

In [1]:

```
#For this project we will attempt to use KMeans Clustering to cluster Universities into  
to two groups, Private and Public.  
#Just note We already have labels for the data here, which in the real world would not  
happen while using K-means clustering as it is  
#Unsupervised Learning algorithm.  
  
# This is basically a test for how K means fares as a unsupervised Learning algo.  
#We will use a data frame with 777 observations on the following 18 variables.  
  
#Private A factor with levels No and Yes indicating private or public university  
#Apps Number of applications received  
#Accept Number of applications accepted  
#Enroll Number of new students enrolled  
#Top10perc Pct. new students from top 10% of H.S. class  
#Top25perc Pct. new students from top 25% of H.S. class  
#F.Undergrad Number of fulltime undergraduates  
#P.Undergrad Number of parttime undergraduates  
#Outstate Out-of-state tuition  
#Room.Board Room and board costs  
#Books Estimated book costs  
#Personal Estimated personal spending  
#PhD Pct. of faculty with Ph.D.'s  
#Terminal Pct. of faculty with terminal degree  
#S.F.Ratio Student/faculty ratio  
#perc.alumni Pct. alumni who donate  
#Expend Instructional expenditure per student  
#Grad.Rate Graduation rate
```

In [2]:

```
import numpy as np  
import pandas as pd  
import seaborn as sns  
import matplotlib.pyplot as plt  
%matplotlib inline
```

In [7]:

```
univ_df=pd.read_csv('College_Data',index_col=0)
```

In [8]:

```
univ_df.head()
```

Out[8]:

	Private	Apps	Accept	Enroll	Top10perc	Top25perc	F.Undergrad	P.Undergrad
Abilene Christian University	Yes	1660	1232	721	23	52	2885	537
Adelphi University	Yes	2186	1924	512	16	29	2683	1227
Adrian College	Yes	1428	1097	336	22	50	1036	99
Agnes Scott College	Yes	417	349	137	60	89	510	63
Alaska Pacific University	Yes	193	146	55	16	44	249	869

In [9]:

```
univ_df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
Index: 777 entries, Abilene Christian University to York College of Pennsylvania
```

```
Data columns (total 18 columns):
```

#	Column	Non-Null Count	Dtype
0	Private	777 non-null	object
1	Apps	777 non-null	int64
2	Accept	777 non-null	int64
3	Enroll	777 non-null	int64
4	Top10perc	777 non-null	int64
5	Top25perc	777 non-null	int64
6	F.Undergrad	777 non-null	int64
7	P.Undergrad	777 non-null	int64
8	Outstate	777 non-null	int64
9	Room.Board	777 non-null	int64
10	Books	777 non-null	int64
11	Personal	777 non-null	int64
12	PhD	777 non-null	int64
13	Terminal	777 non-null	int64
14	S.F.Ratio	777 non-null	float64
15	perc.alumni	777 non-null	int64
16	Expend	777 non-null	int64
17	Grad.Rate	777 non-null	int64

```
dtypes: float64(1), int64(16), object(1)
```

```
memory usage: 115.3+ KB
```

In [11]:

```
univ_df.describe()
```

Out[11]:

	Apps	Accept	Enroll	Top10perc	Top25perc	F.Undergrad	P.Priv
count	777.000000	777.000000	777.000000	777.000000	777.000000	777.000000	777.000000
mean	3001.638353	2018.804376	779.972973	27.558559	55.796654	3699.907336	85.411648
std	3870.201484	2451.113971	929.176190	17.640364	19.804778	4850.420531	152.801100
min	81.000000	72.000000	35.000000	1.000000	9.000000	139.000000	0.000000
25%	776.000000	604.000000	242.000000	15.000000	41.000000	992.000000	9.000000
50%	1558.000000	1110.000000	434.000000	23.000000	54.000000	1707.000000	35.000000
75%	3624.000000	2424.000000	902.000000	35.000000	69.000000	4005.000000	96.000000
max	48094.000000	26330.000000	6392.000000	96.000000	100.000000	31643.000000	2183.000000

In [12]:

