lib/python3.7/dist-packages/prophet/stan model/prophet model.bin', 'random', 'seed=94025', 'data', 'file=/tmp/tmp9puoo3fo/6sl
q ppw.json', 'init=/tmp/tmp9puoo3fo/r8h3nd89.json', 'output', 'file=/tmp/tmp9puoo3fo/prophet model-20221120132659.csv', 'method=optimize', 'algori
thm=lbfgs', 'iter=10000']
13:26:59 - cmdstanpy - INFO - Chain [1] start processing
INFO:cmdstanpv:Chain [1] start processing
13:26:59 - cmdstanpy - INFO - Chain [1] done processing
INFO:cmdstanpy:Chain [1] done processing
MAF: 101.147
MAPE: 0.862
MAE Nasdag = 101.14739006588715
MAPE Nasdag = 0.8615145731510987
RMSE Nasdag = 104.69679070484187
          ds
                     yhat yhat lower
                                          yhat upper
0 2022-11-14 11609.856679 10883.531257 12276.940227
1 2022-11-15 11725.601947 10970.311106 12425.273763
2 2022-11-16 11792.375573 11068.767770 12571.762469
3 2022-11-17 11788.937624 11074.986754 12575.020092
4 2022-11-18 11740.762380 10944.539819 12424.424048
           ds
0 2022-11-14 11700.94
1 2022-11-15 11871.15
2 2022-11-16 11699.09
3 2022-11-17 11676.86
4 2022-11-18 11677.02
                                     Actual nasdag
11850
                                      Predicted nasdag
11800
11750
11700
11650
```

## Predicting(NASDAQ) for final data 21th Nov to 25th Nov

3.5

1.0

1.5

2.0

2.5

3.0

11600

```
In [372...
         import dateutil
         future = []
         for i in range(3.8.1):
           future.append(DF Nasdag["ds"][DF Nasdag.shape[0]-1] + dateutil.relativedelta.relativedelta(days=i))
         test nasdag = pd.DataFrame({"ds":future})
         test nasdag.drop(test nasdag.index[3], inplace=True)
         test nasdag = test nasdag.reset index(drop=True)
         print(test nasdag)
                    ds
         0 2022-11-21
         1 2022-11-22
         2 2022-11-23
         3 2022-11-25
In [370... def ProhetModel test nasdag(test,train,CountryName):
           #train test split
           test val = test.copv()
           train val = train.copv()
           train val = train val.reset index(drop=True)
           test val = test val.reset index(drop=True)
           # Setup model with holidays
           final model = Prophet(#holidays=holiday,
                                changepoint prior scale= 0.6,
                                #holidays prior scale = 0.3.
                                n changepoints = 5.
                               seasonality mode = 'additive',
                                weekly seasonality=5,
                                daily seasonality = 10,
                                vearly seasonality = True,
                               interval width=0.95)
           final model.add country holidays(country name=CountryName)
           final model.add seasonality('quarterly', period=91.25, fourier order=8, mode='additive')
           final model.add seasonality(name='monthly', period=30.5, fourier order=5)
           #final model.add regressor('regressor', mode='additive')
           #final model.add seasonality(name='quarterly', period=30.5,fourier order=5)
           final model.fit(train val)
           #FutureDates nasdaq=final model.make future dataframe(periods=1)
           #forecast = final model.predict(test val.drop("y", axis=1))
           forecast = final model.predict(test val)
           return forecast , test val,train val
         train_nasdaq=DF_Nasdaq.copy()
         final forecast nasdaq, final test val nasdaq, final train val nasdaq = ProhetModel test nasdaq(test nasdaq, train nasdaq, 'UK')
```

```
DEBUG:cmdstanpy:input tempfile: /tmp/tmp9puoo3fo/f327x1o4.json
DEBUG:cmdstanpy:input tempfile: /tmp/tmp9puoo3fo/v9v1vxfa.json
DEBUG:cmdstanpy:idx 0
DEBUG:cmdstanpy:running CmdStan, num_threads: None
DEBUG:cmdstanpy:CmdStan args: ['/usr/local/lib/python3.7/dist-packages/prophet/stan_model/prophet_model.bin', 'random', 'seed=69377', 'data', 'file=/tmp/tmp9puoo3fo/f32
7x1o4.json', 'init=/tmp/tmp9puoo3fo/v9v1vxfa.json', 'output', 'file=/tmp/tmp9puoo3fo/prophet_modeliodz_3im/prophet_model-20221120132716.csv', 'method=optimize', 'algori
thm=lbfgs', 'iter=10000']
13:27:16 - cmdstanpy - INFO - Chain [1] start processing
INFO:cmdstanpy:Chain [1] start processing
13:27:16 - cmdstanpy - INFO - Chain [1] done processing
INFO:cmdstanpy:Chain [1] done processing
```

#### In [371... print(final\_forecast\_nasdaq[['ds','yhat','yhat\_lower','yhat\_upper']])

```
ds yhat yhat_lower yhat_upper 0 2022-11-21 11751.804315 10996.709732 12444.915795 1 2022-11-22 11800.889696 11116.932340 12549.528707 2 2022-11-23 11824.891905 11081.122156 12561.198107 3 2022-11-25 11849.687751 11116.620986 12545.868979
```