Experiment 12

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
import warnings
warnings.filterwarnings('ignore')
```

df = pd.read_csv('College_Data.csv')

df.head(3)

₹		Unnamed:	Private	Apps	Accept	Enroll	Top10perc	Top25perc	F.Undergrad	P.Undergrad	Outstate	Room.Board
	0	Abilene Christian University	Yes	1660	1232	721	23	52	2885	537	7440	3300
	1	Adelphi University	Yes	2186	1924	512	16	29	2683	1227	12280	6450
	2	Adrian College	Yes	1428	1097	336	22	50	1036	99	11250	3750

df.info()

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#	Column	Non-Null Count	t Dtype						
0	Unnamed: 0	777 non-null	object						
1	Private	777 non-null	object						
2	Apps	777 non-null	int64						
3	Accept	777 non-null	int64						
4	Enroll	777 non-null	int64						
5	Top10perc	777 non-null	int64						
6	Top25perc	777 non-null	int64						
7	F.Undergrad	777 non-null	int64						
8	P.Undergrad	777 non-null	int64						
9	Outstate	777 non-null	int64						
10	Room.Board	777 non-null	int64						
11	Books	777 non-null	int64						
12	Personal	777 non-null	int64						
1 3	PhD	777 non-null	int64						
14	Terminal	777 non-null	int64						
15	S.F.Ratio	777 non-null	float64						
16	perc.alumn <u>i</u>	777 non-null	int64						
17	Expend	777 non-null	int64						
18	Grad.Rate	777 non-null	int64						
<pre>dtypes: float64(1), int64(16), object(2)</pre>									
memory usage: 115.5+ KB									

df.isnull().sum()

```
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```

0

Unnamed: 0 0

Private 0

Apps 0

Accept 0

Enroll 0

Top10perc 0

Top25perc 0

F.Undergrad 0

P.Undergrad 0

Outstate 0

Room.Board 0

Books 0

Personal 0

PhD 0

Terminal 0

S.F.Ratio 0

perc.alumni 0

Expend 0

Grad.Rate 0

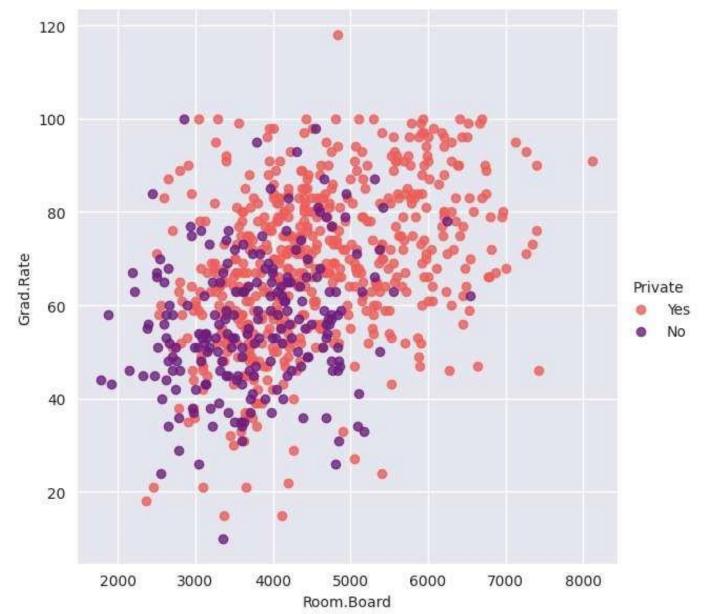
dtype: int64

df.describe()

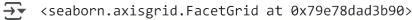
₹		Apps	Accept	Enroll	Top10perc	Top25perc	F.Undergrad	P.Undergrad	Outstate	Rc
	count	777.000000	777.000000	777.000000	777.000000	777.000000	777.000000	777.000000	777.000000	7
	mean	3001.638353	2018.804376	779.972973	27.558559	55.796654	3699.907336	855.298584	10440.669241	43
	std	3870.201484	2451.113971	929.176190	17.640364	19.8 0 4778	4850.420531	1522.431887	4023.016484	10!
	min	81.000000	72.000000	35.000000	1.000000	9.000000	139.000000	1.000000	2340.000000	17≀
	25%	776.000000	604.000000	242.000000	15.000000	41.000000	992.000000	95.000000	7320.000000	35!
	50%	1558.000000	1110.000000	434.000000	23.000000	54.000000	1707.000000	353.000000	9990.000000	420
	75%	3624.000000	2424.000000	902.000000	35.000000	69.000000	4005.000000	967.000000	12925.000000	50!
	max	48094.000000	26330.000000	6392.000000	96.000000	100.000000	31643.000000	21836.000000	21700.000000	81:

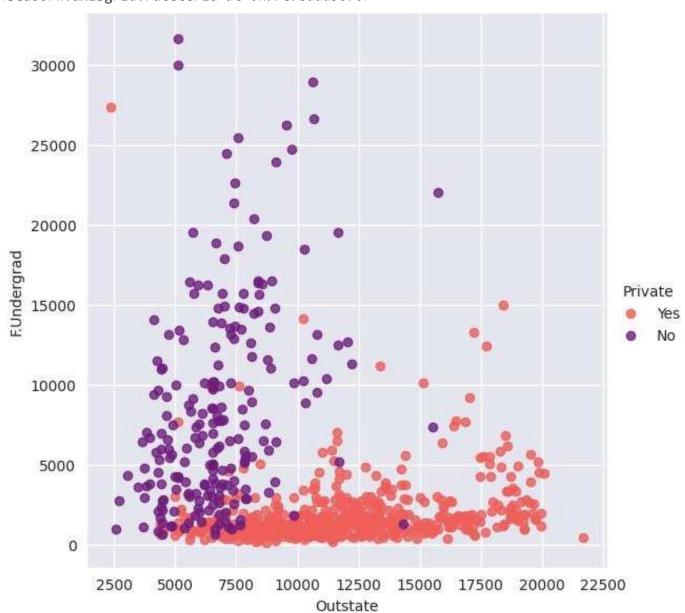
2/5

<seaborn.axisgrid.FacetGrid at 0x79e78de4fe10>



sns.set_style('darkgrid') sns.lmplot(x='Outstate', y='F.Undergrad', data=df, hue='Private', palette='magma_r', height=6, aspect=1, fit_reg=Fals

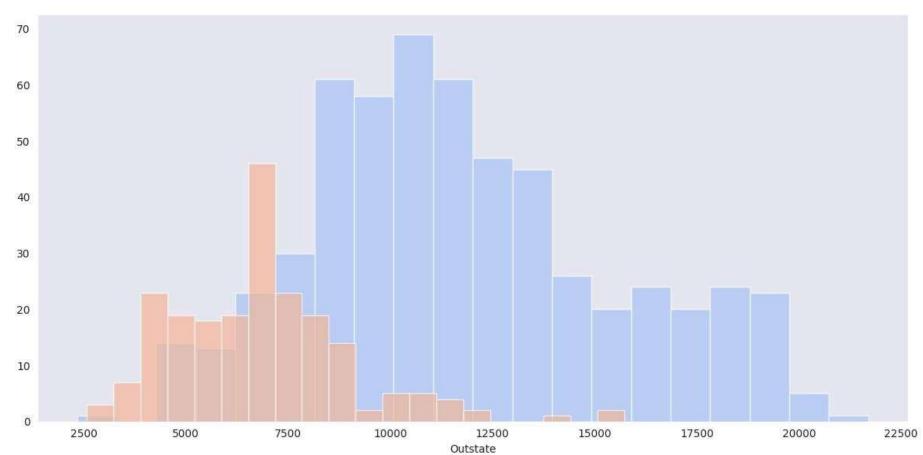




sns.set_style('dark') # Replaced 'size' with 'height' as 'size' is deprecated in newer seaborn versions. h = sns.FacetGrid(df,hue="Private",palette='coolwarm',height=6,aspect=2) h = h.map(plt.hist, 'Outstate', bins=20, alpha=0.7)