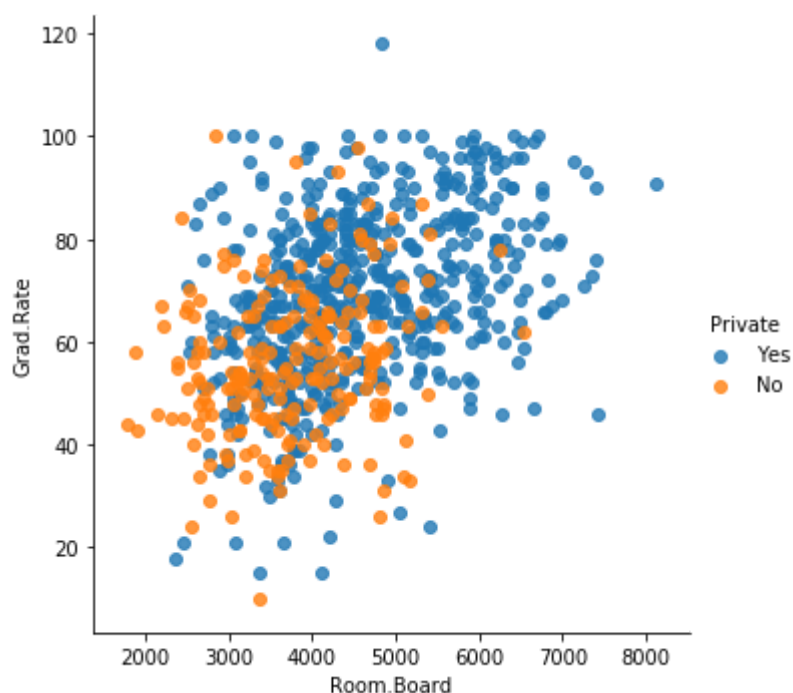
```
#Lets do some exploratory data analysis visualitions
```

In [14]:

```
#Scatter Plot with a hue of Private : Yes or No  
#Just used a lmpot plot and got rid of fit_reg to plot something simi;ar as scatter pl  
ot of matplotlib  
sns.lmplot(x='Room.Board',y='Grad.Rate',data=univ_df,hue='Private',fit_reg=False)
```

Out[14]:

<seaborn.axisgrid.FacetGrid at 0x2a45837dbc8>



In [16]:

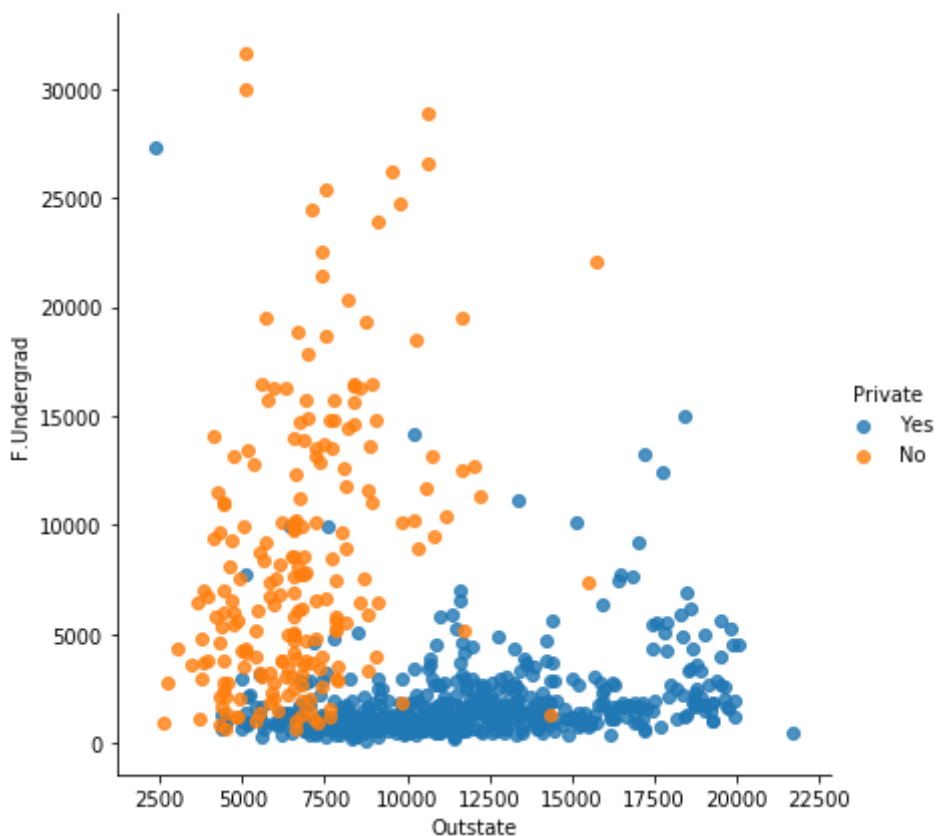
```
#A scatterplot of F.Undergrad versus Outstate where the points are colored by the Private column.  
sns.lmplot(x='Outstate',y='F.Undergrad',data=univ_df,hue='Private',fit_reg=False,size=6,aspect=1)
```

