

In [1]:

```

1 import pandas as pd
2 import numpy as np
3 import matplotlib.pyplot as plt
4 %matplotlib inline
5 import seaborn as sns
6 from IPython import get_ipython
7 import warnings
8 warnings.filterwarnings("ignore")

```

In [2]:

```
1 data = pd.read_csv('admission_data.csv')
```

In [3]:

```
1 data.head()
```

Out[3]:

| | GRE Score | TOEFL Score | University Rating | SOP | LOR | CGPA | Research | Chance of Admit |
|---|-----------|-------------|-------------------|-----|-----|------|----------|-----------------|
| 0 | 337 | 118 | 4 | 4.5 | 4.5 | 9.65 | 1 | 0.92 |
| 1 | 324 | 107 | 4 | 4.0 | 4.5 | 8.87 | 1 | 0.76 |
| 2 | 316 | 104 | 3 | 3.0 | 3.5 | 8.00 | 1 | 0.72 |
| 3 | 322 | 110 | 3 | 3.5 | 2.5 | 8.67 | 1 | 0.80 |
| 4 | 314 | 103 | 2 | 2.0 | 3.0 | 8.21 | 0 | 0.65 |

In [4]:

```
1 data.tail()
```

Out[4]:

| | GRE Score | TOEFL Score | University Rating | SOP | LOR | CGPA | Research | Chance of Admit |
|-----|-----------|-------------|-------------------|-----|-----|------|----------|-----------------|
| 495 | 332 | 108 | 5 | 4.5 | 4.0 | 9.02 | 1 | 0.87 |
| 496 | 337 | 117 | 5 | 5.0 | 5.0 | 9.87 | 1 | 0.96 |
| 497 | 330 | 120 | 5 | 4.5 | 5.0 | 9.56 | 1 | 0.93 |
| 498 | 312 | 103 | 4 | 4.0 | 5.0 | 8.43 | 0 | 0.73 |
| 499 | 327 | 113 | 4 | 4.5 | 4.5 | 9.04 | 0 | 0.84 |

In [5]:



```
1 data.shape
```

Out[5]:

```
(500, 8)
```

In [6]:



```
1 data.columns
```

Out[6]:

```
Index(['GRE Score', 'TOEFL Score', 'University Rating', 'SOP', 'LOR ', 'CG  
PA',  
      'Research', 'Chance of Admit '],  
      dtype='object')
```

In [7]:



```
1 data.duplicated().sum()
```

Out[7]:

```
0
```

In [8]:



```
1 data.isnull().sum()
```

Out[8]:

```
GRE Score      0  
TOEFL Score    0  
University Rating  0  
SOP            0  
LOR            0  
CGPA           0  
Research       0  
Chance of Admit  0  
dtype: int64
```

In [9]:

```
1 data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 500 entries, 0 to 499
Data columns (total 8 columns):
#   Column                Non-Null Count  Dtype
---  -
0   GRE Score              500 non-null   int64
1   TOEFL Score            500 non-null   int64
2   University Rating      500 non-null   int64
3   SOP                    500 non-null   float64
4   LOR                    500 non-null   float64
5   CGPA                   500 non-null   float64
6   Research               500 non-null   int64
7   Chance of Admit        500 non-null   float64
dtypes: float64(4), int64(4)
memory usage: 31.4 KB
```

In [10]:

```
1 data.describe()
```

Out[10]:

| | GRE Score | TOEFL Score | University Rating | SOP | LOR | CGPA | Research | Chance of Admit |
|-------|------------|-------------|-------------------|------------|------------|------------|------------|-----------------|
| count | 500.000000 | 500.000000 | 500.000000 | 500.000000 | 500.000000 | 500.000000 | 500.000000 | 500.000000 |
| mean | 316.472000 | 107.192000 | 3.114000 | 3.374000 | 3.484000 | 8.576440 | 0.560000 | 0.528000 |
| std | 11.295148 | 6.081868 | 1.143512 | 0.991004 | 0.925450 | 0.604813 | 0.496884 | 0.497446 |
| min | 290.000000 | 92.000000 | 1.000000 | 1.000000 | 1.000000 | 6.800000 | 0.000000 | 0.000000 |
| 25% | 308.000000 | 103.000000 | 2.000000 | 2.500000 | 3.000000 | 8.127500 | 0.000000 | 0.000000 |
| 50% | 317.000000 | 107.000000 | 3.000000 | 3.500000 | 3.500000 | 8.560000 | 1.000000 | 0.000000 |
| 75% | 325.000000 | 112.000000 | 4.000000 | 4.000000 | 4.000000 | 9.040000 | 1.000000 | 0.000000 |
| max | 340.000000 | 120.000000 | 5.000000 | 5.000000 | 5.000000 | 9.920000 | 1.000000 | 1.000000 |

In [11]:



```
1 data.nunique()
```

Out[11]:

```
GRE Score      49
TOEFL Score    29
University Rating  5
SOP            9
LOR            9
CGPA          184
Research        2
Chance of Admit  61
dtype: int64
```

In [12]:



```
1 data['University Rating'].unique()
```

Out[12]:

```
array([4, 3, 2, 5, 1], dtype=int64)
```

In [13]:



