```
#Shaquiel Pashtunyar
In [149...
           #DSC550
           #Week 3 and 4 homework
In [150...
           #basic imports for packets
           import numpy as np
           import pandas as pd
           import matplotlib.pyplot as plt
In [151...
           #getting the boston data from the packt github page
           url = 'https://raw.githubusercontent.com/tirthajyoti/Packt-Data Wrangling/master/Lesso
           boston = pd.read csv(url)
           #first 10 rows of the data set
In [152...
           boston.head(10)
                        ZN INDUS CHAS
                                           NOX
                                                        AGE
                                                                DIS RAD
                                                                           TAX PTRATIO
                                                                                              B LSTAT PF
Out[152]:
                CRIM
                                                   RM
           0.00632
                       18.0
                               2.31
                                                        65.2
                                                             4.0900
                                                                            296
                                                                                     15.3 396.90
                                                                                                   4.98
                                        0 0.538
                                                 6.575
                                                                        1
           1 0.02731
                        0.0
                               7.07
                                                                            242
                                                                                     17.8 396.90
                                                                                                   9.14
                                        0 0.469 6.421
                                                         78.9
                                                             4.9671
                                                                        2
             0.02729
                        0.0
                               7.07
                                        0 0.469
                                                7.185
                                                        61.1 4.9671
                                                                        2
                                                                            242
                                                                                     17.8 392.83
                                                                                                   4.03
                                                                                                   2.94
           3 0.03237
                        0.0
                               2.18
                                        0 0.458
                                                6.998
                                                        45.8
                                                             6.0622
                                                                            222
                                                                                     18.7 394.63
              0.06905
                        0.0
                               2.18
                                        0 0.458 7.147
                                                         54.2 6.0622
                                                                        3
                                                                            222
                                                                                     18.7 396.90
                                                                                                   5.33
           5 0.02985
                        0.0
                               2.18
                                        0 0.458
                                                6.430
                                                         58.7 6.0622
                                                                            222
                                                                                     18.7 394.12
                                                                                                   5.21
              0.08829
                       12.5
                               7.87
                                        0 0.524
                                                 6.012
                                                                           311
                                                                                     15.2 395.60
                                                                                                  12.43
                                                        66.6 5.5605
                                                                        5
           7 0.14455
                               7.87
                                        0 0.524 6.172
                                                         96.1 5.9505
                                                                                     15.2 396.90
                       12.5
                                                                           311
                                                                                                  19.15
           8 0.21124
                       12.5
                               7.87
                                        0 0.524
                                                 5.631
                                                        100.0
                                                              6.0821
                                                                        5
                                                                            311
                                                                                     15.2 386.63
                                                                                                  29.93
           9 0.17004 12.5
                               7.87
                                        0 0.524 6.004
                                                        85.9
                                                             6.5921
                                                                           311
                                                                                     15.2 386.71
                                                                                                  17.10
           #Shape gives us data set shape, so 506 rows and 13 columns
In [153...
           boston.shape
           (506, 14)
Out[153]:
           boston2= boston.drop(['NOX', 'B', 'LSTAT', 'CHAS'], axis=1)
In [154...
In [155...
           #new dataset with less columns
           boston2
```

Out[155]:		CRIM	ZN	INDUS	RM	AGE	DIS	RAD	TAX	PTRATIO	PRICE
	0	0.00632	18.0	2.31	6.575	65.2	4.0900	1	296	15.3	24.0
	1	0.02731	0.0	7.07	6.421	78.9	4.9671	2	242	17.8	21.6
	2	0.02729	0.0	7.07	7.185	61.1	4.9671	2	242	17.8	34.7
	3	0.03237	0.0	2.18	6.998	45.8	6.0622	3	222	18.7	33.4
	4	0.06905	0.0	2.18	7.147	54.2	6.0622	3	222	18.7	36.2
	•••	•••									•••
	501	0.06263	0.0	11.93	6.593	69.1	2.4786	1	273	21.0	22.4
	502	0.04527	0.0	11.93	6.120	76.7	2.2875	1	273	21.0	20.6
	503	0.06076	0.0	11.93	6.976	91.0	2.1675	1	273	21.0	23.9
	504	0.10959	0.0	11.93	6.794	89.3	2.3889	1	273	21.0	22.0
	505	0.04741	0.0	11.93	6.030	80.8	2.5050	1	273	21.0	11.9