C:\Users\anike\anaconda3\lib\site-packages\seaborn\regression.py:574: UserWarning: The `size` parameter has been renamed to `height`; please update your code.

```
warnings.warn(msg, UserWarning)
```

Out[16]:

<seaborn.axisgrid.FacetGrid at 0x2a458c007c8>



In [17]:

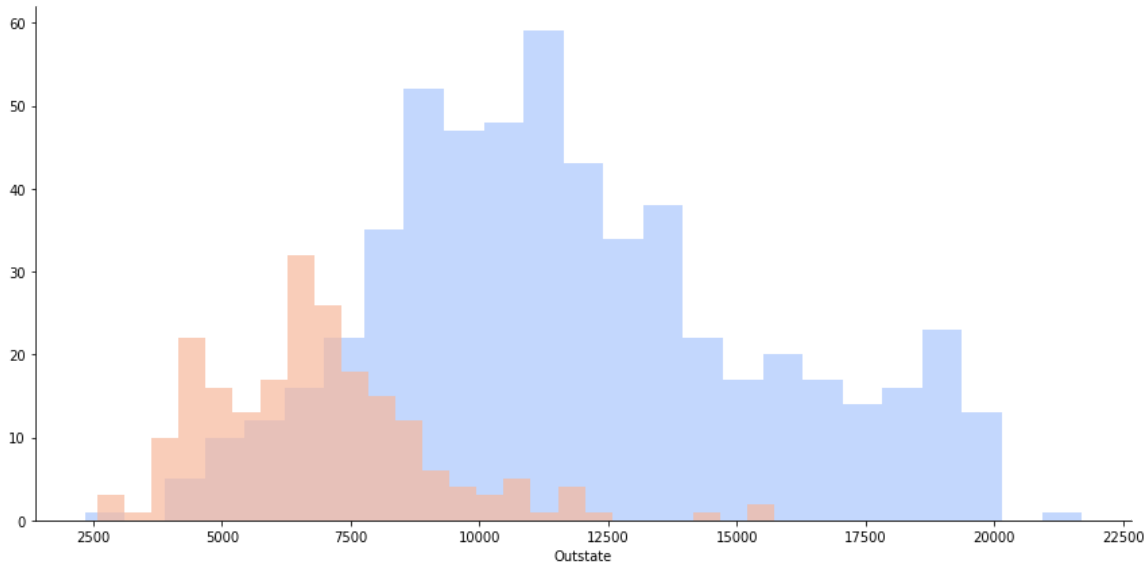
```
#We can already tell that tuition for private schools is way higher
```

In [19]:

```
#A stacked histogram showing Out of State Tuition based on the Private column.  
g=sns.FacetGrid(univ_df,hue='Private',palette='coolwarm',size=6,aspect=2)  
g=g.map(plt.hist,'Outstate',bins=25,alpha=0.7)
```