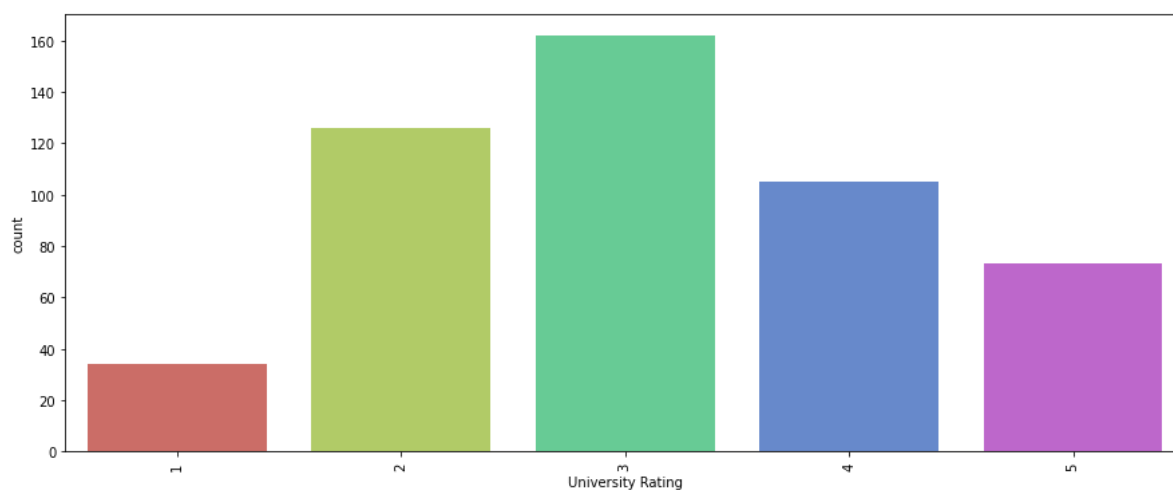
```
1 data['University Rating'].value_counts()
```

Out[13]:

```
3    162
2    126
4    105
5     73
1     34
Name: University Rating, dtype: int64
```

In [14]:

```
1 plt.figure(figsize=(15,6))
2 sns.countplot('University Rating', data = data, palette = 'hls')
3 plt.xticks(rotation = 90)
4 plt.show()
```



In [15]:

```
1 data['SOP'].unique()
```

Out[15]:

```
array([4.5, 4. , 3. , 3.5, 2. , 5. , 1.5, 1. , 2.5])
```

In [16]:

```
1 data['SOP'].value_counts()
```

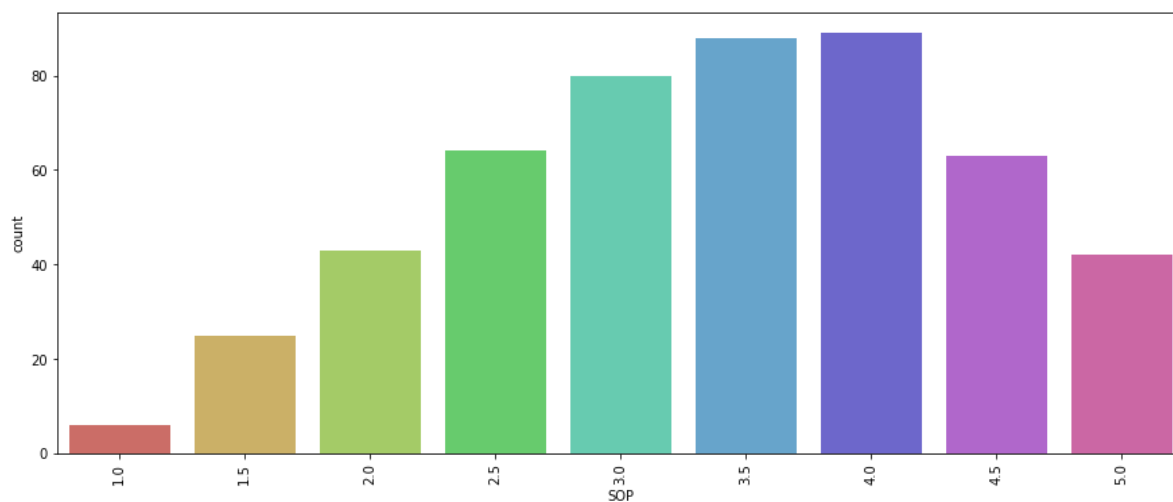
Out[16]:

```
4.0    89
3.5    88
3.0    80
2.5    64
4.5    63
2.0    43
5.0    42
1.5    25
1.0     6
Name: SOP, dtype: int64
```

In [17]:



```
1 plt.figure(figsize=(15,6))
2 sns.countplot('SOP', data = data, palette = 'hls')
3 plt.xticks(rotation = 90)
4 plt.show()
```



In [18]:



```
1 data['LOR '].unique()
```

Out[18]:

```
array([4.5, 3.5, 2.5, 3. , 4. , 1.5, 2. , 5. , 1. ])
```

In [19]:

