```
import pandas as pd
In [1]:
         import numpy as np
         import matplotlib.pyplot as plt
         %matplotlib inline
         airline = pd.read_csv('airline_passengers.csv',index_col="Month")
In [3]:
        airline.head()
In [4]:
Out[4]:
                 Thousands of Passengers
          Month
         1949-01
                                112.0
         1949-02
                                 118.0
```

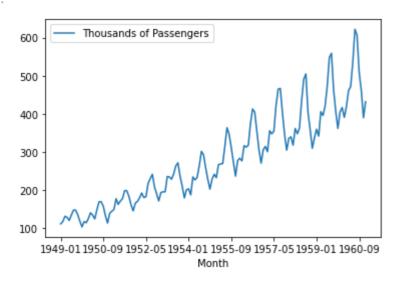
In [5]: airline.plot()

1949-03

1949-04

1949-05

Out[5]: <AxesSubplot:xlabel='Month'>



132.0

129.0

121.0

ETS

We can use an additive model when it seems that the trend is more linear and the seasonality and trend components seem to be constant over time (e.g. every year we add 10,000 passengers). A multiplicative model is more appropriate when we are increasing (or decreasing) at a non-linear rate (e.g. each year we double the amount of passengers).

Based off this chart, it looks like the trend in these earlier days is slightly increasing at a higher rate than just linear (although it is a bit hard to tell from this one plot).

```
In [51]: # Get data in correct format
airline.dropna(inplace=True)
airline.index = pd.to_datetime(airline.index)
```

In [52]: airline.head()

Out[52]: Thousands of Passengers

Month	
1949-01-01	112.0
1949-02-01	118.0
1949-03-01	132.0
1949-04-01	129.0
1949-05-01	121.0

```
In [58]: from statsmodels.tsa.seasonal import seasonal_decompose
  result = seasonal_decompose(airline['Thousands of Passengers'], model='multiplicative')
  result.plot()
```

```
Out[58]:
                Observed
                   500
                    250
                Trend
                    400
                   200
                    1.2
                 Seasonal
                    1.0
                    0.8
                    1.1
                 Residual
                    1.0
                    0.9
                                     1951
                                                    1953
                                                                  1955
                                                                                 1957
                                                                                                1959
                       1949
                                                                  Month
                Observed
                    500
                    250
                Trend
                    400
                   200
                    1.2
                 Seasonal
                    1.0
                    0.8
                    1.1
                 Residual
                    1.0
                    0.9
                                     1951
                                                    1953
                                                                  1955
                                                                                 1957
                                                                                                1959
                       1949
                                                                  Month
```