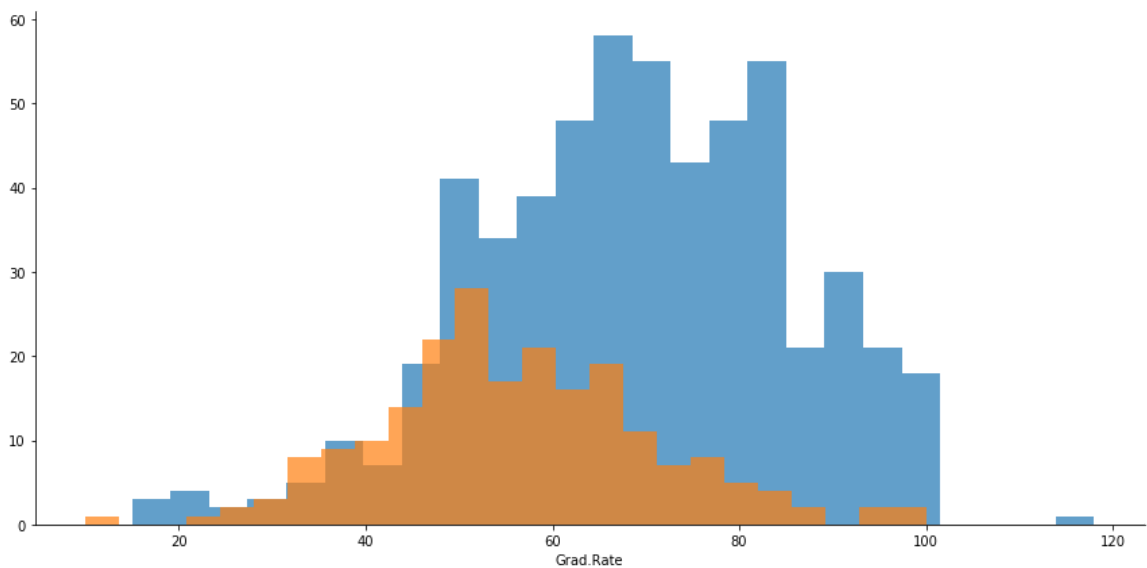
C:\Users\anike\anaconda3\lib\site-packages\seaborn\axisgrid.py:243: UserWarning: The `size` parameter has been renamed to `height`; please update your code.

```
warnings.warn(msg, UserWarning)
```



In [20]:

```
g=sns.FacetGrid(univ_df,hue='Private',size=6,aspect=2)  
g=g.map(plt.hist,'Grad.Rate',bins=25,alpha=0.7)
```



In [21]:

```
#Seems to be a private school with grad rate greater than 100, that  
#isnt possible is it?? Lets find out  
univ_df[univ_df['Grad.Rate']>100]
```

Out[21]:

	Private	Apps	Accept	Enroll	Top10perc	Top25perc	F.Undergrad	P.Undergrad
Cazenovia College	Yes	3847	3433	527	9	35	1010	12

In [23]:

```
#Lets fix this nonsensical number  
univ_df['Grad.Rate']['Cazenovia College']=100
```

C:\Users\anike\anaconda3\lib\site-packages\ipykernel_launcher.py:2: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame

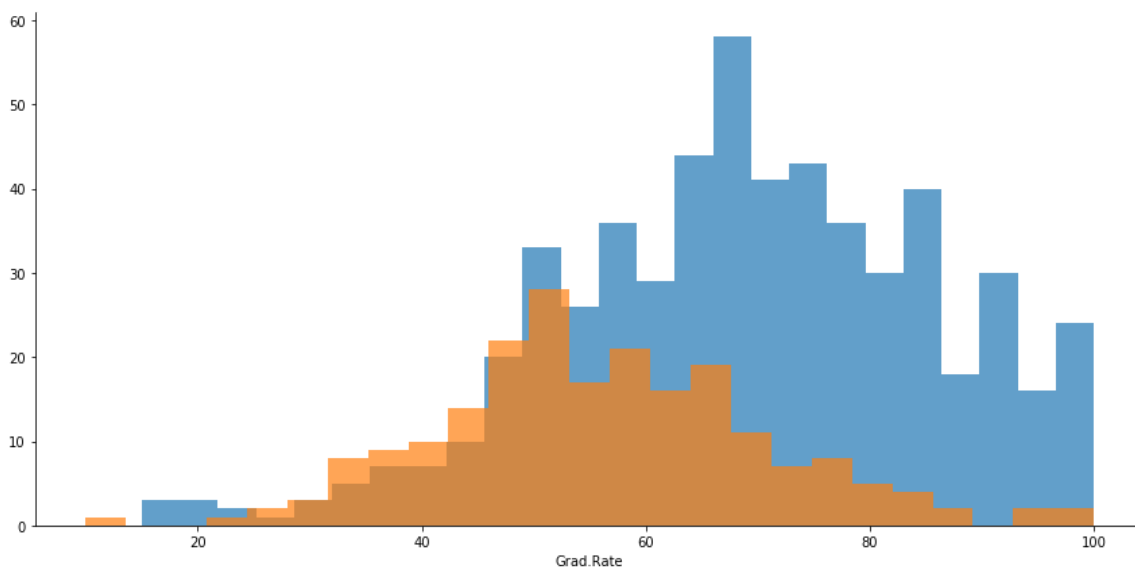
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

