→ Stock Price prediction using Prophet

```
#Importing all the necessary libraries

import pandas as pd
import plotly.express as px
from prophet import Prophet

#Initializing Plotly
import plotly.io as pio
pio.renderers.default='colab'

FVRR_df=pd.read_csv("FVRR.csv")
QFIN_df=pd.read_csv("FVRR.csv")
ROKU_df=pd.read_csv("ROKU.csv")
SAVA_df=pd.read_csv("SAVA.csv")

TWLO_df=pd.read_csv("TWLO.csv")

from google.colab import drive
drive.mount('/content/drive')
```

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).

#read_csv function from pandas
FVRR_df

	Date	0pen	High	Low	Close	Adj Close	Volume	7		
0	2019-06-13	26.000000	41.680000	25.549999	39.900002	39.900002	22046000			
1	2019-06-14	41.639999	44.250000	31.490000	31.490000	31.490000	10275800			
2	2019-06-17	32.810001	34.880001	31.541000	34.380001	34.380001	3789100			
3	2019-06-18	35.259998	35.919998	31.040001	31.150000	31.150000	3274000			
4	2019-06-19	31.799999	31.799999	28.250000	28.670000	28.670000	1902000			
960	2023-04-05	34.389999	34.639999	33.365002	34.029999	34.029999	1394900			
961	2023-04-06	33.849998	34.889999	32.880001	34.529999	34.529999	481100			
962	2023-04-10	35.110001	35.369999	34.310001	35.049999	35.049999	434900			
963	2023-04-11	35.119999	35.790001	34.869999	35.430000	35.430000	527100			
964	2023-04-12	36.230000	36.750000	34.650002	34.830002	34.830002	590500			
965 rd	965 rows × 7 columns									

QFIN_df



	Date	0pen	High	Low	Close	Adj Close	Volume	1
0	2019-06-13	104.849998	105.400002	102.540001	104.970001	104.970001	7629200	
1	2019-06-14	104.330002	106.120003	101.980003	102.019997	102.019997	7326900	
2	2019-06-17	101.550003	104.330002	101.010002	103.839996	103.839996	5528300	
3	2019-06-18	104.750000	104.879997	101.169998	104.389999	104.389999	6591700	
4	2019-06-19	104.070000	106.550003	102.732002	106.489998	106.489998	5663000	
960	2023-04-05	65.099998	65.389999	61.119999	61.270000	61.270000	9049600	
961	2023-04-06	60.950001	64.169998	59.459999	64.080002	64.080002	9170100	
962	2023-04-10	63.040001	64.449997	61.790001	64.320000	64.320000	7988400	
963	2023-04-11	64.320000	64.980003	62.639999	63.970001	63.970001	5914800	
964	2023-04-12	65.470001	65.470001	60.419998	60.470001	60.470001	9032800	
965 rd	ows × 7 colum	ns						

965 rows × 7 columns

SAVA_df

	Date	0pen	High	Low	Close	Adj Close	Volume	i
0	2019-06-13	1.110000	1.12	1.100000	1.110000	1.110000	77100	
1	2019-06-14	1.100000	1.22	1.100000	1.200000	1.200000	281600	
2	2019-06-17	1.230000	1.23	1.150000	1.190000	1.190000	70600	
3	2019-06-18	1.180000	1.22	1.110000	1.140000	1.140000	324900	
4	2019-06-19	1.130000	1.16	1.130000	1.160000	1.160000	104500	
960	2023-04-05	23.879999	24.08	23.309999	23.840000	23.840000	411200	
961	2023-04-06	23.930000	24.18	23.410000	23.969999	23.969999	426100	
962	2023-04-10	23.879999	23.93	22.420000	23.110001	23.110001	950300	
963	2023-04-11	23.100000	23.82	23.049999	23.400000	23.400000	571000	
964	2023-04-12	23.510000	23.84	22.850000	22.900000	22.900000	432500	
06E ==	v 7 aaliim							

965 rows × 7 columns

TWLO_df

```
1
                Date
                           0pen
                                      High
                                                   Low
                                                            Close Adj Close Volume
FVRR_df.info()
QFIN_df.info()
ROKU_df.info()
SAVA df.info()
TWLO_df.info()
     dtypes: float64(5), int64(1), object(1)
    memory usage: 52.9+ KB
     <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 964 entries, 0 to 963
    Data columns (total 7 columns):
        Column
                    Non-Null Count Dtype
     0
                     964 non-null
                                    object
         Date
      1
         0pen
                    964 non-null
                                    float64
      2
                     964 non-null
                                     float64
         High
                     964 non-null
                                    float64
      3
         Low
                    964 non-null
      4
                                    float64
         Close
         Adj Close 964 non-null
                                     float64
                    964 non-null
         Volume
                                    int64
     dtypes: float64(5), int64(1), object(1)
    memory usage: 52.8+ KB
     <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 965 entries, 0 to 964
    Data columns (total 7 columns):
         Column
                    Non-Null Count Dtype
                     965 non-null
     0
         Date
                                    object
      1
         0pen
                    965 non-null
                                    float64
      2
         High
                     965 non-null
                                     float64
                    965 non-null
                                    float64
      3
         Low
      4
                    965 non-null
                                    float64
         Close
         Adj Close 965 non-null
                                     float64
                    965 non-null
         Volume
                                    int64
     dtypes: float64(5), int64(1), object(1)
     memory usage: 52.9+ KB
     <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 965 entries, 0 to 964
    Data columns (total 7 columns):
         Column
                    Non-Null Count Dtype
     0
                    965 non-null
                                    object
         Date
      1
         0pen
                    965 non-null
                                    float64
      2
                     965 non-null
                                     float64
         High
                    965 non-null
                                    float64
      3
         Low
                    965 non-null
                                    float64
         Close
         Adj Close 965 non-null
                                     float64
         Volume
                    965 non-null
                                    int64
     dtypes: float64(5), int64(1), object(1)
    memory usage: 52.9+ KB
     <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 965 entries, 0 to 964
    Data columns (total 7 columns):
         Column
                    Non-Null Count Dtype
          -----
     0
                     965 non-null
         Date
                                    object
      1
         0pen
                    965 non-null
                                    float64
                     965 non-null
                                    float64
      2
         High
                    965 non-null
                                    float64
      3
         Low
         Close
                    965 non-null
                                    float64
         Adj Close
                    965 non-null
                                     float64
         Volume
                    965 non-null
                                     int64
    dtypes: float64(5), int64(1), object(1)
     memory usage: 52.9+ KB
FVRR_df.describe()
```

17/23, 11	, 11:21 PM Untitled4.ip					ntitled4.ipynb - 0	Colabora	
		Open	High	Low	Close	Adj Close	Volume	%
	count	965.000000	965.000000	965.000000	965.000000	965.000000	9.650000e+02	
	mean	95.570161	98.453501	92.323667	95.448663	95.448663	8.063907e+05	
	std	78.452421	80.577269	75.831312	78.323589	78.323589	9.852608e+05	
	min	17.500000	18.170000	17.110001	17.680000	17.680000	3.170000e+04	
QFIN_c	df.desc	ribe()						
		Open	High	Low	Close	Adj Close	Volume	<i>7</i> .
	count	964.000000	964.000000	964.000000	964.000000	964.000000	9.640000e+02	
	mean	16.266308	16.782992	15.718861	16.260830	15.347037	1.477851e+06	
	std	6.956765	7.235050	6.614239	6.953779	6.526019	1.620863e+06	
	min	6.500000	7.030000	6.370000	6.510000	6.049567	4.560000e+04	
	25%	10.760000	11.065000	10.527500	10.787500	10.045439	5.647250e+05	
	50%	14.330000	14.775000	13.915000	14.340000	13.765850	9.627000e+05	
	75%	20.925000	21.492500	20.205001	20.830000	19.601273	1.764275e+06	
	max	44.285000	45.000000	42.610001	44.049999	40.934464	2.344910e+07	
ROKU_c	df.desc	ribe()						
		0pen	High	Low	Close	Adj Close	Volume	1
	count	965.000000	965.000000	965.000000	965.000000	965.000000	9.650000e+02	
	mean	180.436633	185.262047	175.461893	180.299166	180.299166	8.852012e+06	
	std	116.726871	119.095466	114.048885	116.507743	116.507743	7.345221e+06	
	min	39.169998	39.889999	38.259998	38.799999	38.799999	1.443700e+06	
	25%	92.730003	96.764999	89.720001	92.900002	92.900002	4.396200e+06	
	50%	134.509995	138.509995	131.119995	134.639999	134.639999	6.796000e+06	
	75%	276.029999	280.600006	271.339996	276.459991	276.459991	1.052060e+07	
	max	477.200012	490.760986	468.779999	479.500000	479.500000	6.658520e+07	
SAVA_c	df.desc	ribe()						
		0pen	High	Low	Close	Adj Close	Volume	%
	count	965.000000	965.000000	965.000000	965.000000	965.000000	9.650000e+02	
	mean	27.206373	28.855047	25.849959	27.257389	27.257389	3.387860e+06	
	std	25.493375	27.271539	23.935321	25.481802	25.481802	9.279942e+06	
	min	1.050000	1.100000	1.030000	1.040000	1.040000	1.620000e+04	
	25%	4.850000	5.410000	4.630000	4.860000	4.860000	7.577000e+05	
	50%	24.500000	25.410000	23.879999	24.610001	24.610001	1.312700e+06	
	75%	42.070000	44.599998	40.799999	42.349998	42.349998	2.777500e+06	
	may	145 000000	146 160004	126 020000	135 300003	135 300003	1 7500720+08	

max 145.000000 146.160004 126.029999 135.300003 135.300003 1.750972e+08

TWLO_df.describe()

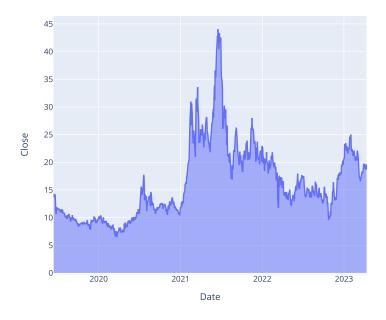
	0pen	High	Low	Close	Adj Close	Volume	1
count	965.000000	965.000000	965.000000	965.000000	965.000000	9.650000e+02	
mean	194.543392	198.973826	189.756916	194.367616	194.367616	3.345260e+06	
min	43.290001	44.160000	41.000000	42.740002	42.740002	6.775000e+05	

Data Visualization using plotly express- Visualizing the historical performance of the stocks

```
max 441 000000 457 299988 437 000000 443 489990 4484080e+07

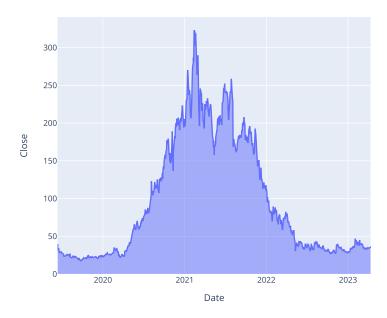
#Line graph, Area graph , box plot (Analyzing price and volume)

px.area(QFIN_df, x="Date", y="Close")
```

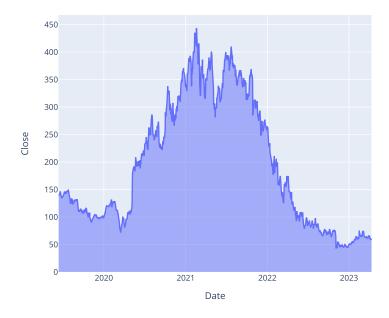


px.area(ROKU_df, x="Date", y="Close")

px.area(FVRR_df, x="Date", y="Close")

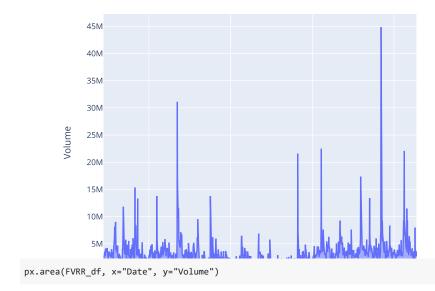


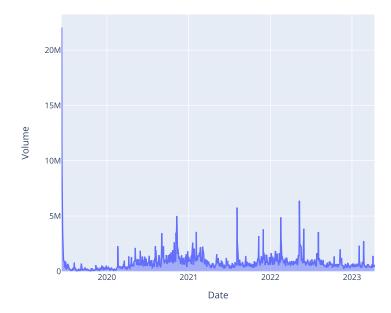
px.area(TWLO_df, x="Date", y="Close")



Double-click (or enter) to edit

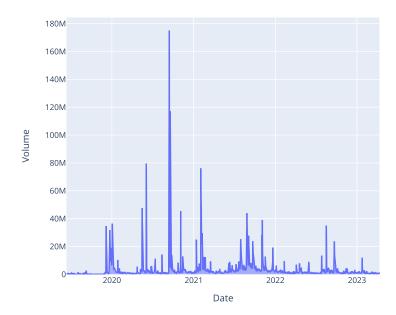
px.area(TWLO_df, x="Date", y="Volume")



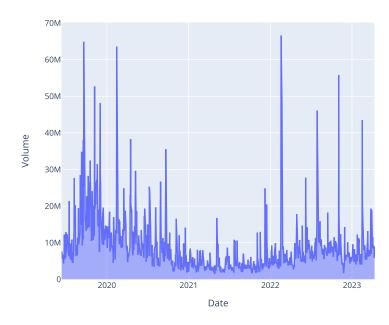


px.area(QFIN_df, x="Date", y="Volume")

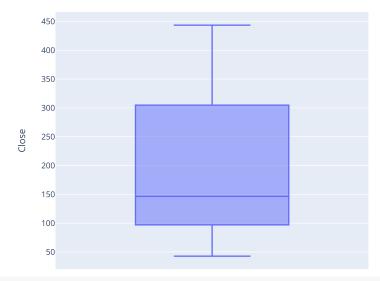




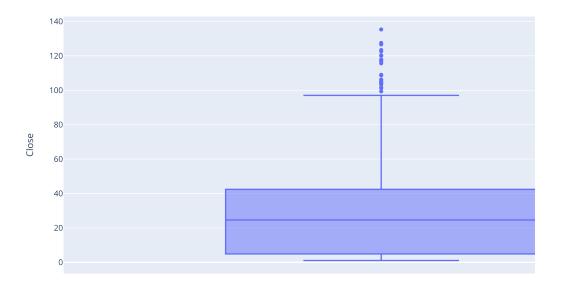
px.area(ROKU_df, x="Date", y="Volume")



px.box(TWLO_df, y="Close")



px.box(SAVA_df, y="Close")



px.box(ROKU_df, y="Close")





px.box(QFIN_df, y="Close")



Data Preperation

```
columns=['Date', "Close"]
ndf = pd.DataFrame(TWLO_df, columns =columns)
ndf
                                    10
                Date
                           Close
           2019-06-13 141.059998
       0
           2019-06-14 140.169998
           2019-06-17 140.710007
           2019-06-18 142.289993
           2019-06-19 146.500000
      960
           2023-04-05
                       59.320000
      961 2023-04-06
                       60.759998
      962 2023-04-10
                       59.639999
      963 2023-04-11
                       59.040001
      964 2023-04-12
                       58.130001
     965 rows × 2 columns
prophet_df = ndf.rename (columns = {'Date':'ds','Close': 'y' })
prophet_df
                                    1
                  ds
           2019-06-13 141.059998
           2019-06-14 140.169998
           2019-06-17 140.710007
           2019-06-18 142.289993
           2019-06-19 146.500000
       ...
      960 2023-04-05
                       59.320000
      961 2023-04-06
                       60.759998
      962 2023-04-10
                       59.639999
      963 2023-04-11
                       59.040001
      964 2023-04-12 58.130001
     965 rows × 2 columns
# Creating Prophet Model
m = Prophet()
m.fit(prophet_df)
     INFO:prophet:Disabling daily seasonality. Run prophet with daily_seasonality=True to override this.
     {\tt DEBUG:cmdstanpy:input\ tempfile:\ /tmp/tmpeqadd0kw/gjapk1od.json}
     DEBUG:cmdstanpy:input tempfile: /tmp/tmpeqadd0kw/9svs_r1u.json
     DEBUG:cmdstanpy:idx 0
     {\tt DEBUG:cmdstanpy:running\ CmdStan,\ num\_threads:\ None}
     DEBUG:cmdstanpy:CmdStan args: ['/usr/local/lib/python3.9/dist-packages/prophet/stan_model/prophet_model.bin', 'random', 'seed=61893', '
     17:15:51 - cmdstanpy - INFO - Chain [1] start processing
     INFO:cmdstanpy:Chain [1] start processing
     17:15:51 - cmdstanpy - INFO - Chain [1] done processing
     INFO:cmdstanpy:Chain [1] done processing
     cprophet.forecaster.Prophet at 0x7f9ab3895970>
```

```
future = m.make_future_dataframe(periods=365*5)
forecast_TWLO= m.predict(future)
```

 ${\tt forecast_TWLO}$

	ds	trend	yhat_lower	yhat_upper	trend_lower	trend_upper	additive_terms	additive_t€				
0	2019- 06-13	118.093845	96.476678	141.976335	118.093845	118.093845	1.423845					
1	2019- 06-14	118.075798	100.318063	142.224496	118.075798	118.075798	2.582881					
2	2019- 06-17	118.021659	102.307748	146.829930	118.021659	118.021659	6.382362					
3	2019- 06-18	118.003612	103.340396	148.284609	118.003612	118.003612	7.895676					
4	2019- 06-19	117.985565	106.403811	148.992951	117.985565	117.985565	8.977375					
2785	2028- 04-06	79.466711	-1502.214041	1735.614311	-1489.846815	1764.378874	-15.090059					
2786	2028- 04-07	79.471694	-1505.397939	1747.772583	-1491.432313	1765.004340	-14.643960					
2787	2028- 04-08	79.476676	-1488.747977	1763.607902	-1493.017811	1765.629807	0.979862					
2788	2028- 04-09	79.481658	-1486.464217	1764.877773	-1494.603310	1766.255273	1.374563					
2789	2028- 04-10	79.486641	-1512.587154	1741.265443	-1496.188808	1766.880739	-13.681702					
2790 rd	2790 rows × 19 columns											

1

Display the underlying forecast dataframe (tail)
forecast_TWLO[['ds', 'yhat', 'yhat_lower', 'yhat_upper']].tail()

```
        ds
        yhat
        yhat_lower
        yhat_upper

        2785
        2028-04-06
        64.376652
        -1502.214041
        1735.614311

        2786
        2028-04-07
        64.827734
        -1505.397939
        1747.772583

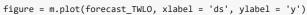
        2787
        2028-04-08
        80.456538
        -1488.747977
        1763.607902

        2788
        2028-04-09
        80.856221
        -1486.464217
        1764.877773

        2789
        2028-04-10
        65.804938
        -1512.587154
        1741.265443
```

px.line(forecast_TWLO, x='ds',y='yhat')





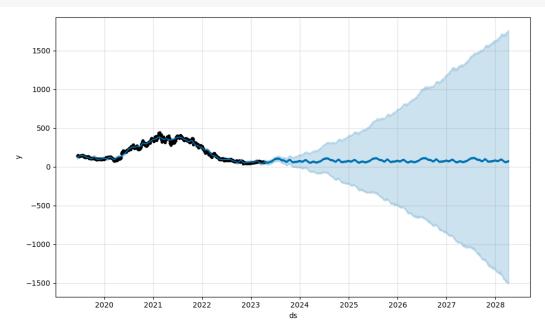


figure2=m.plot_components(forecast_TWLO)

```
1500
          1000
           500
             0
          -500
         -1000
         -1500
                       2020
                                2021
                                          2022
                                                    2023
                                                             2024
                                                                       2025
                                                                                 2026
                                                                                          2027
                                                                                                    2028
                                                             ds
          10.0
           7.5
columns=['Date', "Close"]
ndf = pd.DataFrame(FVRR_df, columns =columns)
prophet_FVRR_df = ndf.rename (columns = {'Date':'ds','Close': 'y' })
# Creating Prophet Model
m = Prophet()
m.fit(prophet_FVRR_df)
     INFO:prophet:Disabling daily seasonality. Run prophet with daily_seasonality=True to override this.
     {\tt DEBUG:cmdstanpy:input\ tempfile:\ /tmp/tmpeqadd0kw/fafddmtu.json}
     DEBUG:cmdstanpy:input tempfile: /tmp/tmpeqadd0kw/s0i1jfiv.json
     DEBUG:cmdstanpy:idx 0
     {\tt DEBUG:cmdstanpy:running\ CmdStan,\ num\_threads:\ None}
     DEBUG:cmdstanpy:CmdStan args: ['/usr/local/lib/python3.9/dist-packages/prophet/stan_model/prophet_model.bin', 'random', 'seed=96934', '
     17:15:54 - cmdstanpy - INFO - Chain [1] start processing
     INFO:cmdstanpy:Chain [1] start processing
     17:15:55 - cmdstanpy - INFO - Chain [1] done processing
     INFO:cmdstanpy:Chain [1] done processing
     cprophet.forecaster.Prophet at 0x7f9ab039aee0>
     4
future = m.make_future_dataframe(periods=365*5)
forecast_FVRR= m.predict(future)
forecast_FVRR
```

	ds	trend	yhat_lower	yhat_upper	trend_lower	trend_upper	additive_terms	additive_te
0	2019- 06-13	25.175468	3.201889	37.567580	25.175468	25.175468	-5.338177	
1	2019- 06-14	25.165478	3.502947	37.274032	25.165478	25.165478	-4.372060	

Display the underlying forecast dataframe (tail)
forecast_FVRR[['ds', 'yhat', 'yhat_lower', 'yhat_upper']].tail()

	ds	yhat	yhat_lower	yhat_upper
2785	2028-04-06	-17.282245	-1393.897618	1305.706130
2786	2028-04-07	-16.884033	-1386.912969	1301.914866
2787	2028-04-08	-7.049663	-1377.089628	1310.823308
2788	2028-04-09	-7.044627	-1384.909096	1314.678813
2789	2028-04-10	-17.672504	-1400.725390	1309.141639
	VT-V1			

px.line(forecast_FVRR, x='ds',y='yhat')

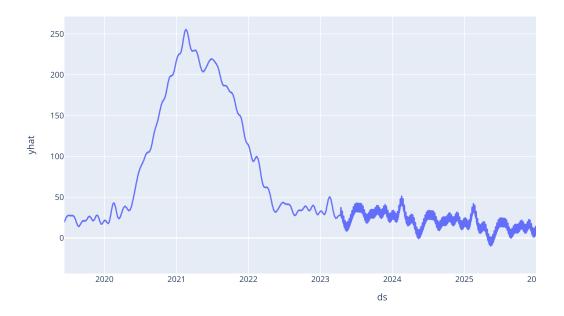
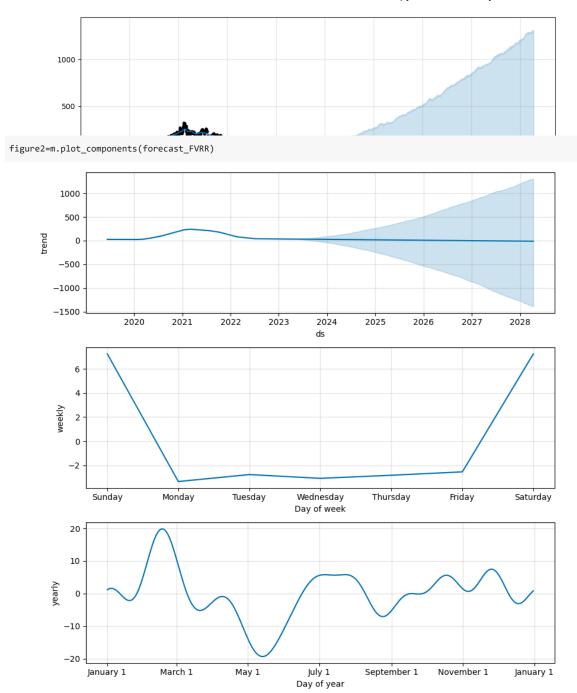


figure = m.plot(forecast_FVRR, xlabel = 'ds', ylabel = 'y')



```
columns=['Date', "Close"]
ndf = pd.DataFrame(ROKU_df, columns =columns)

prophet_ROKU_df = ndf.rename (columns = {'Date':'ds','Close': 'y' })

# Creating Prophet Model

m = Prophet()
m.fit(prophet_ROKU_df)

INFO:prophet:Disabling daily seasonality. Run prophet with daily_seasonality=True to override this.
DEBUG:cmdstanpy:input tempfile: /tmp/tmpeqadd0kw/jgtv66hw.json
```

DEBUG:cmdstanpy:input tempfile: /tmp/tmpeqadd0kw/beq39rm2.json

DEBUG:cmdstanpy:running CmdStan, num_threads: None

DEBUG:cmdstanpy:idx 0

DEBUG:cmdstanpy:CmdStan args: ['/usr/local/lib/python3.9/dist-packages/prophet/stan_model/prophet_model.bin', 'random', 'seed=74627', '17:15:59 - cmdstanpy - INFO - Chain [1] start processing INFO:cmdstanpy:Chain [1] start processing 17:15:59 - cmdstanpy - INFO - Chain [1] done processing INFO:cmdstanpy:Chain [1] done processing cprophet.forecaster.Prophet at 0x7f9ab00d23d0>

future = m.make_future_dataframe(periods=365*5) forecast_ROKU= m.predict(future)

forecast_ROKU

	ds	trend	yhat_lower	yhat_upper	trend_lower	trend_upper	additive_terms	additive_te			
0	2019- 06-13	95.568665	54.815878	110.998125	95.568665	95.568665	-11.421010				
1	2019- 06-14	95.841312	57.121079	111.278049	95.841312	95.841312	-10.434995				
2	2019- 06-17	96.659251	63.111600	119.585030	96.659251	96.659251	-5.038165				
3	2019- 06-18	96.931897	66.674354	121.115988	96.931897	96.931897	-2.854840				
4	2019- 06-19	97.204543	67.202674	121.007794	97.204543	97.204543	-1.770651				
2785	2028- 04-06	-69.117646	-2570.760262	2136.060850	-2558.184390	2144.822144	-13.386239				
2786	2028- 04-07	-69.184856	-2555.111452	2133.083051	-2562.812904	2147.515832	-13.119558				
2787	2028- 04-08	-69.252065	-2546.477668	2137.460434	-2566.112496	2150.209520	-6.213885				
2788	2028- 04-09	-69.319275	-2562.704950	2146.668549	-2567.946195	2152.903208	-5.627422				
2789	2028- 04-10	-69.386484	-2545.200436	2143.270452	-2569.745345	2155.596895	-10.811091				
2790 rd	2790 rows × 19 columns										



Display the underlying forecast dataframe (tail) forecast_ROKU[['ds', 'yhat', 'yhat_lower', 'yhat_upper']].tail()

	ds	yhat	yhat_lower	yhat_upper
2785	2028-04-06	-82.503884	-2570.760262	2136.060850
2786	2028-04-07	-82.304413	-2555.111452	2133.083051
2787	2028-04-08	-75.465950	-2546.477668	2137.460434
2788	2028-04-09	-74.946697	-2562.704950	2146.668549
2789	2028-04-10	-80.197575	-2545.200436	2143.270452

px.line(forecast_ROKU, x='ds',y='yhat')



figure = m.plot(forecast_ROKU, xlabel = 'ds', ylabel = 'y')

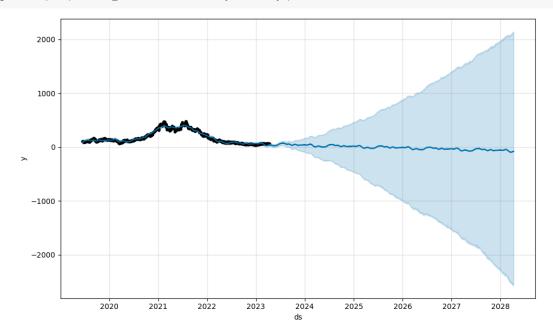
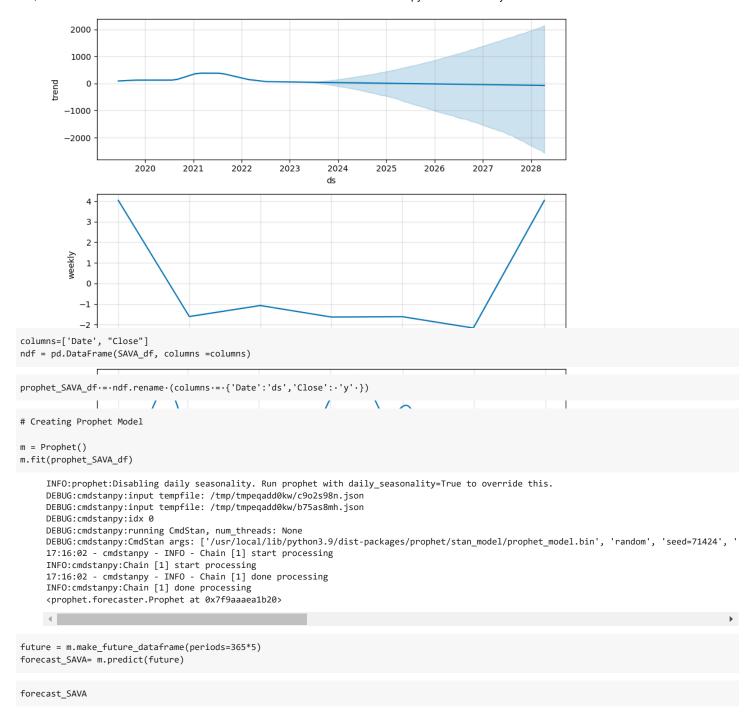


figure2=m.plot_components(forecast_ROKU)



		ds	trend	yhat_lower	yhat_upper	trend_lower	trend_upper	additive_terms	additive_term:	
	0	2019- 06-13	-0.313874	-10.751076	11.593996	-0.313874	-0.313874	0.464975	0	
	1	2019- 06-14	-0.298771	-11.619794	11.311607	-0.298771	-0.298771	0.322636	0	
	2	2019- 06-17	-0.253462	-12.274528	12.165609	-0.253462	-0.253462	0.438944	0	
	3	2019- 06-18	-0.238359	-10.917909	11.929222	-0.238359	-0.238359	0.744279	0	
		2010_								
<pre>px.line(forecast_SAVA, x='ds',y='yhat')</pre>										

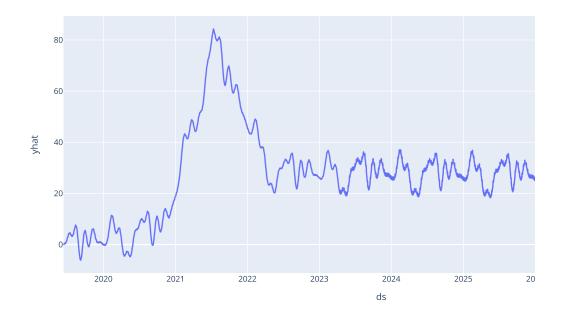
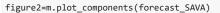
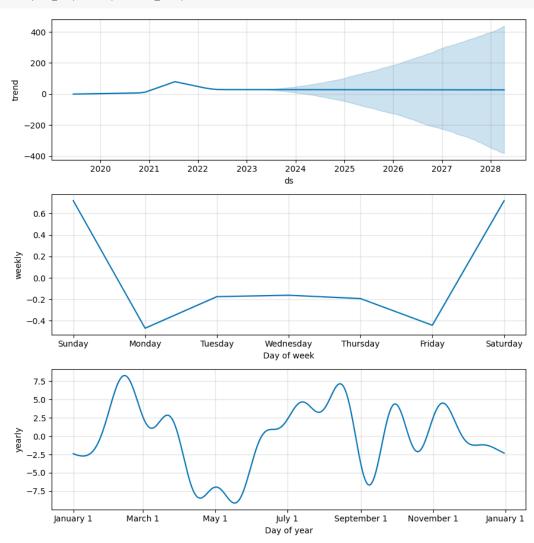


figure = m.plot(forecast_SAVA, xlabel = 'ds', ylabel = 'y')





```
columns=['Date', "Close"]
ndf = pd.DataFrame(QFIN_df, columns =columns)
prophet_QFIN_df = ndf.rename (columns = {'Date':'ds','Close': 'y' })
# Creating Prophet Model
m = Prophet()
m.fit(prophet_QFIN_df)
     INFO:prophet:Disabling daily seasonality. Run prophet with daily_seasonality=True to override this.
     DEBUG:cmdstanpy:input tempfile: /tmp/tmpeqadd0kw/3yc_pu9v.json
     DEBUG:cmdstanpy:input tempfile: /tmp/tmpeqadd0kw/wog29t52.json
     DEBUG:cmdstanpy:idx 0
     DEBUG::mdstanpy:running CmdStan, num_threads: None
DEBUG::mdstanpy:CmdStan args: ['/usr/local/lib/python3.9/dist-packages/prophet/stan_model/prophet_model.bin', 'random', 'seed=74700', '
     17:16:05 - cmdstanpy - INFO - Chain [1] start processing
     INFO:cmdstanpy:Chain [1] start processing 17:16:05 - cmdstanpy - INFO - Chain [1] done processing
     INFO:cmdstanpy:Chain [1] done processing
     cprophet.forecaster.Prophet at 0x7f9aaac2dc10>
     4
```

future = m.make_future_dataframe(periods=365*5)
forecast_QFIN= m.predict(future)

px.line(forecast_QFIN, x='ds',y='yhat')

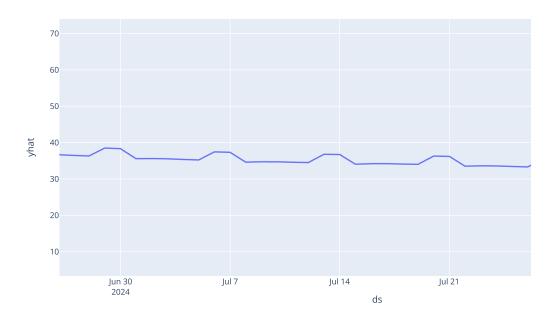


figure -- m.plot(forecast_QFIN, .xlabel -- 'ds', .ylabel -- 'y')

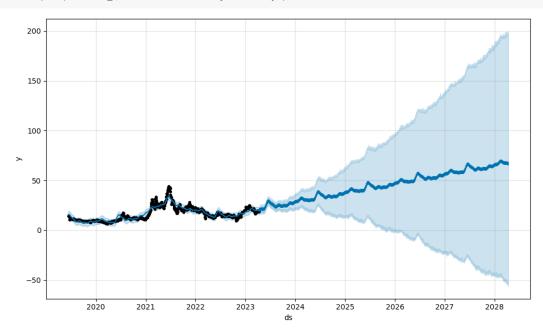
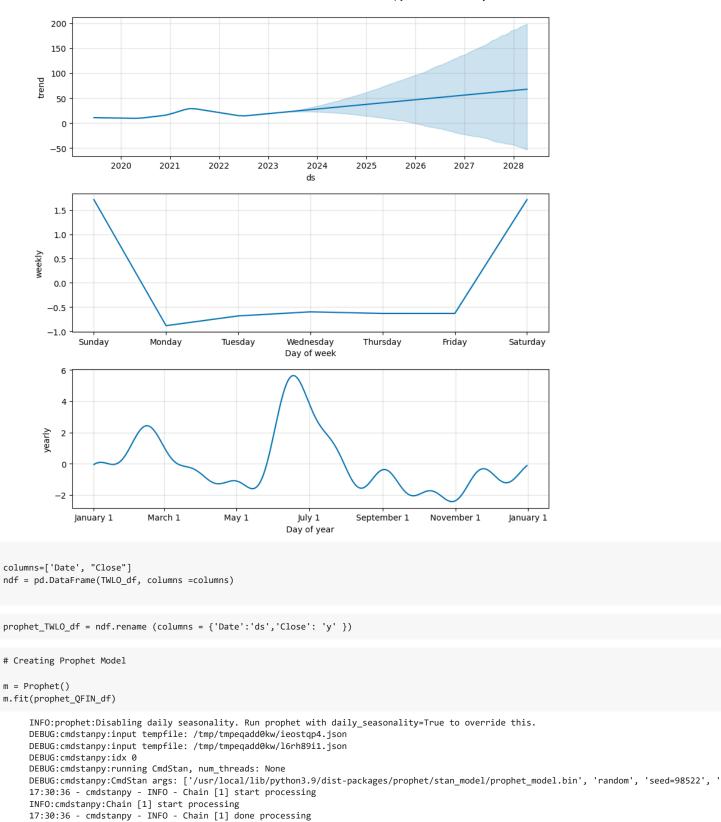


figure2=m.plot_components(forecast_QFIN)



```
future = m.make_future_dataframe(periods=365*5)
forecast_TWLO= m.predict(future)
```

 $px.line(forecast_TWLO, x='ds',y='yhat')$

INFO:cmdstanpy:Chain [1] done processing
<prephet.forecaster.Prophet at 0x7f9ab36a7d00>

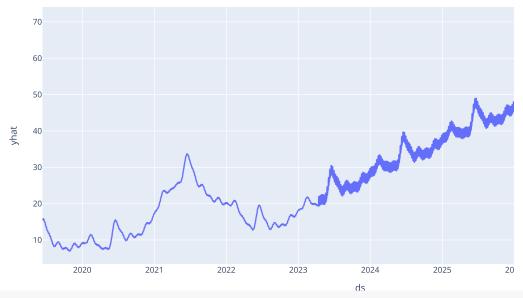


figure = m.plot(forecast_TWLO, xlabel = 'ds', ylabel = 'y')

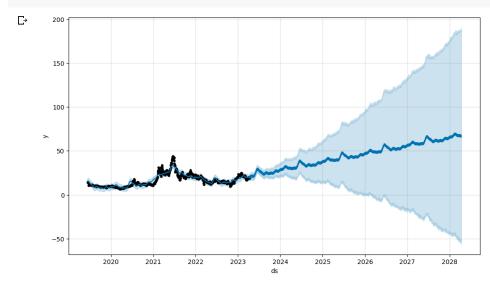
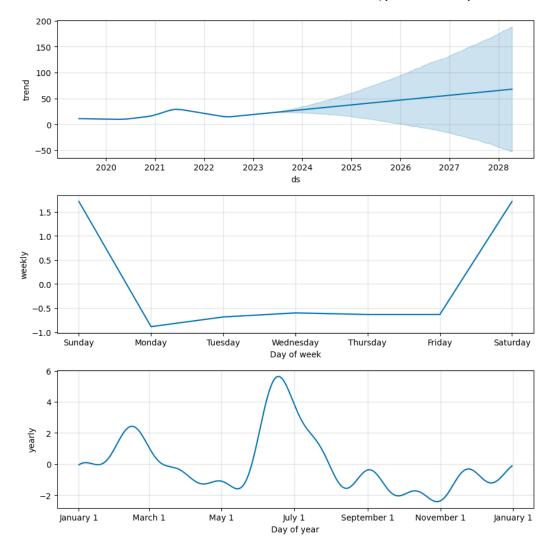


figure2=m.plot_components(forecast_TWLO)



✓ 1s completed at 1:34 PM