Business & Data Understaning

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import pandas as pd
df=pd.read_csv("/content/AirPassengers.csv")
df.head()

	Month	#Passengers	
0	1949-01	112	ılı
1	1949-02	118	
2	1949-03	132	
3	1949-04	129	
4	1949-05	121	

df.tail()



df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 144 entries, 0 to 143
Data columns (total 2 columns):
Column Non-Null Count Dtype
--- 0 Month 144 non-null object
1 #Passengers 144 non-null int64
dtypes: int64(1), object(1)
memory usage: 2.4+ KB

This dataset include history of 12 years from 1949-01 to 1960-12

```
import matplotlib.pyplot as plt
plt.xlabel('Months')
plt.ylabel('Number of Passengers')
plt.title('Distribution of the number of passengers over 12 years')
plt.plot(df['#Passengers'])
plt.show()
```

Distribution of the number of passengers over 12 years

Data Preparation

```
-
                                                                                      - 11
                                                                                             - 1
def load_data(data, seq_len):
  X = []
   y = []
  for i in range(seq_len, len(data)):
     X.append(data.iloc[i-seq_len : i, 1])
     y.append(data.iloc[i,1])
   return X,y
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                                                        IN I M''
X,y=load_data(df,20)
                                        -
                                                                                               I
len(y)
      124
           100 ]
                                                                                               1
X_train=X[:100]
y_train=y[:100]
X_test=X[100:]
y_test=y[100:]
import numpy as np
X_train=np.array(X_train)
y_train=np.array(y_train)
X_test=np.array(X_test)
y_test=np.array(y_test)
X_train
      [301, 356, 348, ..., 491, 505, 404],
[356, 348, 355, ..., 505, 404, 359],
                [348, 355, 422, ..., 404, 359, 310]])
print('x_train.shape = ',X_train.shape)
print('y_train.shape = ', y_train.shape)
print('x_test.shape = ', X_test.shape)
print('y_test.shape = ', y_test.shape)
      x_train.shape = (100, 20)
y_train.shape = (100,)
      x_{\text{test.shape}} = (24, 20)
      y_{\text{test.shape}} = (24,)
X_{train} = np.reshape(X_{train}, (100, 20, 1))
X_{\text{test}} = \text{np.reshape}(X_{\text{test}}, (24, 20, 1))
print('x_train.shape = ',X_train.shape)
print( x_train.shape = ', x_train.shape)
print('y_train.shape = ', y_train.shape)
print('x_test.shape = ', X_test.shape)
print('y_test.shape = ', y_test.shape)
      x_{train.shape} = (100, 20, 1)
      y_train.shape = (100,)
x_test.shape = (24, 20, 1)
y_test.shape = (24,)
```

