In [1]:

```
## Airline Passengers Forecasting Using FbProphet

import pandas as pd
import fbprophet
import matplotlib.pyplot as plt
%matplotlib inline
```

In [3]:

```
df=pd.read_csv('/content/airline_passengers.csv')
```

In [4]:

df.head()

Out[4]:

Month	Thousands of Passengers
0 1949-01	112.0
1 1949-02	118.0
2 1949-03	132.0
3 1949-04	129.0
4 1949-05	121.0

In [5]:

df.tail()

Out[5]:

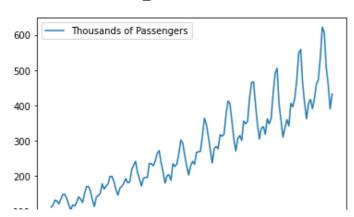
	Month	Thousands of Passengers
140	1960-09	508.0
141	1960-10	461.0
142	1960-11	390.0
143	1960-12	432.0
144 International airline passe	engers: monthly tota	NaN

In [7]:

df.plot()

Out[7]:

<matplotlib.axes. subplots.AxesSubplot at 0x7f05d9f32210>



```
100
                                       120
In [8]:
df.columns = ['ds','y']
df.head()
Out[8]:
       ds
0 1949-01 112.0
1 1949-02 118.0
2 1949-03 132.0
3 1949-04 129.0
4 1949-05 121.0
In [9]:
df.drop(144,axis=0,inplace=True)
In [10]:
df.tail()
Out[10]:
        ds
139 1960-08 606.0
140 1960-09 508.0
141 1960-10 461.0
142 1960-11 390.0
143 1960-12 432.0
In [11]:
df['ds'] = pd.to datetime(df['ds'])
In [12]:
df.head()
Out[12]:
         ds
0 1949-01-01 112.0
1 1949-02-01 118.0
2 1949-03-01 132.0
3 1949-04-01 129.0
4 1949-05-01 121.0
In [13]:
df.tail()
Out[13]:
           ds
139 1960-08-01 606.0
```

```
140 1960-09-01 508.0
141 1960-10-01 461.0
142 1960-11-01 390.0
143 1960-12-01 432.0
In [14]:
from fbprophet import Prophet
In [15]:
dir (Prophet)
Out[15]:
__dict__',
__dir__',
   __doc__',
__eq__',
  ___format__',
  __ge__',
  __getattribute__',
  __gt__',
  __hash__',
__init__',
   __init_subclass__',
   _le__',
_lt__',
   __module__',
  __ne__',
  new ',
 '_reduce ',
  reduce ex ',
 '__repr__',
 '__setattr__',
'__sizeof__',
 '__str__',
 "__subclasshook__',
"_weakref__',
"_load_stan_backend',
 'add_country_holidays',
 'add_group_component',
 'add_regressor',
 'add_seasonality',
 'construct holiday dataframe',
 'fit',
 'flat_growth_init',
 'flat trend',
 'fourier series',
 'initialize_scales',
 'linear growth init',
 'logistic_growth_init',
 'make_all_seasonality_features',
 'make_future_dataframe',
 'make_holiday_features',
 'make seasonality features',
 'parse_seasonality_args',
 'percentile',
 'piecewise_linear',
 'piecewise_logistic',
 'plot',
 'plot_components',
 'predict',
 'predict seasonal components',
 'predict trend',
 'predict uncertainty',
 'predictive_samples',
```

```
'sample_model',
 'sample posterior predictive',
 'sample_predictive_trend',
 'set_auto_seasonalities',
 'set changepoints',
 'setup_dataframe',
 'validate column name',
 'validate inputs']
In [16]:
# Initialize the Model
model=Prophet()
In [17]:
df.columns
Out[17]:
Index(['ds', 'y'], dtype='object')
In [18]:
df.dropna(axis=0,inplace=True)
INFO: numexpr.utils: NumExpr defaulting to 2 threads.
In [19]:
df.head()
Out[19]:
         ds
               У
0 1949-01-01 112.0
1 1949-02-01 118.0
2 1949-03-01 132.0
3 1949-04-01 129.0
4 1949-05-01 121.0
In [20]:
model.fit(df)
INFO: fbprophet: Disabling weekly seasonality. Run prophet with weekly seasonality=True to
override this.
INFO: fbprophet: Disabling daily seasonality. Run prophet with daily seasonality=True to ov
erride this.
Out[20]:
<fbprophet.forecaster.Prophet at 0x7f05d9a00410>
In [21]:
df.tail()
Out[21]:
          ds
139 1960-08-01 606.0
140 1960-09-01 508.0
141 1960-10-01 461.0
142 1960-11-01 390.0
```

'regressor column matrix',

143 1960-12-6 432.V

In [22]:

Create future dates of 365 days
future_dates=model.make_future_dataframe(periods=365)

In [23]:

future dates.tail()

Out[23]:

ds

504 1961-11-27

505 1961-11-28

506 1961-11-29

507 1961-11-30

508 1961-12-01

In [24]:

prediction=model.predict(future_dates)

In [25]:

prediction.head()

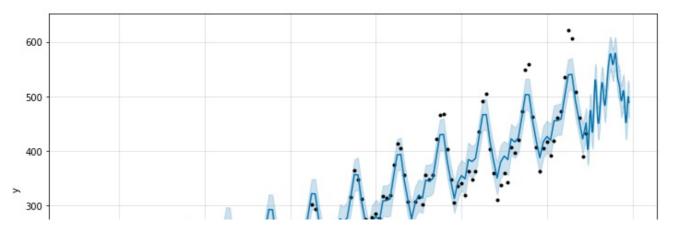
Out[25]:

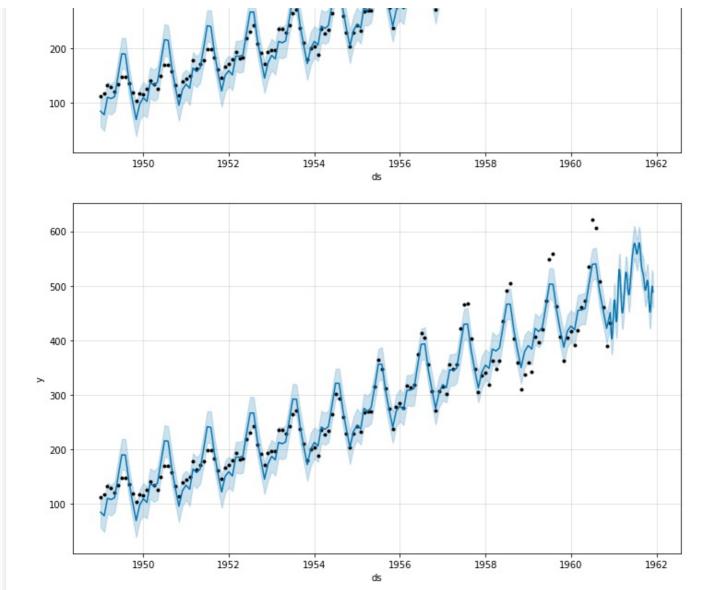
	ds	trend	yhat_lower	yhat_upper	trend_lower	trend_upper	additive_terms	additive_terms_lower	additive_terms_u
0	1949- 01-01	106.727984	56.578938	110.369495	106.727984	106.727984	-21.920631	-21.920631	-21.92
1	1949- 02-01	108.901201	48.320578	109.039991	108.901201	108.901201	-30.682847	-30.682847	-30.68
2	1949- 03-01	110.864107	81.549103	138.690395	110.864107	110.864107	-0.502710	-0.502710	-0.50
3	1949- 04-01	113.037323	78.395219	134.615024	113.037323	113.037323	-5.185501	-5.185501	-5.18
4	1949- 05-01	115.140437	84.207209	141.213038	115.140437	115.140437	-3.788702	-3.788702	-3.78
4									Þ

In [26]:

plot the predicted projection
model.plot(prediction)

Out[26]:

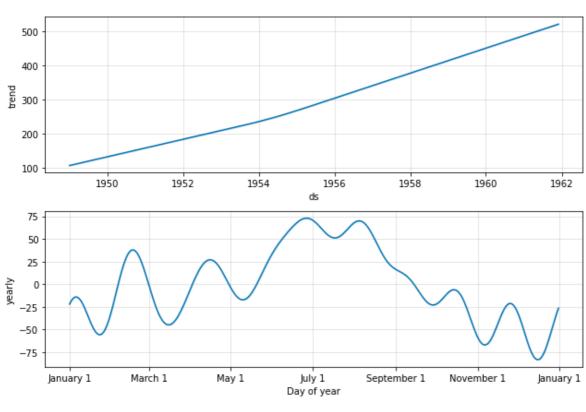


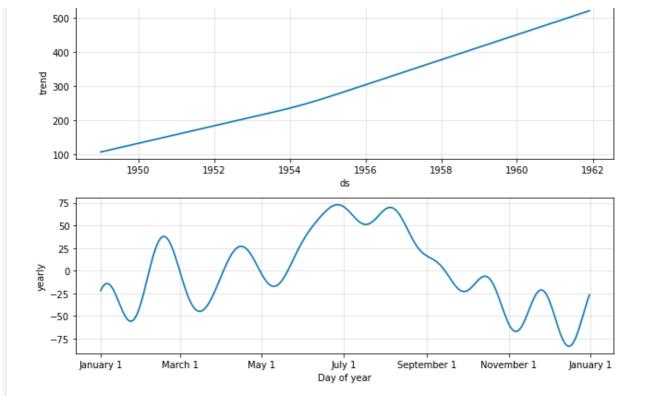


In [29]:

Visualize Each Components[Trends, yearly]
model.plot_components(prediction)

Out[29]:





In [30]:

df.head()

Out[30]:

	ds	У
0	1949-01-01	112.0
1	1949-02-01	118.0
2	1949-03-01	132.0
3	1949-04-01	129.0
4	1949-05-01	121.0

In [31]:

```
from fbprophet.diagnostics import cross_validation
df_cv = cross_validation(model, initial='730 days', period='180 days', horizon = '365 da
ys')
df_cv.head()
```

INFO:fbprophet:Making 19 forecasts with cutoffs between 1951-01-18 00:00:00 and 1959-12-0 $2\ 00:00:00$

```
INFO:fbprophet:n_changepoints greater than number of observations. Using 19. INFO:fbprophet:n_changepoints greater than number of observations. Using 23.
```

Out[31]:

	ds	yhat	yhat_lower	yhat_upper	у	cutoff
0	1951-02-01	158.013969	157.798379	158.230357	150.0	1951-01-18
1	1951-03-01	171.224502	170.530372	171.935670	178.0	1951-01-18
2	1951-04-01	159.628157	158.215258	161.056428	163.0	1951-01-18
3	1951-05-01	145.666500	143.483629	147.909767	172.0	1951-01-18
4	1951-06-01	180.679910	177.532311	183.900682	178.0	1951-01-18

In [32]:

from fbprophet.diagnostics import performance_metrics

```
df_p = performance_metrics(df_cv)
df_p.head()
```

Out[32]:

	horizon	mse	rmse	mae	mape	mdape	coverage
0	40 days	1083.476338	32.916202	25.895616	0.078758	0.073725	0.227273
1	42 days	1009.201759	31.767936	25.246674	0.077080	0.073725	0.181818
2	46 days	906.510734	30.108317	24.749820	0.079413	0.084057	0.136364
3	47 days	1039.076713	32.234713	26.867165	0.082299	0.085796	0.136364
4	48 days	1027.847500	32.060061	26.277771	0.080195	0.085796	0.181818

In [33]:

```
df_p.head()
```

Out[33]:

	horizon	mse	rmse	mae	mape	mdape	coverage
0	40 days	1083.476338	32.916202	25.895616	0.078758	0.073725	0.227273
1	42 days	1009.201759	31.767936	25.246674	0.077080	0.073725	0.181818
2	46 days	906.510734	30.108317	24.749820	0.079413	0.084057	0.136364
3	47 days	1039.076713	32.234713	26.867165	0.082299	0.085796	0.136364
4	48 days	1027.847500	32.060061	26.277771	0.080195	0.085796	0.181818

In [34]:

```
from fbprophet.plot import plot_cross_validation_metric
fig = plot_cross_validation_metric(df_cv, metric='rmse')
```

