```
In [1]:
import warnings
warnings.filterwarnings("ignore")
In [2]:
import numpy as np
import pandas as pd
In [3]:
df = pd.read_csv('/content/Airline_Passangers.csv')
In [4]:
df.isnull().sum()
Out[4]:
Month
               0
Passengers
dtype: int64
In [5]:
df.shape
Out[5]:
(144, 2)
Building ARMA(p,q) model:
 • order = (1,0,1)
In [6]:
order = (1,0,1)
In [7]:
from statsmodels.tsa.arima model import ARIMA
In [10]:
model = ARIMA(df.Passengers[:50],order=order)
In [11]:
ARMA model = model.fit()
In [12]:
ARMA_model.summary2()
Out[12]:
          Model:
                         ARMA
                                          BIC: 428.5331
Dependent Variable:
                     Passengers
                                  Log-Likelihood:
                                                -206.44
                     2022-08-29
           Date:
                                         Scale:
                                                 1.0000
                          18:56
  No. Observations:
                                       Method:
                            50
                                               css-mle
```

0

Sample:

3

47

Df Model:

D4 D = = : d . . = ! = .

```
S.D. of
                          1.0000
                                                   14.770
       Converged:
                                      innovations:
     No. Iterations:
                         17.0000
                                           HQIC: 423.797
             AIC:
                        420.8850
                   Coef. Std.Err.
                                      t
                                          P>ltl
                                                 [0.025
                                                         0.975]
          const 158.8138 15.1420 10.4883 0.0000 129.1361 188.4916
 ar.L1.Passengers
                  0.8397
                          0.0843
                                 9.9666 0.0000
                                                0.6746
                                                         1.0048
ma.L1.Passengers
                  0.2816
                         0.1407
                                 2.0013 0.0454
                                                0.0058
                                                         0.5574
        Real Imaginary Modulus Frequency
 AR.1 1.1909
                0.0000
                        1.1909
                                  0.0000
MA.1 -3.5510
                0.0000
                        3.5510
                                  0.5000
In [14]:
def get mape(actual, pred):
  mape = np.round(np.mean(np.abs(100*(actual-pred)/actual)),2)
  return mape
In [16]:
from sklearn.metrics import mean squared error
def get rmse(actual, pred):
  rmse = np.round(np.sqrt(mean squared error(actual, pred)),2)
  return rmse
In [18]:
ARMA actual = df.Passengers[50:60]
ARMA forecast = ARMA model.predict(50,60)
In [21]:
df ARMA = pd.DataFrame({'actual':ARMA actual, 'pred':np.round(ARMA forecast,0)})
df_ARMA
Out[21]:
    actual pred
50
   236.0 192.0
51
    235.0 186.0
52 229.0 182.0
    243.0 178.0
53
    264.0 175.0
54
    272.0 173.0
55
    237.0 170.0
56
57
    211.0 169.0
58
     180.0 167.0
59
    201.0 166.0
60
     NaN 165.0
In [24]:
ARMA mape = get mape(df ARMA.actual[:9], df ARMA.pred[:9])
```

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DI Residuais:

ARMA mape

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Out[24]:
23.59
In [26]:
ARMA_rmse = get_rmse(df_ARMA.actual[:9], df_ARMA.pred[:9])
ARMA_rmse
Out[26]:
62.29
In []:
```