In [24]:

```
g=sns.FacetGrid(univ_df,hue='Private',size=6,aspect=2)  
g=g.map(plt.hist,'Grad.Rate',bins=25,alpha=0.7)
```

C:\Users\anike\anaconda3\lib\site-packages\seaborn\axisgrid.py:243: UserWarning: The `size` parameter has been renamed to `height`; please update your code.

```
warnings.warn(msg, UserWarning)
```



In [25]:

```
#FIXED
```

In [26]:

```
#TIME to create the cluster labels
```

In [27]:

```
from sklearn.cluster import KMeans
```

In [28]:

```
kmeans=KMeans(n_clusters=2)
```

In [30]:

```
kmeans.fit(univ_df.drop('Private',axis=1))
```

Out[30]:

```
KMeans(algorithm='auto', copy_x=True, init='k-means++', max_iter=300,  
       n_clusters=2, n_init=10, n_jobs=None, precompute_distances='auto',  
       random_state=None, tol=0.0001, verbose=0)
```

In [32]:

```
kmeans.cluster_centers_  
#dimensions are similar to the number of features on the data set
```

Out[32]:

```
array([[1.81323468e+03, 1.28716592e+03, 4.91044843e+02, 2.53094170e+01,  
       5.34708520e+01, 2.18854858e+03, 5.95458894e+02, 1.03957085e+04,  
       4.31136472e+03, 5.41982063e+02, 1.28033632e+03, 7.04424514e+01,  
       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, 2.46486111e+03, 1.07191759e+04,  
       4.64347222e+03, 5.95212963e+02, 1.71420370e+03, 8.63981481e+01,  
       9.13333333e+01, 1.40277778e+01, 2.00740741e+01, 1.41705000e+04,  
       6.75925926e+01]])
```

In [34]:

```
#There is no perfect way to evaluate clustering if you don't have the labels,we do have  
the labels,  
#so we take advantage of this to evaluate our clusters, keep in mind,  
#you usually won't have this luxury in the real world.
```

```
# Creating a new column for df called 'Cluster', which is a 1 for a Private school, and  
a 0 for a public school.
```

In [35]:

```
#just converting from yes or no strings to 0 or 1 values  
def convert(private):  
    if(private=='Yes'):  
        return 1  
    else:  
        return 0
```

In [37]:

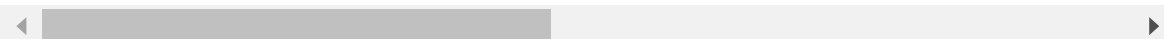
```
univ_df['Cluster']=univ_df['Private'].apply(convert)
```

In [38]:

```
univ_df.head()
```

Out[38]:

	Private	Apps	Accept	Enroll	Top10perc	Top25perc	F.Undergrad	P.Undergrad
Abilene Christian University	Yes	1660	1232	721	23	52	2885	537
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In [39]:

```
from sklearn.metrics import confusion_matrix,classification_report
```

In [41]:

```
print(confusion_matrix(univ_df['Cluster'],kmeans.labels_))
print('\n')
print(classification_report(univ_df['Cluster'],kmeans.labels_))
```

```
[[138  74]
 [531  34]]
```

	precision	recall	f1-score	support
0	0.21	0.65	0.31	212
1	0.31	0.06	0.10	565
accuracy			0.22	777
macro avg	0.26	0.36	0.21	777
weighted avg	0.29	0.22	0.16	777

In []: