

Brian_Reppeto540Week3_4

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0.0.1 DSC 540 Week 2 Data Wrangling with Python:

0.0.2 Activity 5 Generating Stats from a csv file

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```
[94]: # import libraries

import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

```
[95]: #read the Boston housing dataset

bost_hs_df=pd.read_csv("Boston_housing.csv")
```

```
[96]: # read the first 10 rows

bost_hs_df.head(10)
```

```
[96]:
```

	CRIM	ZN	INDUS	CHAS	NOX	RM	AGE	DIS	RAD	TAX	PTRATIO	\
0	0.00632	18.0	2.31	0	0.538	6.575	65.2	4.0900	1	296	15.3	
1	0.02731	0.0	7.07	0	0.469	6.421	78.9	4.9671	2	242	17.8	
2	0.02729	0.0	7.07	0	0.469	7.185	61.1	4.9671	2	242	17.8	
3	0.03237	0.0	2.18	0	0.458	6.998	45.8	6.0622	3	222	18.7	
4	0.06905	0.0	2.18	0	0.458	7.147	54.2	6.0622	3	222	18.7	
5	0.02985	0.0	2.18	0	0.458	6.430	58.7	6.0622	3	222	18.7	
6	0.08829	12.5	7.87	0	0.524	6.012	66.6	5.5605	5	311	15.2	
7	0.14455	12.5	7.87	0	0.524	6.172	96.1	5.9505	5	311	15.2	
8	0.21124	12.5	7.87	0	0.524	5.631	100.0	6.0821	5	311	15.2	
9	0.17004	12.5	7.87	0	0.524	6.004	85.9	6.5921	5	311	15.2	

	B	LSTAT	PRICE
0	396.90	4.98	24.0
1	396.90	9.14	21.6
2	392.83	4.03	34.7
3	394.63	2.94	33.4
4	396.90	5.33	36.2
5	394.12	5.21	28.7

```

6  395.60  12.43  22.9
7  396.90  19.15  27.1
8  386.63  29.93  16.5
9  386.71  17.10  18.9

```

```
[97]: # Use Np shape array to find the rows and column count. 506 rows 14 columns
```

```
bost_hs_df.shape
```

```
[97]: (506, 14)
```

```
[98]: # Smaller DF with selected columns from the original df
```

```
bost_hs_sm_df=bost_hs_df[['CRIM','ZN','INDUS','RM','AGE','DIS','RAD','TAX','PTRATIO','PRICE']]
```

```
[99]: # Check last 7 records using tail (7) to grab the last 7 rows
```

```
bost_hs_sm_df.tail(7)
```

```
[99]:
```

	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

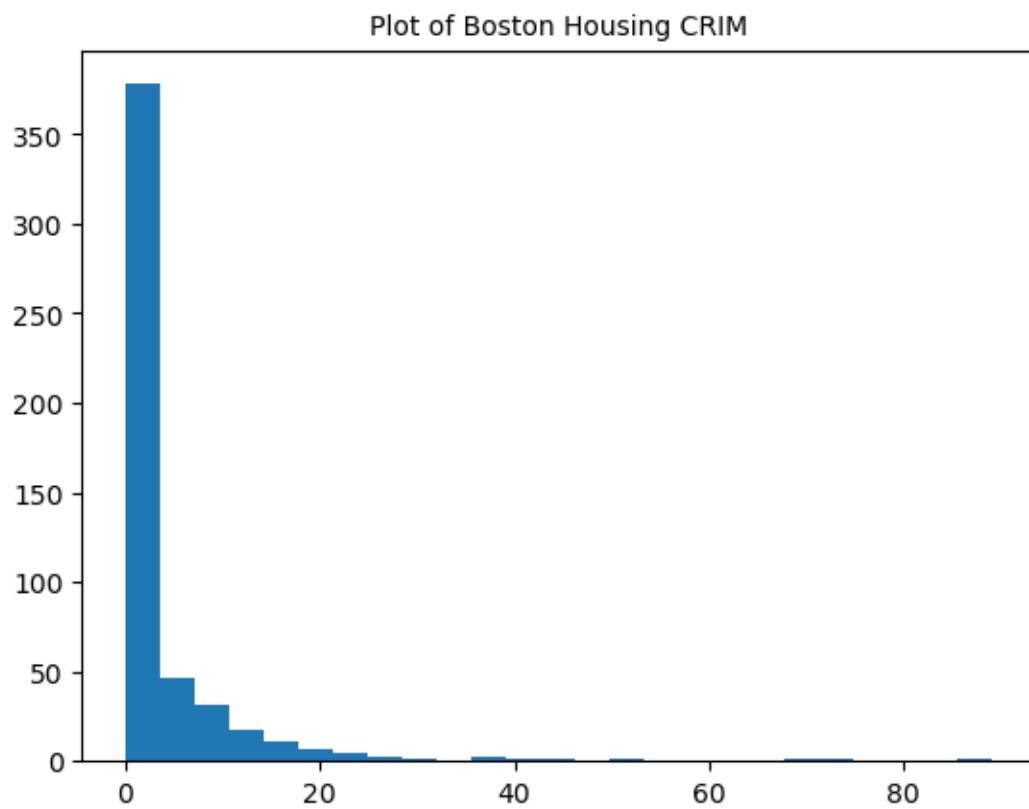
```
[100]: # Using matplotlib and the bost_hs_sm df create histograms for each of the columns
```