Neural Network from Keras

```
from keras.models import Sequential
from keras.layers import Dense

model = Sequential()
model.add(Dense(40, input_shape=(20,),activation='relu'))
model.add(Dense(40,activation='relu'))
model.add(Dense(1))
```

```
model.compile(loss='mean_squared_error', optimizer='adam')
model.fit(X_train, y_train, epochs=500, batch_size=2, verbose=2)
    50/50 - 0s - loss: 53.7200 - 104ms/epoch - 2ms/step
    Epoch 473/500
    50/50 - 0s - loss: 60.3407 - 108ms/epoch - 2ms/step
    Epoch 474/500
    50/50 - 0s - loss: 56.6721 - 123ms/epoch - 2ms/step
    Epoch 475/500
    50/50 - 0s - loss: 89.6014 - 117ms/epoch - 2ms/step
    Epoch 476/500
    50/50 - 0s - loss: 59.4444 - 126ms/epoch - 3ms/step
    Epoch 477/500
    50/50 - 0s - loss: 62.7189 - 119ms/epoch - 2ms/step
    Epoch 478/500
    50/50 - 0s - loss: 61.8451 - 103ms/epoch - 2ms/step
    Epoch 479/500
    50/50 - 0s - loss: 66.5758 - 105ms/epoch - 2ms/step
    Epoch 480/500
    50/50 - 0s - loss: 57.3906 - 109ms/epoch - 2ms/step
    Epoch 481/500
    50/50 - 0s - loss: 83.2053 - 106ms/epoch - 2ms/step
    Epoch 482/500
    50/50 - 0s - loss: 114.6352 - 95ms/epoch - 2ms/step
    Epoch 483/500
    50/50 - 0s - loss: 96.4560 - 124ms/epoch - 2ms/step
    Epoch 484/500
    50/50 - 0s - loss: 117.0442 - 113ms/epoch - 2ms/step
    Epoch 485/500
    50/50 - 0s - loss: 82.1807 - 114ms/epoch - 2ms/step
    Epoch 486/500
    50/50 - 0s - loss: 86.4101 - 117ms/epoch - 2ms/step
    Epoch 487/500
    50/50 - 0s - loss: 74.5774 - 118ms/epoch - 2ms/step
    Epoch 488/500
    50/50 - 0s - loss: 63.9437 - 105ms/epoch - 2ms/step
    Epoch 489/500
    50/50 - 0s - loss: 86.3503 - 113ms/epoch - 2ms/step
    Epoch 490/500
    50/50 - 0s - loss: 139.2220 - 113ms/epoch - 2ms/step
    Epoch 491/500
    50/50 - 0s - loss: 296.7442 - 115ms/epoch - 2ms/step
    Epoch 492/500
    50/50 - 0s - loss: 140.0320 - 107ms/epoch - 2ms/step
    Epoch 493/500
    50/50 - 0s - loss: 156.4481 - 108ms/epoch - 2ms/step
    Epoch 494/500
    50/50 - 0s - loss: 98.1925 - 110ms/epoch - 2ms/step
    Epoch 495/500
    50/50 - 0s - loss: 88.9676 - 112ms/epoch - 2ms/step
    Epoch 496/500
    50/50 - 0s - loss: 80.3688 - 111ms/epoch - 2ms/step
    Epoch 497/500
    50/50 - 0s - loss: 131.3052 - 106ms/epoch - 2ms/step
    Epoch 498/500
    50/50 - 0s - loss: 89.6818 - 113ms/epoch - 2ms/step
    Epoch 499/500
    50/50 - 0s - loss: 95.3477 - 113ms/epoch - 2ms/step
    Epoch 500/500
    50/50 - 0s - loss: 70.0680 - 116ms/epoch - 2ms/step
    <keras.src.callbacks.History at 0x7b88c1a7d660>
y_pred=model.predict(X_test)
    1/1 [======] - 0s 89ms/step
y_pred
    array([[351.3825],
            [334.73746],
            [374.88666],
            [375.75555],
            [396.2213],
           [489.9771],
            [540.4173],
            [549.36145],
            [450.62213],
            [389.89178],
            [336.34064],
            [383.15656],
            [413.22577],
            [401.8603 ].
            [457.02637],
            [422.505
            [457.09988],
           [538.73566],
            [611.03424],
            [619.6796],
```

```
[509.58765],
[442.91708],
[392.5771],
[432.24]], dtype=float32)

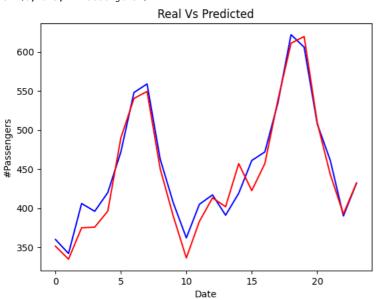
y_test

array([360, 342, 406, 396, 420, 472, 548, 559, 463, 407, 362, 405, 417,
391, 419, 461, 472, 535, 622, 606, 508, 461, 390, 432])

plt.plot(y_test, color='blue', label='Real')
plt.plot(y_pred, color='red', label='Predicted')
plt.title('Real Vs Predicted')
plt.xlabel('Date')
```

Text(0, 0.5, '#Passengers')

plt.ylabel('#Passengers')



```
from sklearn.metrics import mean_squared_error
rmse = np.sqrt(mean_squared_error(y_test, y_pred)).round(2)
mae = np.round(np.mean(np.abs(y_test-y_pred))*100,2)
mape = np.round(np.mean(np.abs(y_test-y_pred)/y_test)*100,2)

print("RMSE=",rmse)
print("MAE=",mae)
print("MAPE=",mape)

RMSE= 18.33
    MAE= 8624.88
    MAPE= 18.89
```