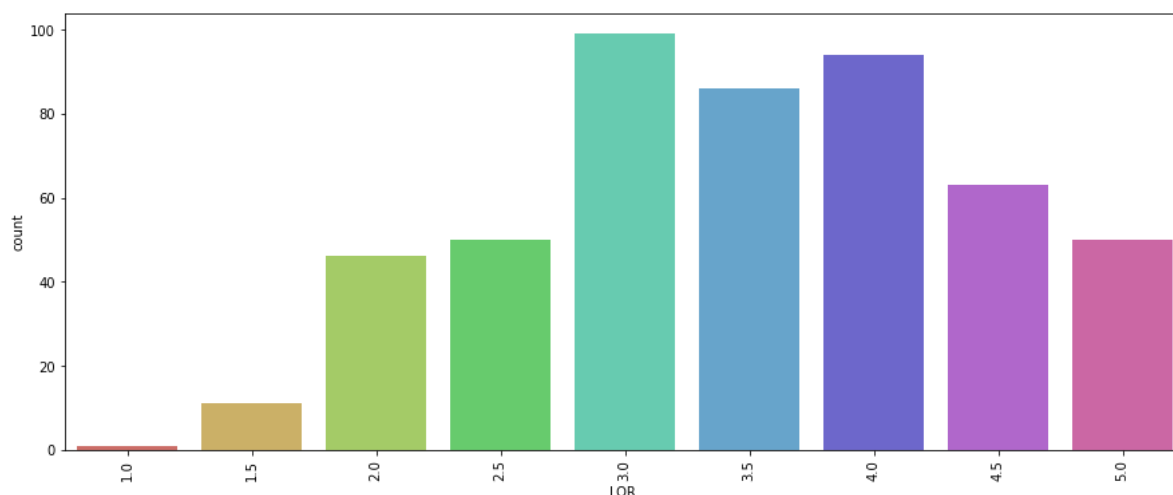
```
1 data['LOR '].value_counts()
```

Out[19]:

```
3.0    99
4.0    94
3.5    86
4.5    63
2.5    50
5.0    50
2.0    46
1.5    11
1.0     1
Name: LOR , dtype: int64
```

In [20]:

```
1 plt.figure(figsize=(15,6))
2 sns.countplot('LOR ', data = data, palette = 'hls')
3 plt.xticks(rotation = 90)
4 plt.show()
```



In [21]:

```
1 data['Research'].unique()
```

Out[21]:

```
array([1, 0], dtype=int64)
```

In [22]:

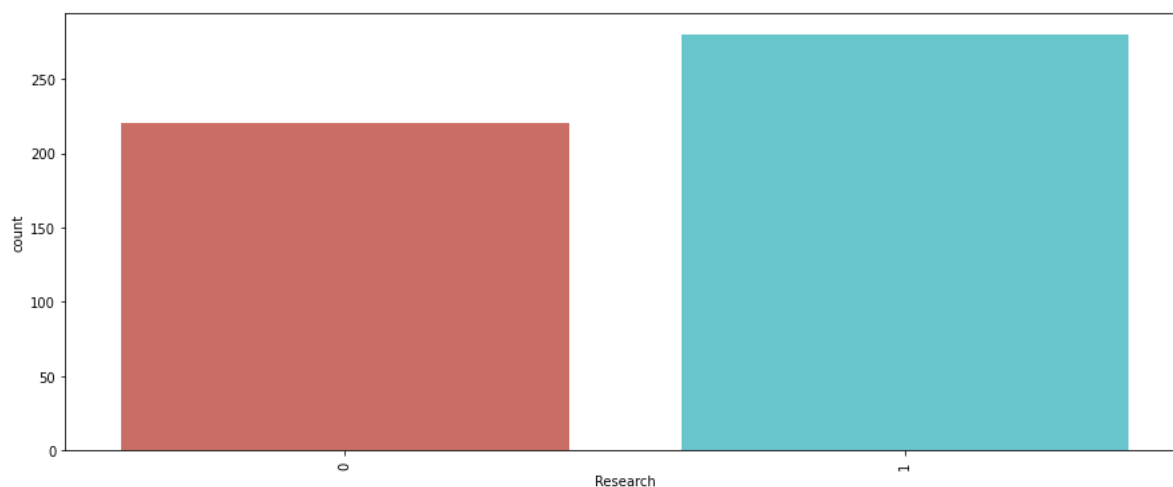
```
1 data['Research'].value_counts()
```

Out[22]:

```
1    280
0    220
Name: Research, dtype: int64
```

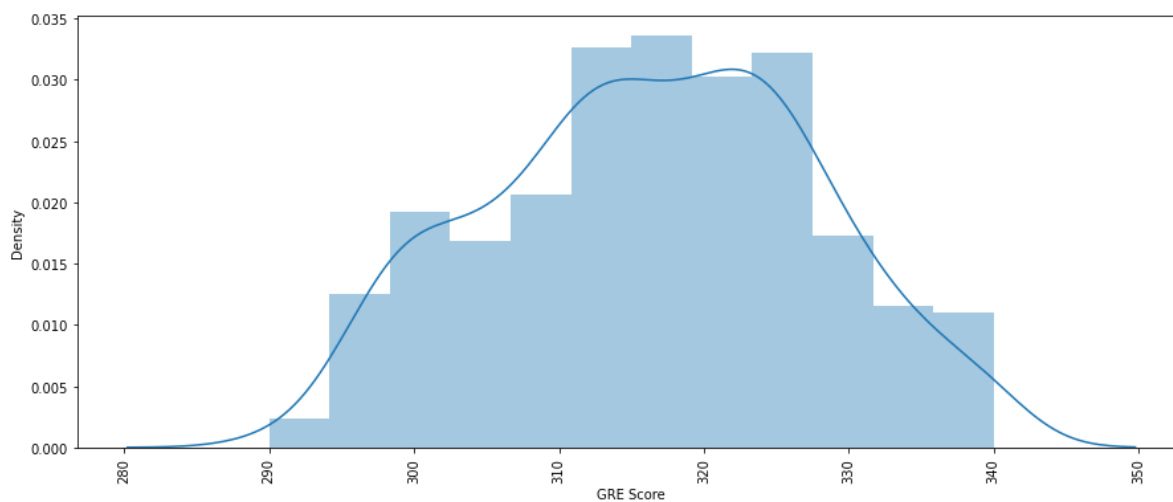
In [23]:

```
1 plt.figure(figsize=(15,6))
2 sns.countplot('Research', data = data, palette = 'hls')
3 plt.xticks(rotation = 90)
4 plt.show()
```