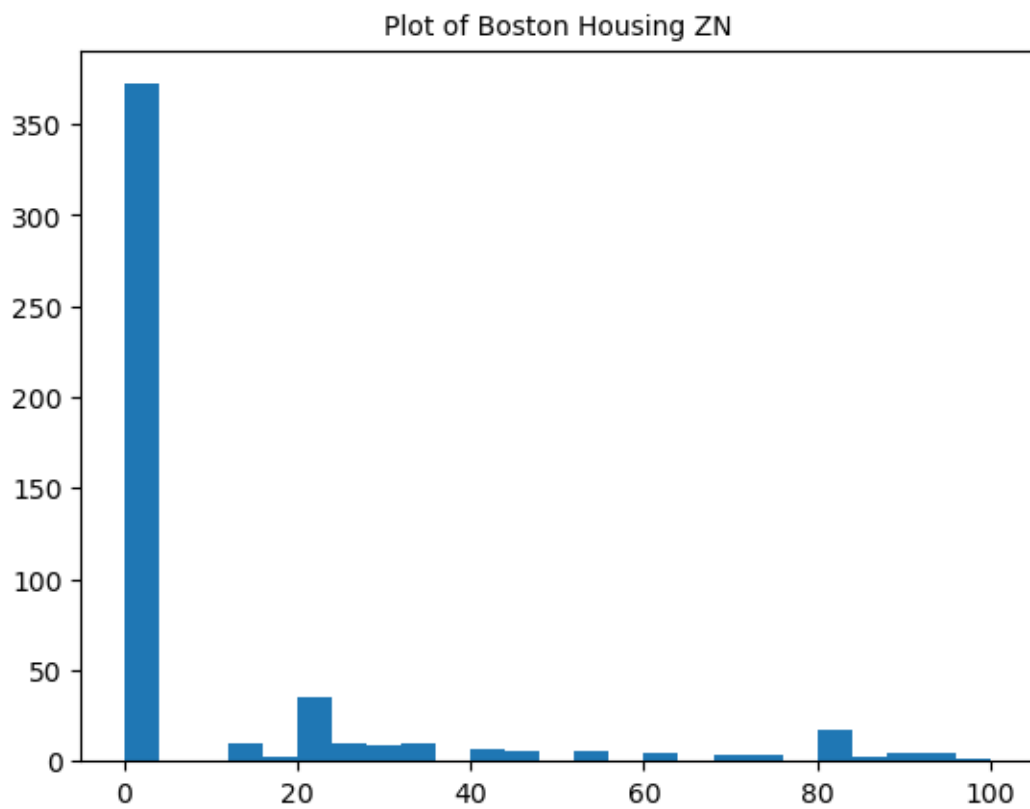
```
# Iterate over columns
```

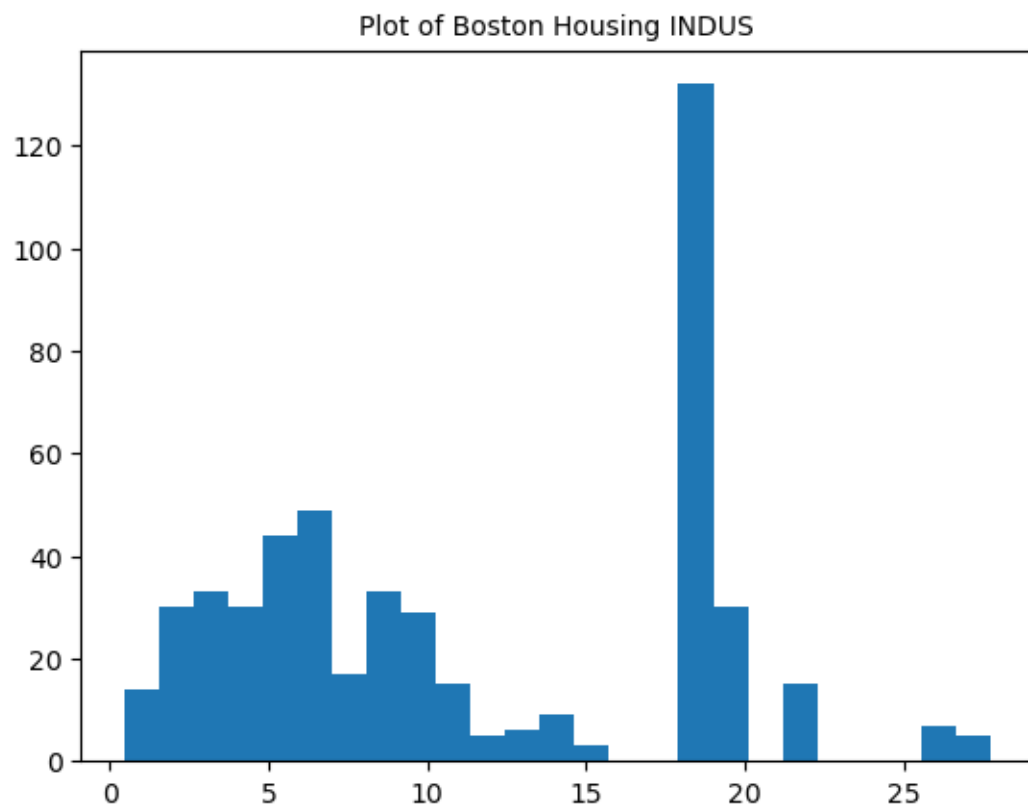
```

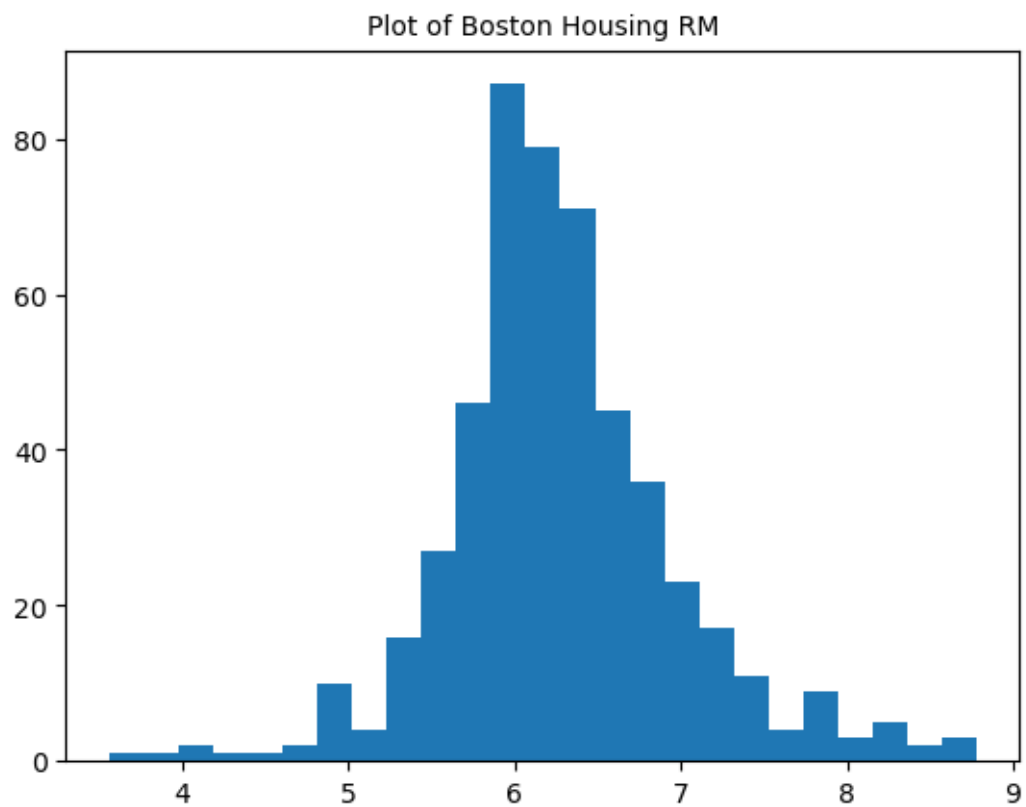
