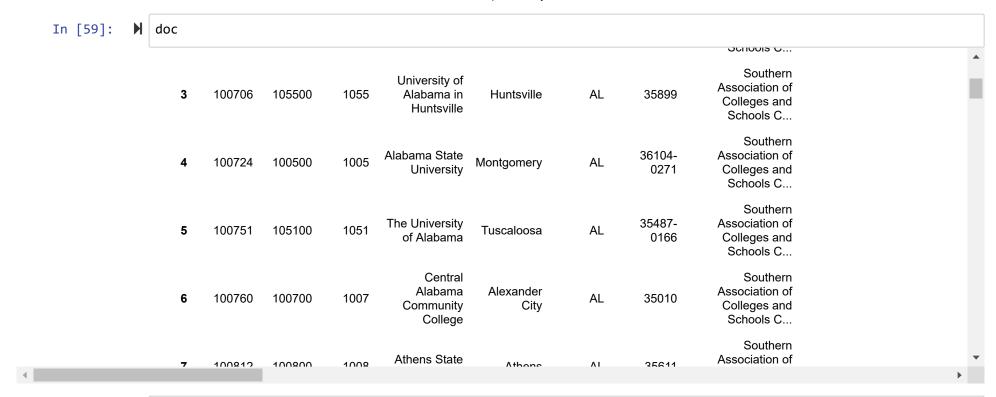
```
In [55]:
          ⋈ import numpy as np
             import pandas as pd
             import matplotlib.pyplot as plt
             import seaborn as sns

    doc = pd.read csv('Pro2 mostrcntdata.csv')

In [56]:
          ▶ doc.shape
In [57]:
    Out[57]: (7058, 1977)
          ▶ doc.columns
In [58]:
   Out[58]: Index(['UNITID', 'OPEID', 'OPEID6', 'INSTNM', 'CITY', 'STABBR', 'ZIP',
                    'ACCREDAGENCY', 'INSTURL', 'NPCURL',
                    'OMAWDP8_NOTFIRSTTIME_POOLED_SUPP', 'OMENRUP_NOTFIRSTTIME_POOLED_SUPP',
                    'OMENRYP_FULLTIME_POOLED_SUPP', 'OMENRAP_FULLTIME_POOLED_SUPP',
                    'OMAWDP8_FULLTIME_POOLED_SUPP', 'OMENRUP_FULLTIME_POOLED_SUPP',
                    'OMENRYP_PARTTIME_POOLED_SUPP', 'OMENRAP_PARTTIME_POOLED_SUPP',
                    'OMAWDP8_PARTTIME_POOLED_SUPP', 'OMENRUP_PARTTIME_POOLED_SUPP'],
                   dtype='object', length=1977)
```



In [60]: print(doc.columns.values)

['UNITID' 'OPEID' 'OPEID6' ... 'OMENRAP_PARTTIME_POOLED_SUPP' 'OMAWDP8_PARTTIME_POOLED_SUPP' 'OMENRUP_PARTTIME_POOLED_SUPP']

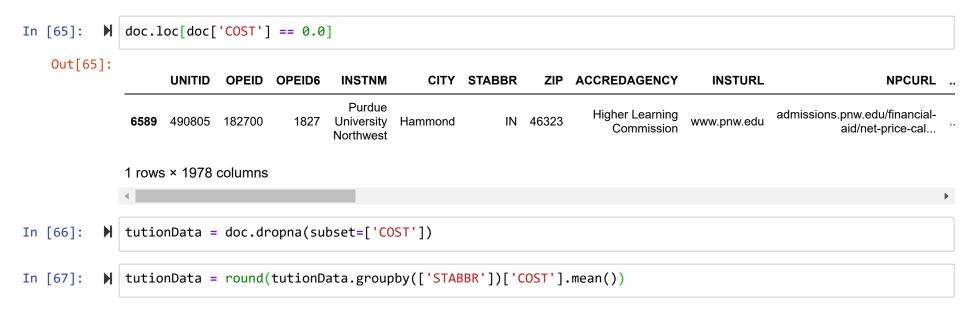
In [61]: ▶	doc.dtypes		
Out[61]:	UNITID	int64	
	OPEID	int64	
	OPEID6	int64	
	INSTNM	object	
	CITY	object	
	STABBR	object	
	ZIP	object	
	ACCREDAGENCY	object	
	INSTURL	object	
	NPCURL	object	
	SCH_DEG	float64	
	HCM2	int64	
	MAIN	int64	
	NUMBRANCH	int64	
	PREDDEG	int64	
	HIGHDEG	int64	
	CONTROL	int64	
	ST_FIPS	int64	
	REGION	int64	
	LOCALE	float64	
	LOCALE2	float64	
	LATITUDE	float64	
	LONGITUDE	float64	
	CCBASIC	float64	
	CCUGPROF	float64	
	CCSIZSET	float64	
	HBCU	float64	
	PBI	float64	
	ANNHI	float64	
	TRIBAL	float64	
		•••	
	OMAWDP8_NOTFIRSTTIME	float64	
	OMENRUP_NOTFIRSTTIME	float64	
	OMENRYP_FULLTIME	float64	
	OMENRAP_FULLTIME	float64	
	OMAWDP8_FULLTIME	float64	
	OMENRUP_FULLTIME	float64	
	OMENRYP_PARTTIME	float64	
	OMENRAP_PARTTIME	float64	
	OMAWDP8_PARTTIME	float64	