In [24]:

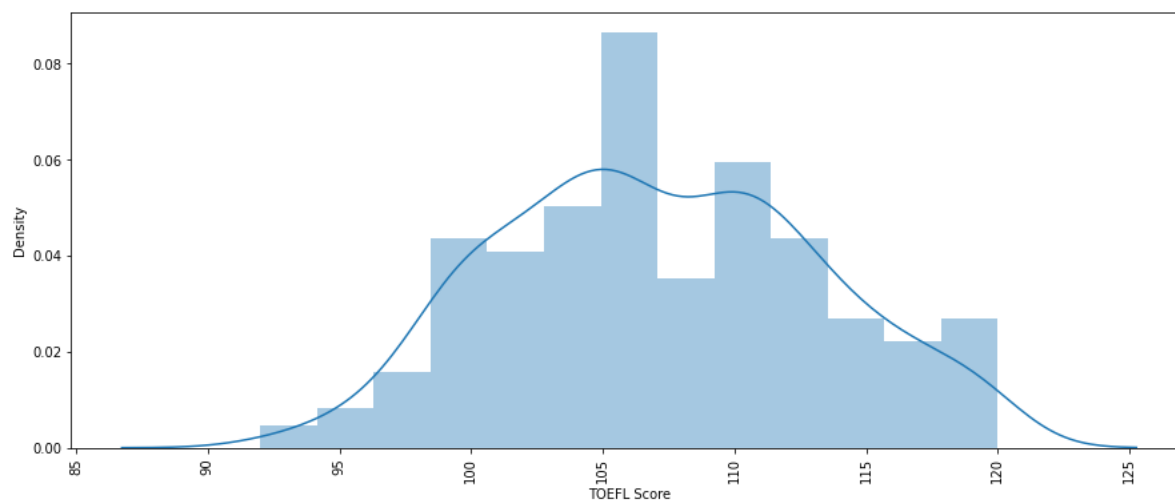
```
1 plt.figure(figsize=(15,6))
2 sns.distplot(data['GRE Score'])
3 plt.xticks(rotation = 90)
4 plt.show()
```



In [25]:



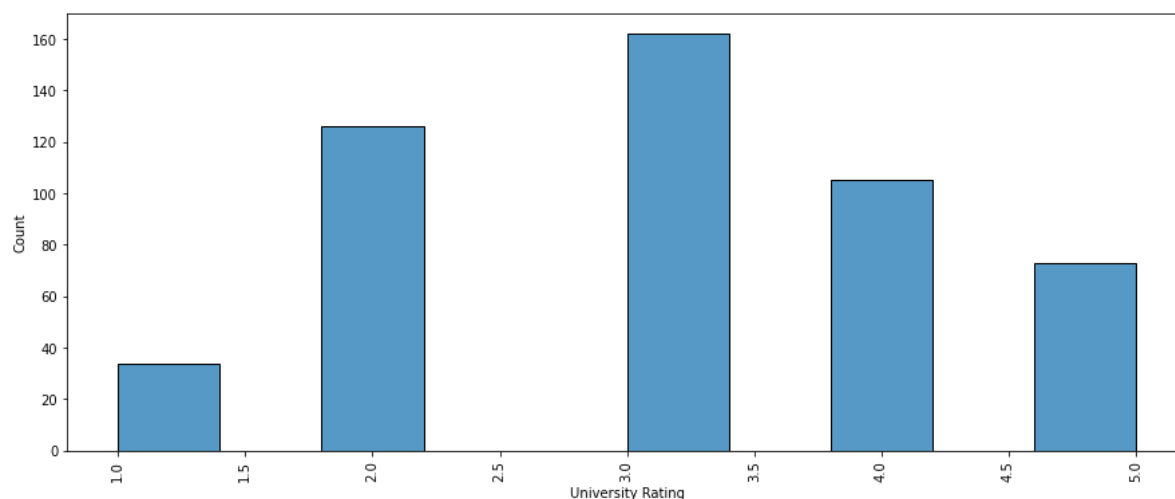
```
1 plt.figure(figsize=(15,6))
2 sns.distplot(data['TOEFL Score'])
3 plt.xticks(rotation = 90)
4 plt.show()
```



In [26]:



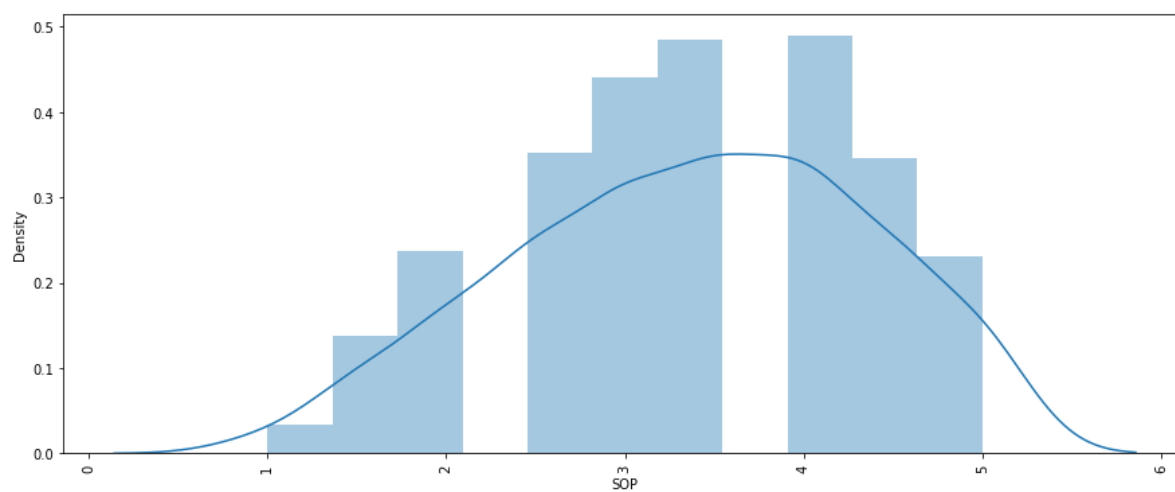
```
1 plt.figure(figsize=(15,6))
2 sns.histplot(data['University Rating'])
3 plt.xticks(rotation = 90)
4 plt.show()
```



In [27]:



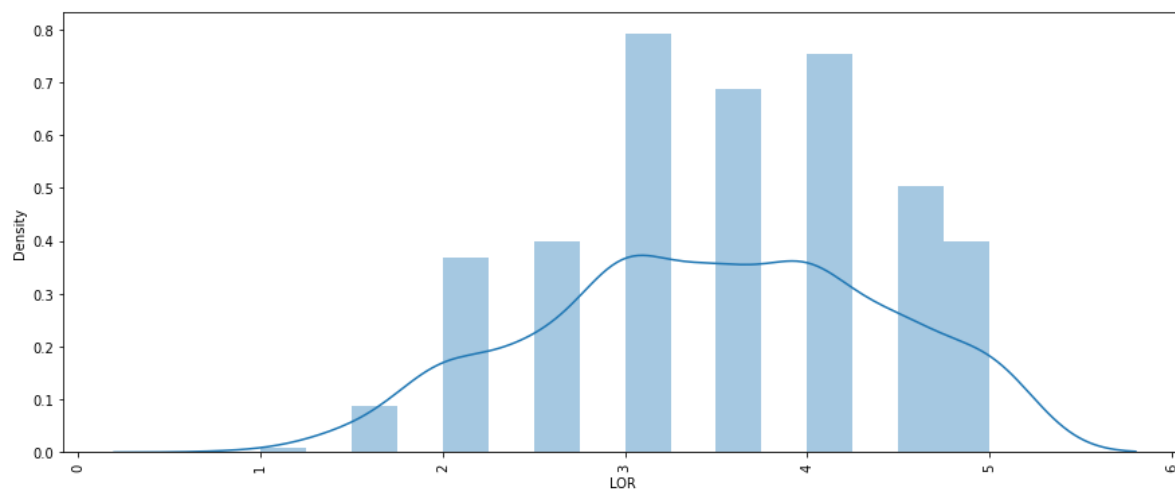
```
1 plt.figure(figsize=(15,6))
2 sns.distplot(data['SOP'])
3 plt.xticks(rotation = 90)
4 plt.show()
```



In [28]:

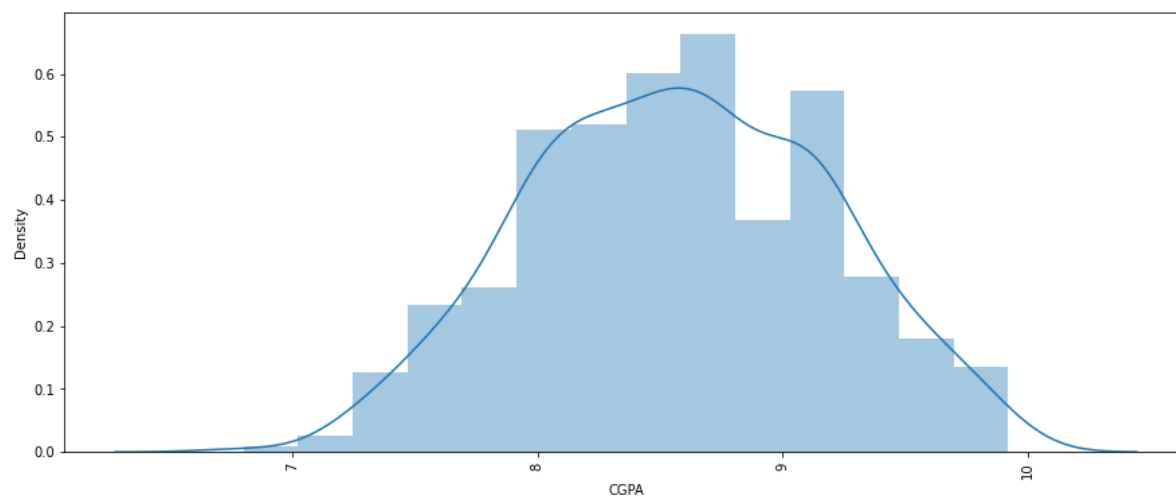


```
1 plt.figure(figsize=(15,6))
2 sns.distplot(data['LOR '])
3 plt.xticks(rotation = 90)
4 plt.show()
```



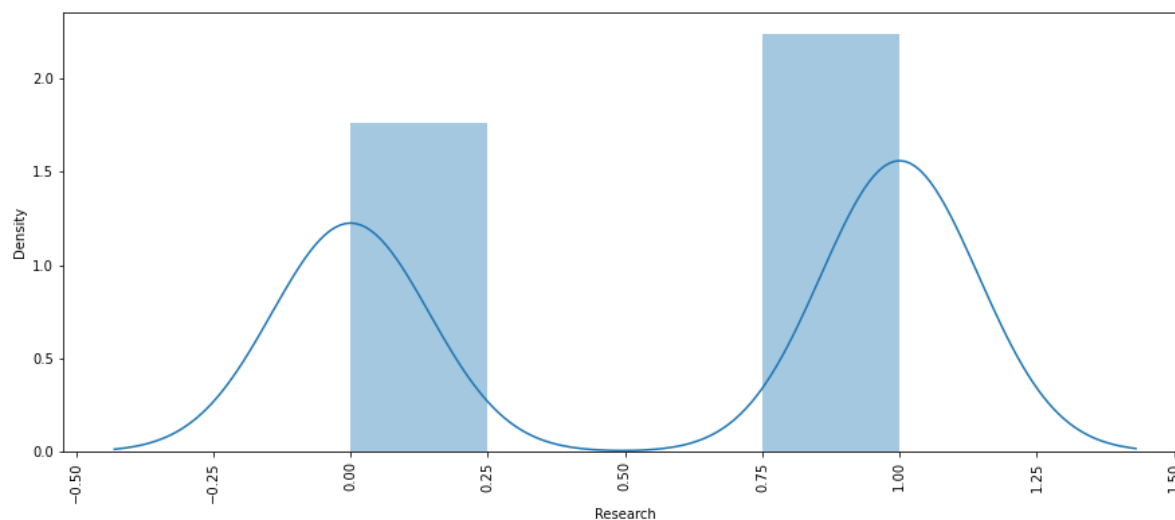
In [29]:

```
1 plt.figure(figsize=(15,6))
2 sns.distplot(data['CGPA'])
3 plt.xticks(rotation = 90)
4 plt.show()
```



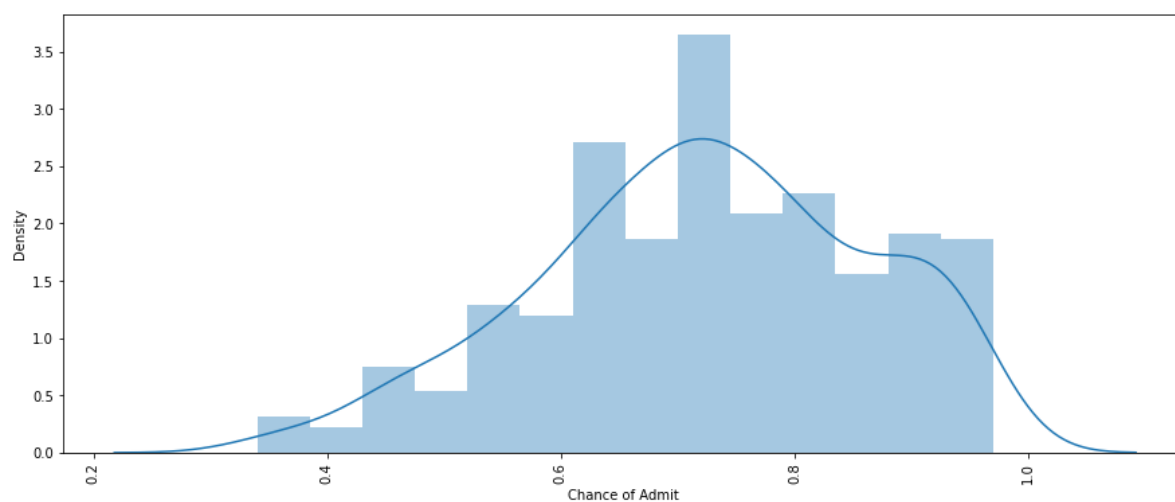
In [30]:

```
1 plt.figure(figsize=(15,6))
2 sns.distplot(data['Research'])
3 plt.xticks(rotation = 90)
4 plt.show()
```



In [31]:

```
1 plt.figure(figsize=(15,6))
2 sns.distplot(data['Chance of Admit '])
3 plt.xticks(rotation = 90)
4 plt.show()
```



In [32]:



```
1 corrmat = data.corr()
2 corrmat
```

Out[32]:

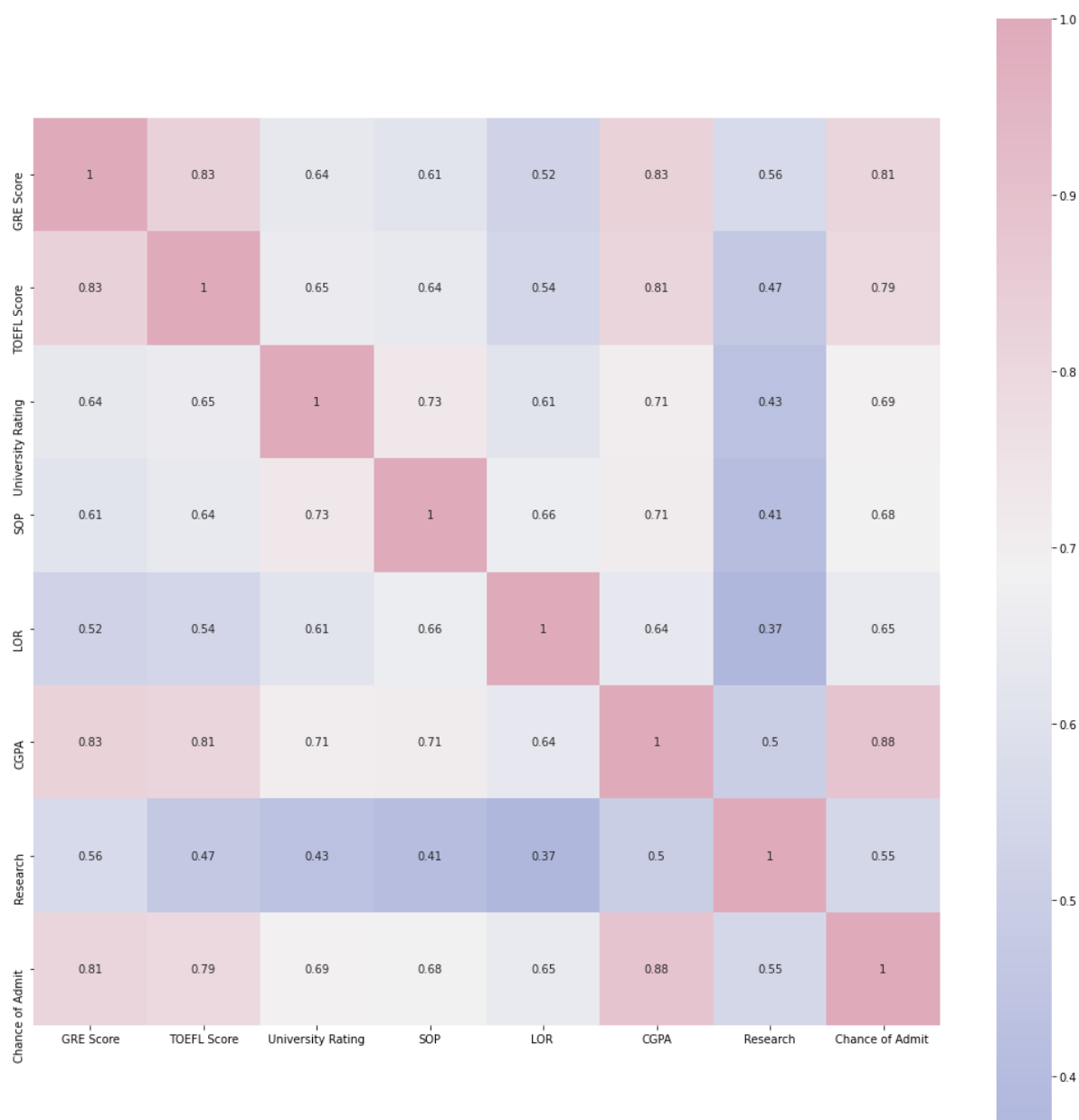
| | GRE Score | TOEFL Score | University Rating | SOP | LOR | CGPA | Research | Chance of Admit |
|----------------------|--------------|----------------|----------------------|----------|----------|----------|----------|--------------------|
| GRE Score | 1.000000 | 0.827200 | 0.635376 | 0.613498 | 0.524679 | 0.825878 | 0.563398 | 0.810351 |
| TOEFL Score | 0.827200 | 1.000000 | 0.649799 | 0.644410 | 0.541563 | 0.810574 | 0.467012 | 0.792228 |
| University Rating | 0.635376 | 0.649799 | 1.000000 | 0.728024 | 0.608651 | 0.705254 | 0.427047 | 0.690132 |
| SOP | 0.613498 | 0.644410 | 0.728024 | 1.000000 | 0.663707 | 0.712154 | 0.408116 | 0.684137 |
| LOR | 0.524679 | 0.541563 | 0.608651 | 0.663707 | 1.000000 | 0.637469 | 0.372526 | 0.645365 |
| CGPA | 0.825878 | 0.810574 | 0.705254 | 0.712154 | 0.637469 | 1.000000 | 0.501311 | 0.882413 |
| Research | 0.563398 | 0.467012 | 0.427047 | 0.408116 | 0.372526 | 0.501311 | 1.000000 | 0.545871 |
| Chance of Admit | 0.810351 | 0.792228 | 0.690132 | 0.684137 | 0.645365 | 0.882413 | 0.545871 | 1.000000 |