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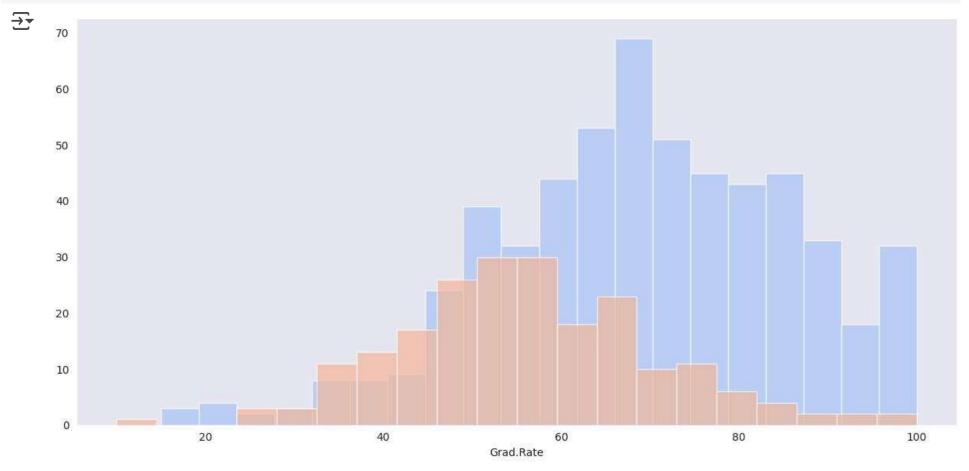


df.set_value(95, 'Grad.Rate', 100) # Deprecated method
df.at[95, 'Grad.Rate'] = 100 # Use .at accessor for label-based indexing
df[df['Grad.Rate'] > 100]

→

Unnamed:
OPrivate Apps Accept Enroll Top10perc Top25perc F.Undergrad P.Undergrad Outstate Room.Board B

Replaced 'size' with 'height' as 'size' is deprecated in newer seaborn versions.
g = sns.FacetGrid(df,hue="Private",palette='coolwarm',height=6,aspect=2)
g = g.map(plt.hist,'Grad.Rate',bins=20,alpha=0.7)



```
Index(['Unnamed: 0', 'Private', 'Apps', 'Accept', 'Enroll', 'Top10perc',
        'Top25perc', 'F.Undergrad', 'P.Undergrad', 'Outstate', 'Room.Board',
        'Books', 'Personal', 'PhD', 'Terminal', 'S.F.Ratio', 'perc.alumni',
        'Expend', 'Grad.Rate'],
       dtype='object')
from sklearn.cluster import KMeans
kmeans = KMeans(n_clusters=2)
kmeans.fit(df.drop(['Private', 'Unnamed: 0'], axis=1))
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                     (i) (?)
           KMeans
     KMeans(n_clusters=2)
means=kmeans.cluster_centers_
print(means)
→ [[1.81323468e+03 1.28716592e+03
                                      4.91044843e+02 2.53094170e+01
       5.34708520e+01
       4.31136472e+03 3:48984888586+03 1:280388346+03 1:0342549846+04
       7.78251121e+01 1.40997010e+01 2.31748879e+01 8.93204634e+03
       6.50926756e+01]
      [1.03631389e+04 6.55089815e+03 2.56972222e+03
                                                     4.14907407e+01
       7.02037037e+01 1.30619352e+04
       4.64347222e+03 5.95212963e+02 1:46486376e+03 1:83981489e+04
       9.13333333e+01 1.40277778e+01 2.00740741e+01 1.41705000e+04
       6.75925926e+01]]
def converter(cluster):
    if cluster=='Yes':
        return 1
    else:
        return 0
df['Cluster'] = df['Private'].apply(converter)
df.head(3)
```

→ ▼	Unnamed:										
		Private	Apps	Accept	Enroll	Top10perc	Top25perc	F.Undergrad	P.Undergrad	Outstate	Room.Board

	Unnamed: 0	Private	Apps	Accept	Enroll	Top10perc	Top25perc	F.Undergrad	P.Undergrad	Outstate	Room.Board	
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2	Adrian	Yes	1428	1097	336	22	50	1036	99	11250	3750	

Conclusion:-

In this experiment, we understand how to evaluate the performance of the K-Means Clustering algorithm using a confusion matrix. It demonstrated how clustering results can be compared with actual labels to assess accuracy and cluster quality. In this experiment, the theoretical approach and the practical approach are the same.