OMENRUP_PARTTIME	float64
OMENRYP_ALL_POOLED_SUPP	object
OMENRAP_ALL_POOLED_SUPP	object
OMAWDP8_ALL_POOLED_SUPP	object
OMENRUP_ALL_POOLED_SUPP	object
OMENRYP_FIRSTTIME_POOLED_SUPP	object
OMENRAP_FIRSTTIME_POOLED_SUPP	object
OMAWDP8_FIRSTTIME_POOLED_SUPP	object
OMENRUP_FIRSTTIME_POOLED_SUPP	object
OMENRYP_NOTFIRSTTIME_POOLED_SUPP	object
OMENRAP_NOTFIRSTTIME_POOLED_SUPP	object
OMAWDP8_NOTFIRSTTIME_POOLED_SUPP	object
OMENRUP_NOTFIRSTTIME_POOLED_SUPP	object
OMENRYP_FULLTIME_POOLED_SUPP	object
OMENRAP_FULLTIME_POOLED_SUPP	object
OMAWDP8_FULLTIME_POOLED_SUPP	object
OMENRUP_FULLTIME_POOLED_SUPP	object
OMENRYP_PARTTIME_POOLED_SUPP	object
OMENRAP_PARTTIME_POOLED_SUPP	object
OMAWDP8_PARTTIME_POOLED_SUPP	object
OMENRUP_PARTTIME_POOLED_SUPP	object
Length: 1977, dtype: object	

```
In [62]:
               doc.groupby
           H
    Out[62]: <bound method NDFrame.groupby of</pre>
                                                           UNITID
                                                                     OPEID OPEID6 \
              0
                       100654
                                100200
                                           1002
              1
                       100663
                                105200
                                           1052
              2
                               2503400
                                          25034
                       100690
              3
                       100706
                                105500
                                           1055
              4
                       100724
                                100500
                                           1005
              5
                       100751
                                105100
                                           1051
              6
                       100760
                                100700
                                           1007
              7
                       100812
                                100800
                                           1008
              8
                       100830
                                831000
                                           8310
              9
                       100858
                                100900
                                           1009
              10
                       100937
                                101200
                                           1012
                       101028
                               1218200
                                          12182
              11
              12
                       101073
                               1055400
                                          10554
              13
                       101116
                                          13039
                               1303906
              14
                       101143
                                101500
                                           1015
              15
                       101161
                                106000
                                           1060
              16
                       101189
                                100300
                                           1003
              17
                                101700
                       101240
                                           1017
                       404077
                                4407200
                                           44070
```

Q1> What is the most costly college? What is the cheapest?

Q2> What is the average cost for college for colleges in different parts of the US?



```
In [68]:
           ▶ print ('The avarge tutition for each state')
              print ('In descending order')
              print ('')
              print (tutionData.sort_values(ascending=False))
              The avarge tutition for each state
              In descending order
              STABBR
              VT
                    39869.0
              DC
                    36921.0
                    35063.0
              MA
                    34898.0
              RΙ
                    29007.0
              PΑ
                    28653.0
              NH
              ME
                    28091.0
                    27910.0
              NY
                    26988.0
              IN
              \mathsf{CT}
                    26631.0
                    25968.0
              IΑ
              MD
                    25329.0
                    25254.0
              CA
                    25170.0
              VA
                    25149.0
              NJ
                    24887.0
              MN
              WΙ
                    24553.0
                    24394.0
              DE
              NE
                    24390.0
              SC
                    24323.0
                    23797.0
              OR
              FL
                    23676.0
                    23428.0
              GΑ
                    23277.0
              OH
                    23245.0
              ΙL
                    23154.0
              CO
              NC
                    22907.0
              TN
                    22750.0
                    22736.0
              ΜI
                    22694.0
              NV
                    22621.0
              MO
              ΚY
                    22480.0
                    22177.0
              WA
```

```
KS
      21793.0
TX
      21573.0
      21327.0
ΑZ
SD
      21214.0
      20874.0
LA
ΗI
      20745.0
ΑL
      20662.0
ΑK
      20203.0
UT
      20058.0
ID
      19609.0
MS
      19143.0
      18277.0
AR
      18102.0
WV
NM
      18100.0
      17511.0
OK
ND
      17174.0
ΜT
      17120.0
VI
      16786.0
WY
      14714.0
GU
      12339.0
PR
      11653.0
FΜ
       9554.0
       8750.0
MH
MP
       8734.0
AS
       7400.0
       6085.0
PW
Name: COST, dtype: float64
```