In [33]:

```
1 cmap = sns.diverging_palette(260,-10,s=50, l=75, n=6, as_cmap=True)
2 plt.subplots(figsize=(18,18))
3 sns.heatmap(corrmat,cmap= cmap,annot=True, square=True)
```

Out[33]:

<AxesSubplot:>

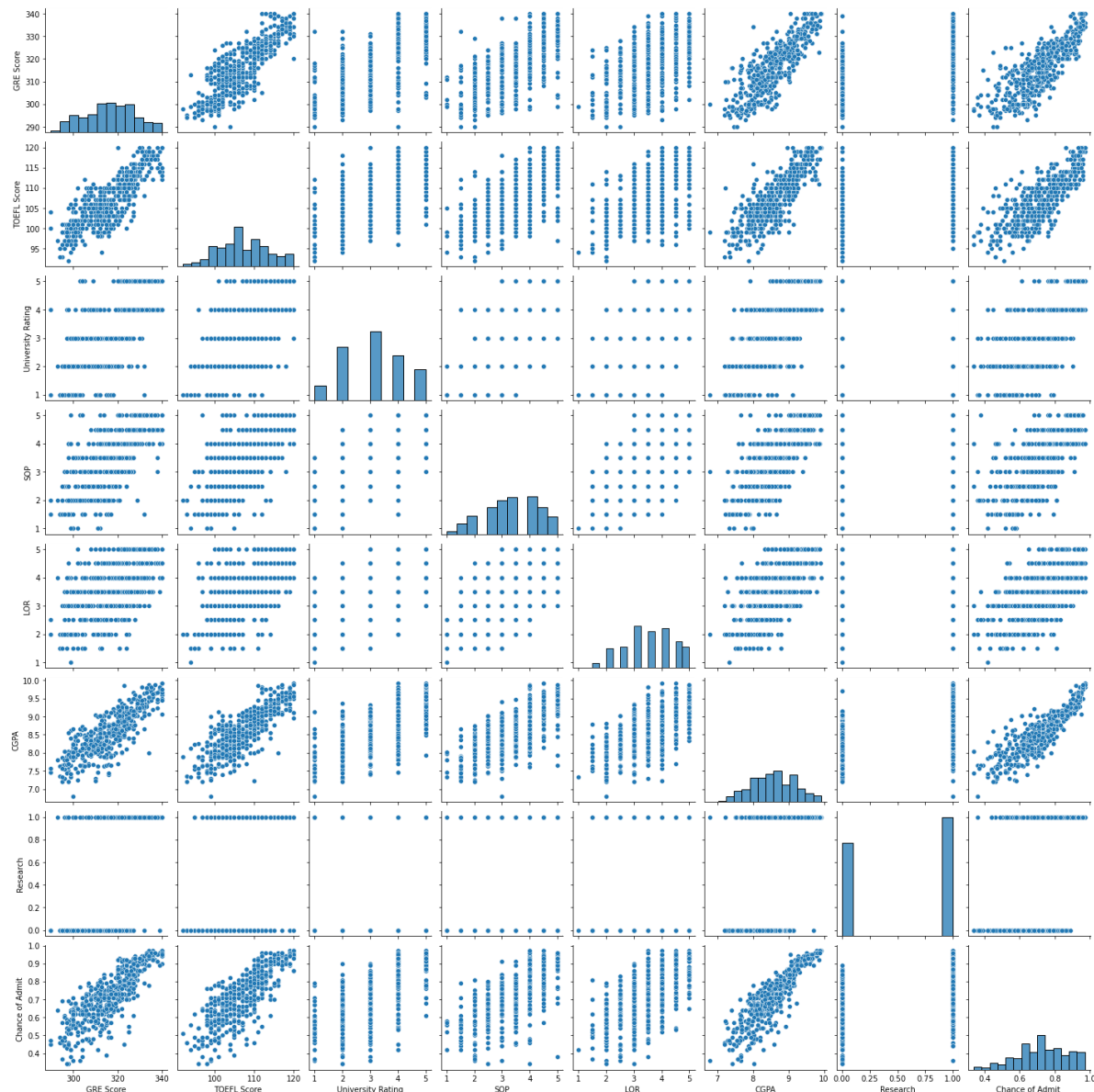


In [34]:

```
1 sns.pairplot(data)
```

Out[34]:

```
<seaborn.axisgrid.PairGrid at 0x6763d6f820>
```



In [35]:

```
1 X=data.drop('Chance of Admit ',axis=1)
2 Y=data['Chance of Admit ']
```

In [36]:

```
1 from sklearn.model_selection import train_test_split
```

In [37]:

```
1 X_train,X_test,y_train,y_test=train_test_split(X,Y,test_size=0.2,
2 random_state=42)
```


In [38]:



```
1 from sklearn.linear_model import LinearRegression
```

In [39]:



```
1 lr=LinearRegression()
```

In [40]:



```
1 lr.fit(X_train,y_train)
```

Out[40]:

```
LinearRegression()
```

In [41]:



```
1 lr_predict=lr.predict(X_test)
```

In [42]:



```
1 train_accuracy_lr=lr.score(X_train,y_train)
2 test_accuracy_lr=lr.score(X_test,y_test)
3 print("Training Accuracy is", train_accuracy_lr)
4 print("Test Accuracy is", test_accuracy_lr)
```

```
Training Accuracy is 0.8210671369321554
```

```
Test Accuracy is 0.8188432567829629
```