Prophet is a procedure for forecasting time series data based on an additive model where non-linear trends are fit with yearly, weekly, and daily seasonality, plus holiday effects. It works best with time series that have strong seasonal effects and several seasons of historical data. Prophet is robust to missing data and shifts in the trend, and typically handles outliers well.

YFinance not only downloads the Stock Price data it also allows us to download all the financial data of a Company since its listing in the stock market.

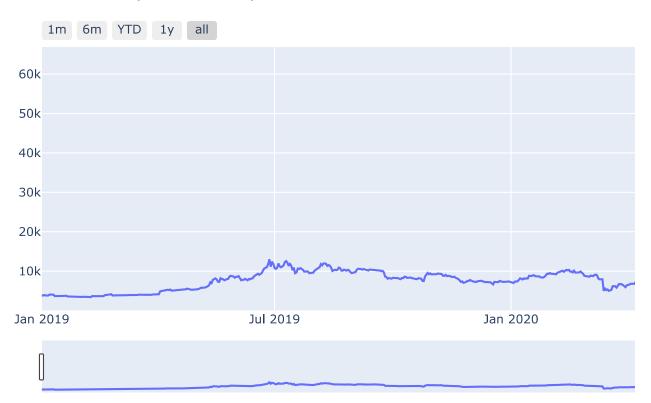
```
today = datetime.today().strftime('%Y-%m-%d')
initial_date = '2019-01-01'

df = yf.download('BTC-USD',initial_date, today)

df.tail()
#df.info()
```

```
[********* 100%********** 1 of 1 completed
                                                      Close Adj Close
                                                                              Volume
                      0pen
                                 High
                                            Low
print(df.columns)
     Index(['Open', 'High', 'Low', 'Close', 'Adj Close', 'Volume'], dtype='object')
      2021-09-02 $48.807.85 $50.343.42 $48.652.32 $49.327.72 $49.327.72 39508070319
df.reset_index(inplace=True)
print(df.columns)
     Index(['Date', 'Open', 'High', 'Low', 'Close', 'Adj Close', 'Volume'], dtype='object')
new_df = df[["Date","Open"]]
new names = {
    "Date": "ds",
    "Open": "y",
}
new_df.rename(columns=new_names, inplace=True)
print(new_df.tail())
                 ds
     969 2021-08-31 $47,024.34
     970 2021-09-01 $47,099.77
     971 2021-09-02 $48,807.85
     972 2021-09-03 $49,288.25
     973 2021-09-04 $49,922.36
x = \text{new df}[\text{"ds"}]
y = \text{new df}["y"]
fig = go.Figure()
fig.add trace(go.Scatter(x=x, y=y))
# Set title
fig.update_layout(
    title text="Time series plot of BTC Open Price",
)
fig.update_layout(
    xaxis=dict(
        rangeselector=dict(
            buttons=list(
                Γ
                    dict(count=1, label="1m", step="month", stepmode="backward"),
                    dict(count=6, label="6m", step="month", stepmode="backward"),
                    dict(count=1, label="YTD", step="year", stepmode="todate"),
                    dict(count=1. label="1v". sten="vear". stenmode="hackward")
```

## Time series plot of BTC Open Price



```
m = Prophet(
    seasonality_mode="multiplicative"
)
m.fit(new_df)

INFO:fbprophet:Disabling daily seasonality. Run prophet with daily_seasonality=True to 
<fbprophet.forecaster.Prophet at 0x7f0a496090d0>
future = m.make_future_dataframe(periods = 365)
```

future.tail()

```
ds

1334 2022-08-31

1335 2022-09-01

1336 2022-09-02

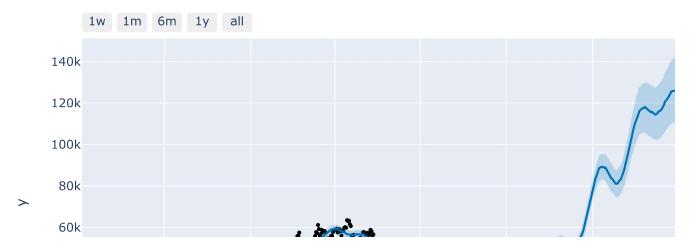
1337 2022-09-03

1338 2022-09-04
```

```
forecast = m.predict(future)
forecast[['ds', 'yhat', 'yhat_lower', 'yhat_upper']].tail()
```

_		ds	yhat	yhat_lower	yhat_upper
•	1334	2022-08-31	\$82,967.83	\$62,401.34	\$104,445.29
	1335	2022-09-01	\$82,993.93	\$63,319.22	\$104,194.60
	1336	2022-09-02	\$82,438.56	\$62,806.70	\$103,202.67
	1337	2022-09-03	\$82,748.57	\$63,146.70	\$103,384.46
	1338	2022-09-04	\$81,955.94	\$62,544.60	\$102,628.90

plot\_plotly(m, forecast)



plot\_components\_plotly(m, forecast)