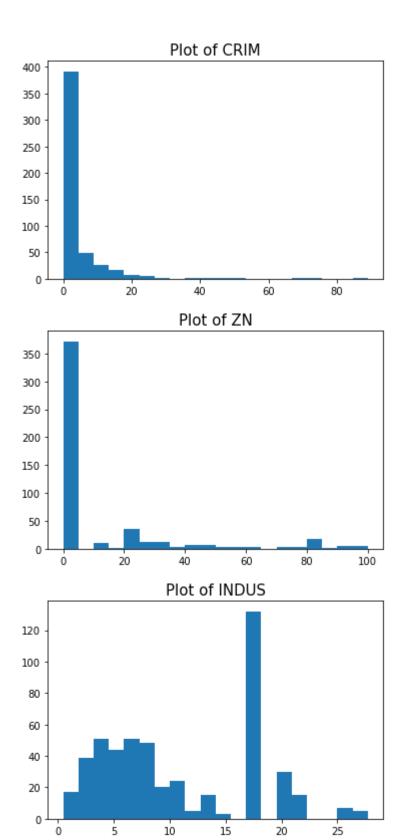
506 rows × 10 columns

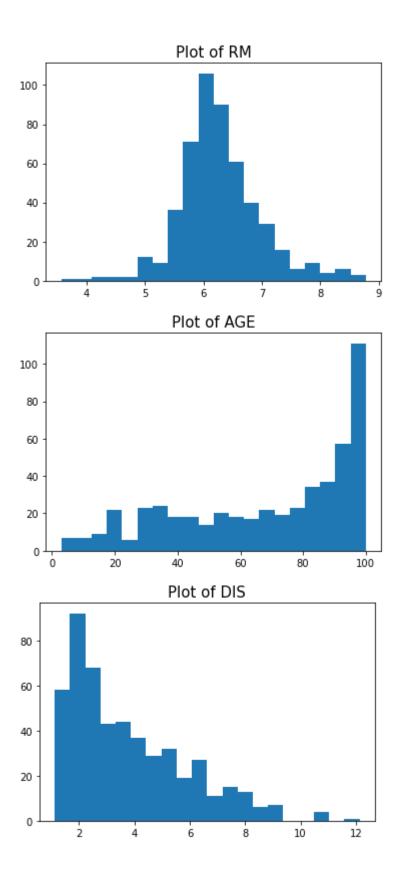
In [156... #final 7 rows of dataset
boston2.tail(7)

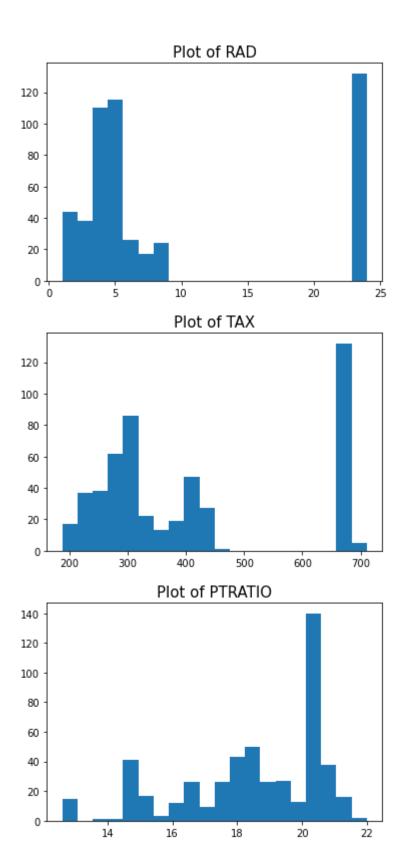
Out[156]:

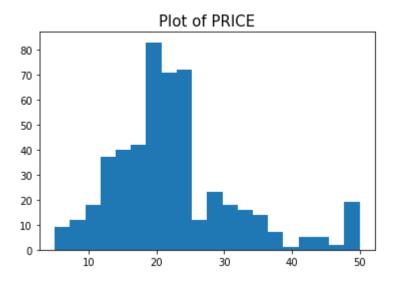
	CRIM	ZN	INDUS	RM	AGE	DIS	RAD	TAX	PTRATIO	PRICE
499	0.17783	0.0	9.69	5.569	73.5	2.3999	6	391	19.2	17.5
500	0.22438	0.0	9.69	6.027	79.7	2.4982	6	391	19.2	16.8
501	0.06263	0.0	11.93	6.593	69.1	2.4786	1	273	21.0	22.4
502	0.04527	0.0	11.93	6.120	76.7	2.2875	1	273	21.0	20.6
503	0.06076	0.0	11.93	6.976	91.0	2.1675	1	273	21.0	23.9
504	0.10959	0.0	11.93	6.794	89.3	2.3889	1	273	21.0	22.0
505	0.04741	0.0	11.93	6.030	80.8	2.5050	1	273	21.0	11.9

```
In [157...
#Wrote a loop that creates all of the plots for me
for c in boston2.columns:
    plt.title("Plot of "+c,fontsize=15)
    plt.hist(boston2[c],bins=20)
    plt.show()
```









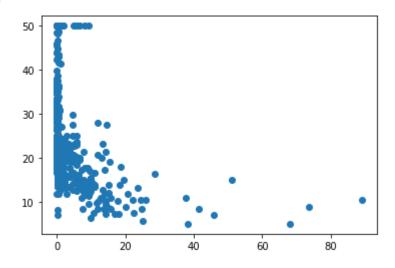
In [158... boston2.head()

\cap	11	+	Γ	1	S.	Q	٦	
\cup	u	L.		_	J	\circ	-	

	CRIM	ZN	INDUS	RM	AGE	DIS	RAD	TAX	PTRATIO	PRICE
0	0.00632	18.0	2.31	6.575	65.2	4.0900	1	296	15.3	24.0
1	0.02731	0.0	7.07	6.421	78.9	4.9671	2	242	17.8	21.6
2	0.02729	0.0	7.07	7.185	61.1	4.9671	2	242	17.8	34.7
3	0.03237	0.0	2.18	6.998	45.8	6.0622	3	222	18.7	33.4
4	0.06905	0.0	2.18	7.147	54.2	6.0622	3	222	18.7	36.2

```
In [159... #Scatter plot creation
    plt.scatter(boston2['CRIM'],boston2['PRICE'])
    plt.show
```

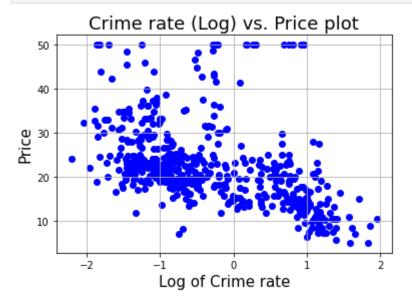
Out[159]: <function matplotlib.pyplot.show(close=None, block=None)>



In [160... display(boston2.dtypes)

```
CRIM
           float64
           float64
ΖN
INDUS
           float64
RM
           float64
           float64
AGE
           float64
DIS
RAD
              int64
              int64
TAX
PTRATIO
           float64
           float64
PRICE
dtype: object
```

```
In [161... #Creating a logrimtic scale to the scatter plot for crim rate. Price going up doesn't
    plt.scatter(np.log10(boston2['CRIM']),boston2['PRICE'],c='blue')
    plt.title("Crime rate (Log) vs. Price plot", fontsize=18)
    plt.xlabel("Log of Crime rate",fontsize=15)
    plt.ylabel("Price",fontsize=15)
    plt.grid(True)
    plt.show()
```



```
boston2.describe
In [162...
           <bound method NDFrame.describe of</pre>
                                                                    INDUS
                                                       CRIM
                                                                ΖN
                                                                               RM
                                                                                    AGE
                                                                                             DIS RAD
Out[162]:
           TAX PTRATIO
                         PRICE
           0
                0.00632
                         18.0
                                 2.31
                                      6.575
                                              65.2
                                                     4.0900
                                                                1
                                                                   296
                                                                            15.3
                                                                                   24.0
           1
                0.02731
                           0.0
                                 7.07
                                       6.421
                                               78.9
                                                     4.9671
                                                                2
                                                                   242
                                                                            17.8
                                                                                   21.6
           2
                                 7.07
                                                                2
                                                                            17.8
                0.02729
                           0.0
                                       7.185
                                               61.1
                                                     4.9671
                                                                   242
                                                                                   34.7
           3
                0.03237
                           0.0
                                 2.18
                                       6.998
                                              45.8
                                                     6.0622
                                                                3
                                                                   222
                                                                            18.7
                                                                                   33.4
           4
                0.06905
                           0.0
                                 2.18
                                       7.147
                                               54.2
                                                     6.0622
                                                                3
                                                                   222
                                                                            18.7
                                                                                   36.2
                                  . . .
                                                                                    . . .
                                11.93
                                       6.593
           501
                0.06263
                           0.0
                                               69.1
                                                     2.4786
                                                                1
                                                                   273
                                                                            21.0
                                                                                   22.4
           502
                0.04527
                           0.0
                                11.93
                                       6.120
                                               76.7
                                                     2.2875
                                                                1
                                                                   273
                                                                            21.0
                                                                                   20.6
           503
                0.06076
                           0.0
                                11.93
                                       6.976
                                               91.0
                                                     2.1675
                                                                1
                                                                   273
                                                                            21.0
                                                                                   23.9
                0.10959
                                11.93
                                       6.794
                                               89.3
                                                                1
                                                                   273
                                                                            21.0
                                                                                   22.0
           504
                           0.0
                                                     2.3889
           505
                0.04741
                           0.0 11.93
                                       6.030 80.8
                                                     2.5050
                                                                1 273
                                                                            21.0
                                                                                   11.9
           [506 rows x 10 columns]>
```

In [163... #mean number of rooms
boston2['RM'].mean()

```
6.284634387351787
Out[163]:
          # median age of house
In [164...
           boston2['AGE'].median()
          77.5
Out[164]:
          # distance of 5 boston employment centers
In [165...
           boston2['DIS'].mean()
           3.795042687747034
Out[165]:
In [166... #create a field with true or false boolean on low price
           low_price=boston2['PRICE']<20</pre>
           # This creates a Boolean array of True, False
           print(low price)
           # True = 1, False = 0, so now if you take an average of this Numpy array, you will kno
           # That many houses are priced below 20,000. So that is the answer.
           # You can convert that into percentage by multiplying with 100
           pcnt=low price.mean()*100
           print("\nPercentage of house with <20,000 price is: ",pcnt)</pre>
          0
                  False
          1
                  False
          2
                  False
          3
                  False
          4
                  False
                  . . .
          501
                 False
          502
                  False
          503
                  False
                  False
          504
          505
                   True
          Name: PRICE, Length: 506, dtype: bool
          Percentage of house with <20,000 price is: 41.50197628458498
          #downloading data file from addult csv
In [167...
           income = pd.read_csv('adult.csv')
In [168...
          income.head()
```

```
Out[168]:
```

In [174...

income.head()

```
educational-
                                                              marital-
               age workclass fnlwgt education
                                                                       occupation relationship
                                                                                                 race gende
                                                        num
                                                                status
                                                                         Machine-
                                                                Never-
           0
                25
                       Private 226802
                                           11th
                                                           7
                                                                                     Own-child
                                                                                                Black
                                                                                                         Ma
                                                               married
                                                                         op-inspct
                                                              Married-
                                                                          Farming-
            1
                38
                       Private
                               89814
                                        HS-grad
                                                           9
                                                                  civ-
                                                                                      Husband White
                                                                                                         Ma
                                                                            fishing
                                                                spouse
                                                              Married-
                                                                        Protective-
                                          Assoc-
           2
                28
                    Local-gov 336951
                                                          12
                                                                  civ-
                                                                                      Husband White
                                                                                                         Ma
                                           acdm
                                                                              serv
                                                                spouse
                                                              Married-
                                                                         Machine-
                                          Some-
                       Private 160323
           3
               44
                                                          10
                                                                                      Husband
                                                                                                Black
                                                                                                         Ma
                                                                  civ-
                                         college
                                                                         op-inspct
                                                                spouse
                                          Some-
                                                                Never-
            4
                18
                           ? 103497
                                                          10
                                                                                ?
                                                                                     Own-child White Fema
                                         college
                                                               married
           #creating script to read txt line by line
In [169...
           #text file open
           names = []
           with open('adultnames.txt','r') as f:
                for line in f:
                    f.readline()
                    var=line.split(":")[0]
                    names.append(var)
In [170...
           #column names
           names
            ['age',
Out[170]:
             'workclass',
             'fnlwgt',
             'education',
             'education-num',
             'marital-status',
             'occupation',
             'relationship',
             'race',
             'sex',
             'capital-gain',
             'capital-loss',
             'hours-per-week',
             'native-country']
           #adding income by using append
In [171...
           names.append('Income')
           #using name column that we created above
In [173...
            url = 'https://raw.githubusercontent.com/TrainingByPackt/Data-Wrangling-with-Python/ma
            income = pd.read_csv(url,names=names)
```

Out[174]:

•		age	workclass	fnlwgt	education	education- num	marital- status	occupation	relationship	race	sex
	0	39	State-gov	77516	Bachelors	13	Never- married	Adm- clerical	Not-in- family	Male	2174
	1	50	Self-emp- not-inc	83311	Bachelors	13	Married- civ- spouse	Exec- managerial	Husband	Male	0
	2	38	Private	215646	HS-grad	9	Divorced	Handlers- cleaners	Not-in- family	Male	0
	3	53	Private	234721	11th	7	Married- civ- spouse	Handlers- cleaners	Husband	Male	0
	4	28	Private	338409	Bachelors	13	Married- civ- spouse	Prof- specialty	Wife	Female	0

In [175... #info on the dataset income.describe()

Out[175]:

	age	fnlwgt	education- num	sex	capital-gain	capital-loss	Income
count	32561.000000	3.256100e+04	32561.000000	32561.000000	32561.000000	32561.000000	0.0
mean	38.581647	1.897784e+05	10.080679	1077.648844	87.303830	40.437456	NaN
std	13.640433	1.055500e+05	2.572720	7385.292085	402.960219	12.347429	NaN
min	17.000000	1.228500e+04	1.000000	0.000000	0.000000	1.000000	NaN
25%	28.000000	1.178270e+05	9.000000	0.000000	0.000000	40.000000	NaN
50%	37.000000	1.783560e+05	10.000000	0.000000	0.000000	40.000000	NaN
75%	48.000000	2.370510e+05	12.000000	0.000000	0.000000	45.000000	NaN
max	90.000000	1.484705e+06	16.000000	99999.000000	4356.000000	99.000000	NaN

In [177... #number of missing datapoints looks like none except for income income.isnull().sum()

```
0
          age
Out[177]:
                                 0
          workclass
           fnlwgt
                                 0
                                 0
           education
           education-num
                                 0
          marital-status
                                 0
           occupation
                                 0
           relationship
                                 0
           race
                                 0
                                 0
           sex
                                 0
           capital-gain
           capital-loss
                                 0
           hours-per-week
                                 0
           native-country
                                 0
           Income
                             32561
           dtype: int64
          #Subset data frame with only select columns
 In [178...
           income_subset = income[['age','education','occupation','race']]
In [179... | #age histogram with a bin of 20
           income_subset['age'].hist(bins=20)
           <AxesSubplot:>
Out[179]:
           3500
           3000
           2500
           2000
           1500
           1000
            500
             0
                   20
                         30
                               40
                                     50
                                           60
                                                 70
                                                       80
          #whitespace striper
In [180...
           def strip whitespace(s):
               return s.strip()
In [183...
          #applying the whitespace stripper to the 3 columns in my subset
           # Education column
           income_subset['education_stripped']=income['education'].apply(strip_whitespace)
           income_subset['education']=income_subset['education_stripped']
           income_subset.drop(labels=['education_stripped'],axis=1,inplace=True)
           # Occupation column
           income_subset['occupation_stripped']=income['occupation'].apply(strip_whitespace)
           income_subset['occupation']=income_subset['occupation_stripped']
           income_subset.drop(labels=['occupation_stripped'],axis=1,inplace=True)
           # Race column
           income_subset['race_stripped']=income['race'].apply(strip_whitespace)
```

```
income_subset['race']=income_subset['race_stripped']
income_subset.drop(labels=['race_stripped'],axis=1,inplace=True)
```

```
C:\Users\spashtunyar\AppData\Local\Temp\ipykernel_34080\240269240.py:3: SettingWithCo
pyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/us
er guide/indexing.html#returning-a-view-versus-a-copy
 income subset['education stripped']=income['education'].apply(strip whitespace)
C:\Users\spashtunyar\AppData\Local\Temp\ipykernel_34080\240269240.py:4: SettingWithCo
pyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row indexer,col indexer] = value instead
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/us
er_guide/indexing.html#returning-a-view-versus-a-copy
  income_subset['education']=income_subset['education_stripped']
C:\Users\spashtunyar\AppData\Local\Temp\ipykernel 34080\240269240.py:5: SettingWithCo
pyWarning:
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/us
er guide/indexing.html#returning-a-view-versus-a-copy
  income_subset.drop(labels=['education_stripped'],axis=1,inplace=True)
C:\Users\spashtunyar\AppData\Local\Temp\ipykernel_34080\240269240.py:8: SettingWithCo
pyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row indexer,col indexer] = value instead
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/us
er guide/indexing.html#returning-a-view-versus-a-copy
  income_subset['occupation_stripped']=income['occupation'].apply(strip_whitespace)
C:\Users\spashtunyar\AppData\Local\Temp\ipykernel_34080\240269240.py:9: SettingWithCo
pyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row indexer,col indexer] = value instead
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/us
er_guide/indexing.html#returning-a-view-versus-a-copy
  income subset['occupation']=income subset['occupation stripped']
C:\Users\spashtunyar\AppData\Local\Temp\ipykernel 34080\240269240.py:10: SettingWithC
opyWarning:
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/us
er guide/indexing.html#returning-a-view-versus-a-copy
  income_subset.drop(labels=['occupation_stripped'],axis=1,inplace=True)
C:\Users\spashtunyar\AppData\Local\Temp\ipykernel_34080\240269240.py:13: SettingWithC
opyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row indexer,col indexer] = value instead
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/us
er guide/indexing.html#returning-a-view-versus-a-copy
  income_subset['race_stripped']=income['race'].apply(strip_whitespace)
C:\Users\spashtunyar\AppData\Local\Temp\ipykernel 34080\240269240.py:14: SettingWithC
opyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row indexer,col indexer] = value instead
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/us
```

```
er_guide/indexing.html#returning-a-view-versus-a-copy
             income_subset['race']=income_subset['race_stripped']
          C:\Users\spashtunyar\AppData\Local\Temp\ipykernel_34080\240269240.py:15: SettingWithC
          opyWarning:
          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/us
          er_guide/indexing.html#returning-a-view-versus-a-copy
             income_subset.drop(labels=['race_stripped'],axis=1,inplace=True)
          # people between 30 and 50
In [186...
           income_filtered=income_subset[(income_subset['age']>=30) & (income_subset['age']<=50)]</pre>
           income_filtered.head()
In [187...
Out[187]:
              age
                   education
                                  occupation
                                               race
               39
                    Bachelors
                                 Adm-clerical
                                              Male
           1
               50
                    Bachelors
                              Exec-managerial
                                              Male
           2
               38
                    HS-grad Handlers-cleaners
                                              Male
           5
               37
                     Masters
                              Exec-managerial Female
           6
               49
                        9th
                                Other-service Female
In [188...
           #number of people who are between 30 and 50
           income_filtered.shape[0]
           16390
Out[188]:
In [193...
           #the mean column give us the average age at each educational level
           income_subset.groupby('education').describe()['age']
```

```
12th
                             433.0 32.000000
                                              14.334625
                                                          17.0
                                                               19.00
                                                                       28.0
                                                                            41.0
                                                                                   79.0
                  1st-4th
                             168.0 46.142857
                                              15.615625
                                                          19.0
                                                               33.00
                                                                       46.0
                                                                             57.0
                                                                                   90.0
                  5th-6th
                             333.0 42.885886
                                              15.557285
                                                          17.0
                                                               29.00
                                                                       42.0
                                                                             54.0
                                                                                   84.0
                  7th-8th
                             646.0 48.445820
                                              16.092350
                                                          17.0
                                                               34.25
                                                                       50.0
                                                                             61.0
                                                                                   90.0
                      9th
                             514.0 41.060311
                                              15.946862
                                                          17.0
                                                               28.00
                                                                       39.0
                                                                             54.0
                                                                                   90.0
              Assoc-acdm
                            1067.0 37.381443
                                              11.095177
                                                          19.0
                                                               29.00
                                                                       36.0
                                                                             44.0
                                                                                   90.0
               Assoc-voc
                            1382.0 38.553546
                                              11.631300
                                                          19.0
                                                               30.00
                                                                       37.0
                                                                             46.0
                                                                                   84.0
                Bachelors
                            5355.0 38.904949
                                              11.912210
                                                          19.0
                                                               29.00
                                                                       37.0
                                                                             46.0
                                                                                   90.0
                Doctorate
                             413.0 47.702179 11.784716
                                                          24.0
                                                               39.00
                                                                       47.0
                                                                             55.0
                                                                                   80.0
                 HS-grad
                           10501.0 38.974479
                                               13.541524
                                                          17.0
                                                               28.00
                                                                       37.0
                                                                             48.0
                                                                                   90.0
                 Masters
                            1723.0 44.049913 11.068935
                                                          18.0
                                                               36.00
                                                                       43.0
                                                                             51.0
                                                                                   90.0
                Preschool
                              51.0 42.764706
                                              15.126914
                                                          19.0
                                                               31.00
                                                                       41.0
                                                                             53.5
                                                                                   75.0
                                                          25.0
                                                               36.00
                                                                                   90.0
              Prof-school
                             576.0 44.746528
                                              11.962477
                                                                       43.0
                                                                             51.0
            Some-college
                            7291.0 35.756275 13.474051
                                                         17.0 24.00
                                                                       34.0
                                                                            45.0
                                                                                   90.0
            #creating data series
In [200...
            Data1 = 7.3, -2.5, 3.4, 1.5
            Data2 = -2.1, 3.6, -1.5, 4, 3.1
 In [202...
            #pandas dataset
            series1 = pd.Series(Data1);
            series2 = pd.Series(Data2);
In [205...
            series1
                  7.3
Out[205]:
            1
                 -2.5
            2
                  3.4
                  1.5
            dtype: float64
In [212...
            #add them together
            Seriesadd = np.add(series1,series2)
            Seriesadd
                  5.2
Out[212]:
                  1.1
                  1.9
            2
            3
                  5.5
            4
                  NaN
            dtype: float64
```

Out[193]:

count

education

10th

11th

mean

1175.0 32.355745 15.545485

933.0 37.429796 16.720713 17.0

std min 25% 50% 75% max

34.0

28.0

52.0

43.0

90.0

90.0

22.00

18.00

17.0