Q3> What is the average cost for college for religious vs. secular institutions?

```
In [73]:  # create new data frame for secular tuition
# drop all rows where religious is NaN
secularTuition = doc.loc[pd.isnull(doc).any(1),:]
# calculate the avarage annual tutition for religious schools
print ('The avarage annual tutition for secullar schools is:')
print (round(secularTuition.COST.mean()),'$')
The avarage annual tutition for secullar schools is:
```

Q4> What percent of colleges have an open admission policy?

There are 4063 schools with open admission which make 58 % of all 7058 schools

Q5> What is the correlation (scatterplot) between admission rates and college cost?

23869 \$

```
AdRt COST.corr()
In [77]:
    Out[77]:
                          ADM_RATE
                                         COST
               ADM_RATE
                             1.000000 -0.301969
                    COST
                                     1.000000
                            -0.301969
In [81]:
           | s = (4,2) |
               plt.scatter(doc.ADM RATE*100, doc.COST, s, color='r')
               plt.title("Correlation Between Admission Rates and College Cost")
               plt.xlabel("admission rates %")
               plt.ylabel("college cost")
               plt.show()
                        correlation between admission rates and college cost
                  70000
                  60000
                  50000
               college cost
                  40000
                  30000
                  20000
                  10000
```

Q6 > What is the correlation between SAT scores and admission rates? Are there any outliers?

100

20

40

admission rates %

60

80

```
In [82]:
          # Observation: There is one outlier just below 600
              doc.SAT AVG.sort values(ascending=True).head(2)
    Out[82]: 825
                      564.0
                      822.0
              2513
             Name: SAT_AVG, dtype: float64
In [83]:

■ sns.boxplot(doc.SAT AVG)
    Out[83]: <matplotlib.axes. subplots.AxesSubplot at 0x27295942f60>
                 600
                         800
                                 1000
                                         1200
                                                 1400
                                                          1600
                                  SAT_AVG
In [85]:
          # Observation: There are outliers close to 0.0
              doc.ADM RATE.sort values(ascending=True).head(5)
    Out[85]: 6065
                      0.000
             5188
                     0.000
              2960
                      0.000
                      0.000
             6610
              2964
                      0.033
```

Q6 >> What colleges have the highest and lowest family income averages?

Name: ADM RATE, dtype: float64

How does that correlate with college costs?

```
In [86]:
          # df['COST'] = df['COST'].fillna(df['COSTT4 P'])
             doc['FAMINC'] = doc['FAMINC'].replace('PrivacySuppressed', np.nan)
             doc['FAMINC'] = doc.FAMINC.astype(float)
          print('The maximal family income registred is: ', round(doc['FAMINC'].max()),'$') # max family income
In [96]:
             print('Institution:',doc.iloc[doc['FAMINC'].idxmax].INSTNM,',', doc.iloc[doc
                                                                                       ['FAMINC'].idxmax].STABBR)
             The maximal family income registred is:
                                                       174263.0 $
             Institution: Jewish Theological Seminary of America , NY
In [91]:
          print('The maximal family income registred is: ', round(doc['FAMINC'].min()),'$')
             print('Institution:',doc.iloc[doc['FAMINC'].idxmin].INSTNM,',', doc.iloc[doc['FAMINC'].idxmin].STABBR)
             The maximal family income registred is:
             Institution: J F Ingram State Technical College , AL
          | famIncm Cost = doc[['FAMINC', 'COST']].copy() # corelation between max income and college cost
In [95]:
            famIncm Cost.corr()
In [94]:
   Out[94]:
                     FAMINC COST
              FAMINC
                      1.0000 0.6758
               COST
                      0.6758 1.0000
 In [ ]:
          H
```