for i in bost_hs_sm_df.columns:
    plt.title("Plot of Boston Housing "+ i,fontsize=10)
    plt.hist(bost_hs_sm_df[i],bins=25)
    plt.show()

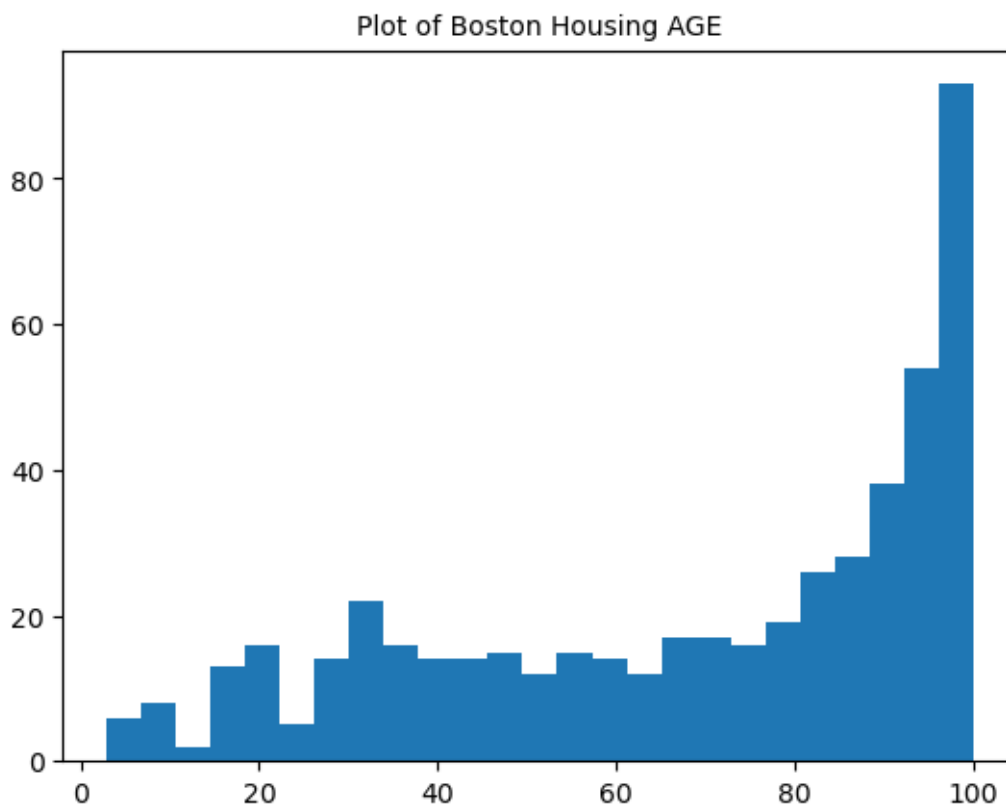
```

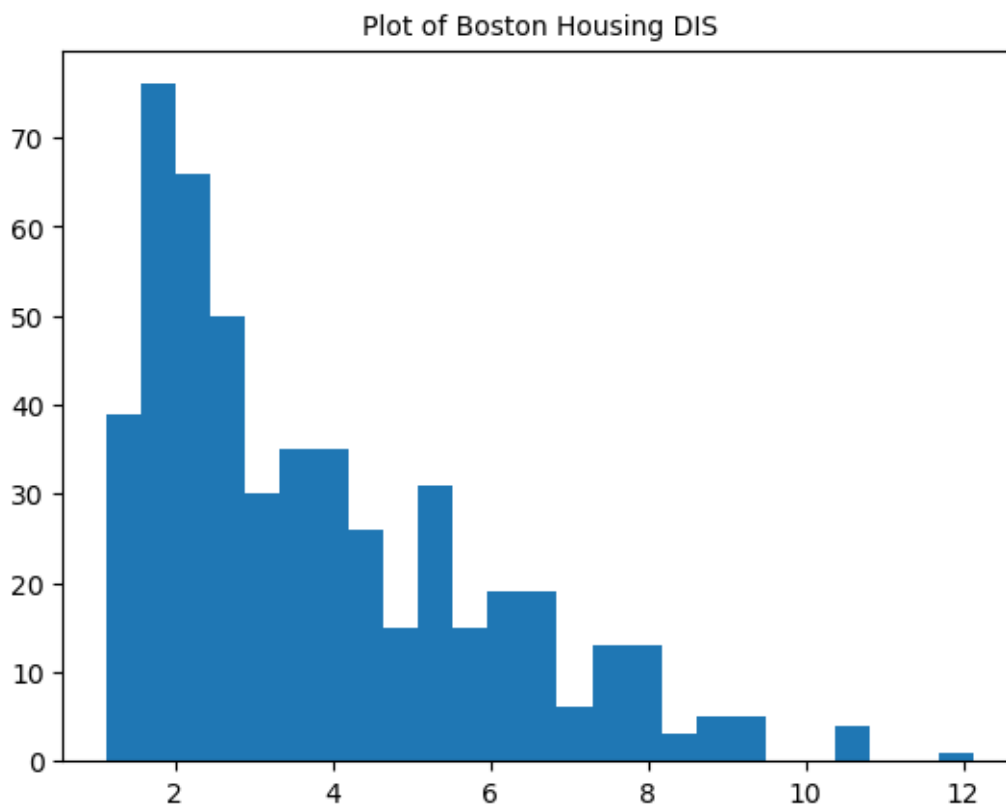


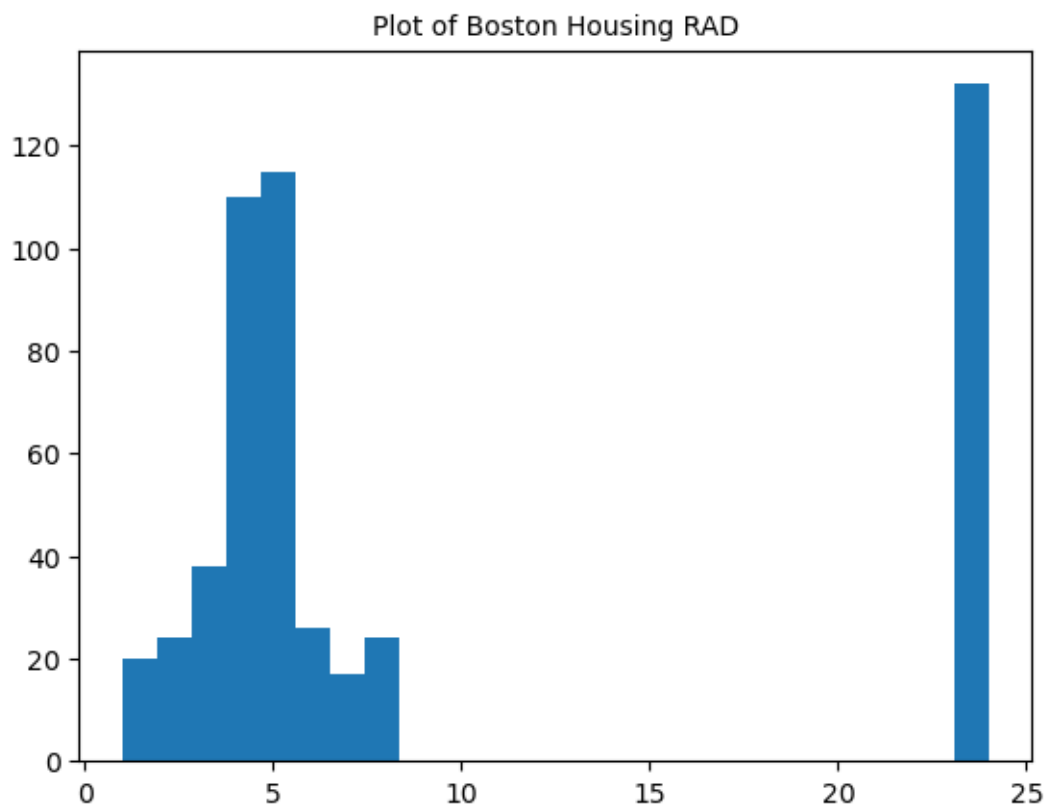


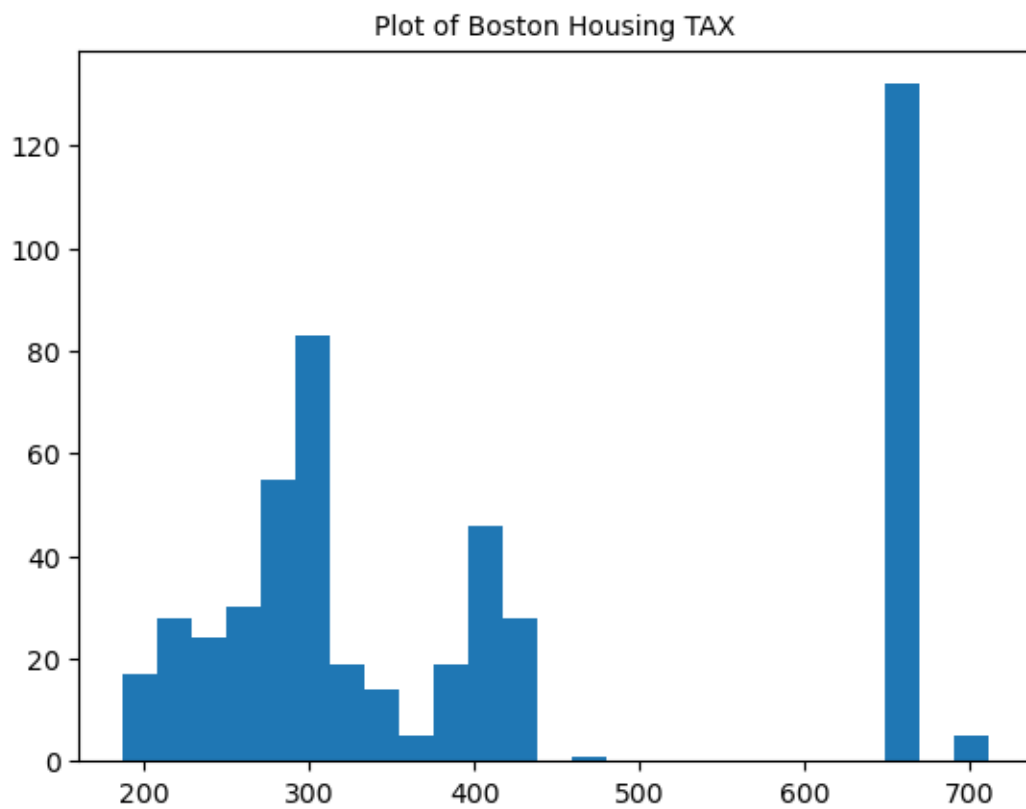


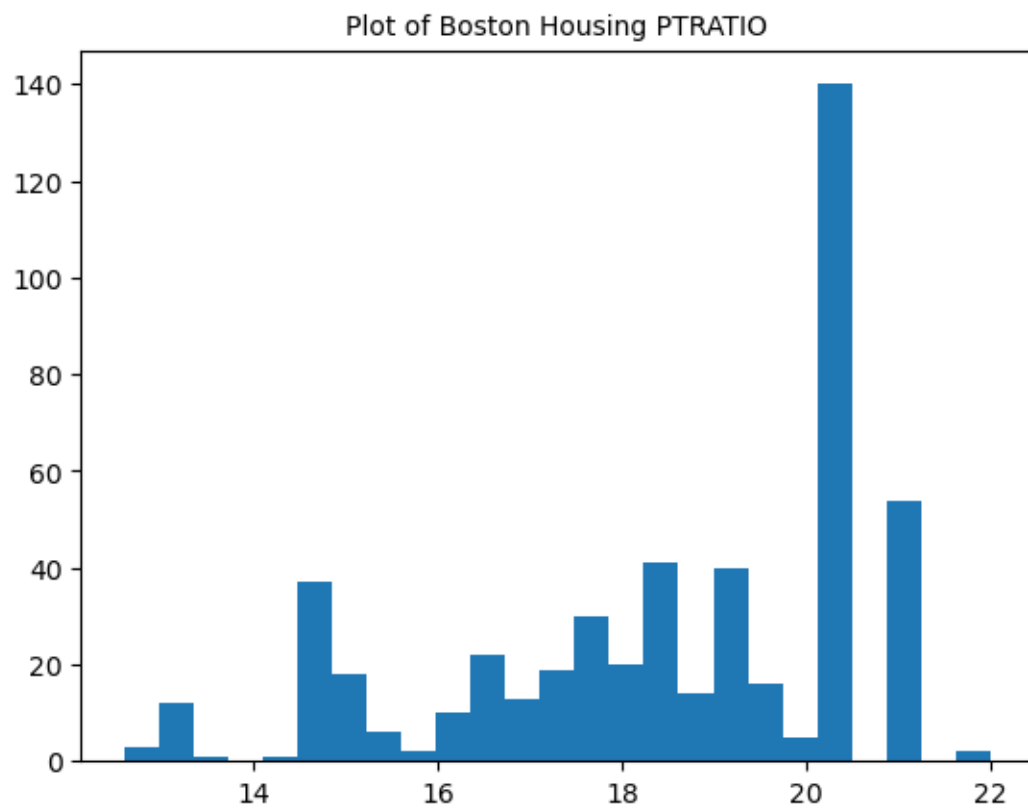


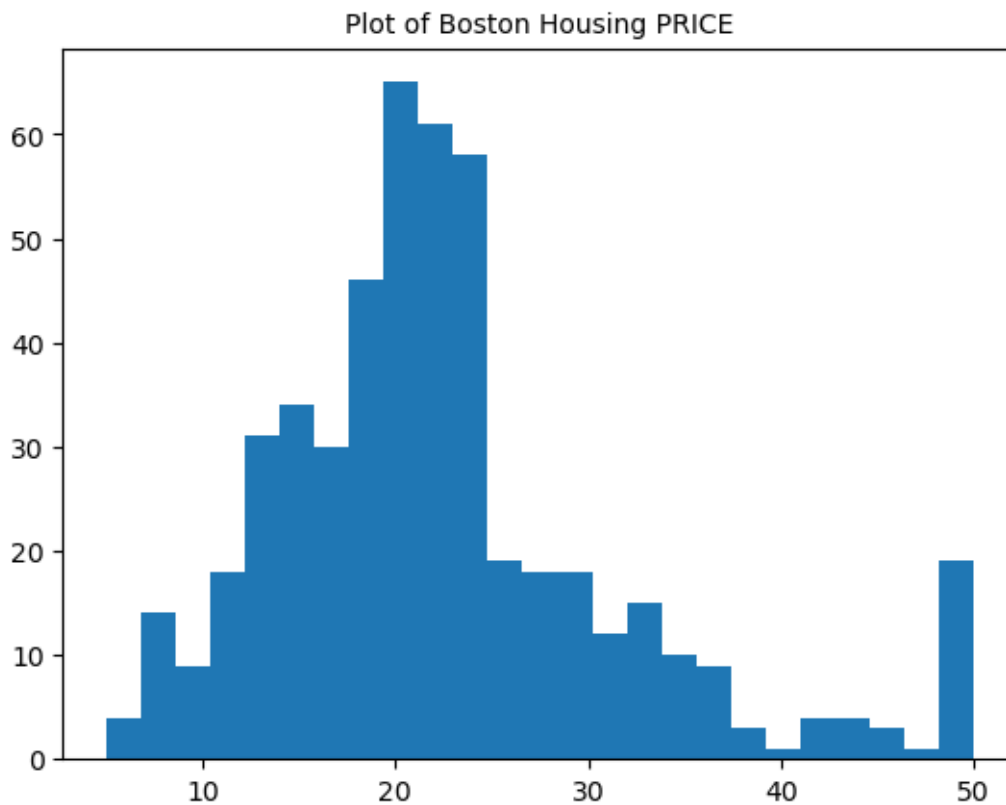






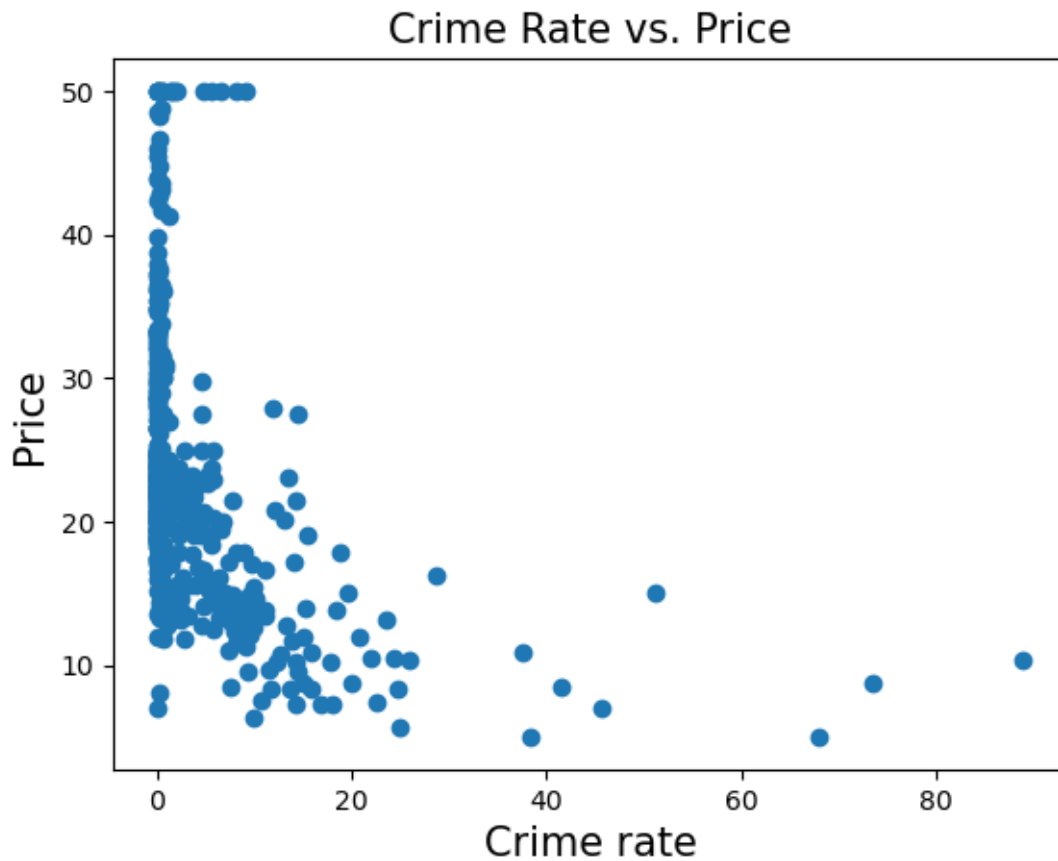






```
[101]: # Scatter plot of crime rate vs price using matplotlib

plt.scatter(bost_hs_sm_df['CRIM'],bost_hs_sm_df['PRICE'])
plt.title('Crime Rate vs. Price', fontsize=15)
plt.xlabel("Crime rate",fontsize=15)
plt.ylabel("Price",fontsize=15)
plt.show()
```

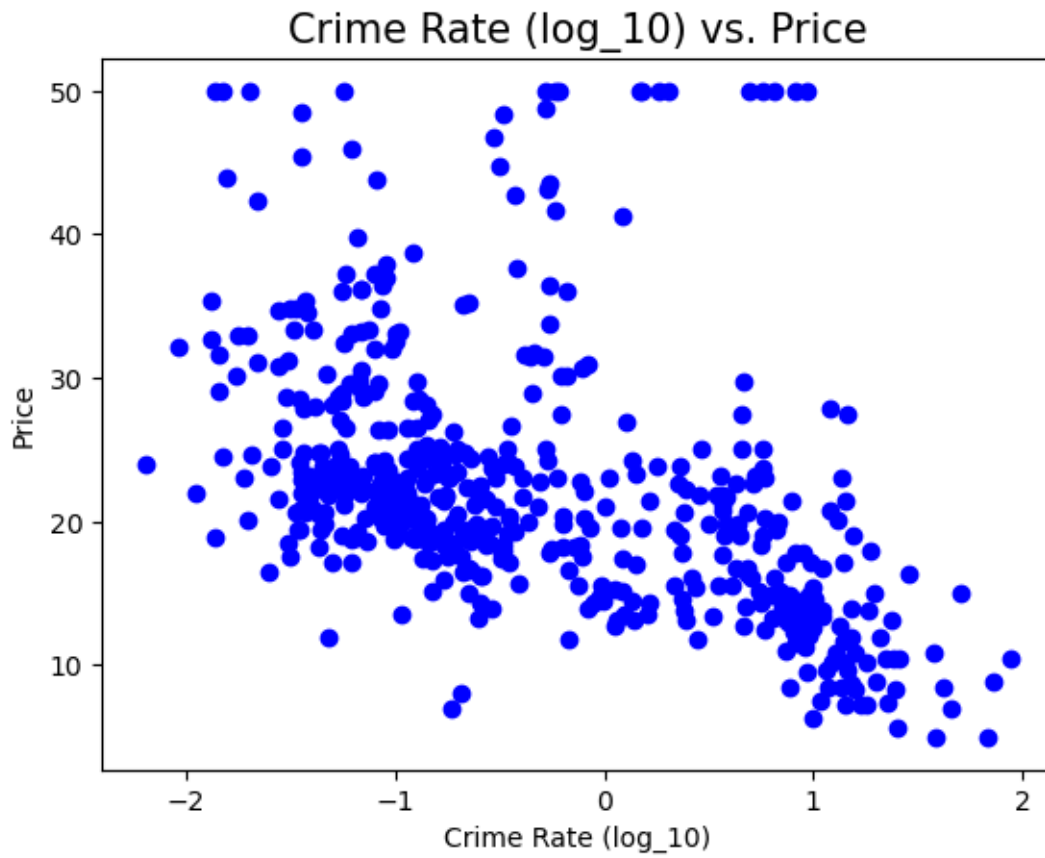


```
[102]: # Using Matplot Using log(10)

# Create a scatter plot
plt.scatter(np.log10(bost_hs_sm_df['CRIM']), bost_hs_sm_df['PRICE'],
            label='CRIM', color='blue')

# Add labels and title
plt.xlabel('Crime Rate (log_10)')
plt.ylabel('Price')
plt.title('Crime Rate (log_10) vs. Price', fontsize=15)

# Show the plot
plt.show()
```



```
[103]: # Calculate the mean of rooms using the Boston housing df.
```

```
bost_hs_sm_df['RM'].mean()
```

```
[103]: 6.284634387351779
```

```
[104]: # Calculate the median age using the Boston housing df.
```

```
bost_hs_sm_df['AGE'].median()
```

```
[104]: 77.5
```

```
[105]: # Calculate the mean distance to five Boston employment centers using the ↵  
↵Boston housing df.
```

```
bost_hs_sm_df['DIS'].mean()
```

```
[105]: 3.795042687747036
```

```
[106]: # create a boolean index to check if the values in the 'PRICE' column of a
        ↪ DataFrame where price is less than 20
```

```
house_less_20=bost_hs_sm_df['PRICE']<20
```

```
[107]: # Print the boolean index where the new boolean is only not equal to false
```

```
print((house_less_20 != "False") )
```

```
0      True
```

```
1      True
```

```
2      True
```

```
3      True
```

```
4      True
```

```
...
```

```
501    True
```

```
502    True
```

```
503    True
```

```
504    True
```

```
505    True
```

```
Name: PRICE, Length: 506, dtype: bool
```

```
[108]: # Find the total number of rows
```

```
num_rows = len(house_less_20)
```

```
print("Total number of rows:", num_rows)
```

```
Total number of rows: 506
```

```
[109]: # Calculate the mean of the house_less_20 variable and then expresses that mean
        ↪ as a percentage
```

```
# by multiplying it by 100. The result is stored in the variable perc.
```

```
perc=house_less_20.mean()*100
```

```
[110]: # using an f-string print a message then the calcuation of the percent of
        ↪ houses < 20,000.
```

```
print(f" % of houses with < 20,000: {perc} ")
```

```
% of houses with < 20,000: 41.50197628458498
```

0.1 Activity 6 Working with the Adult Income Dataset

```
[111]: # import libraries
```

```
import numpy as np
```

```
import pandas as pd
import matplotlib.pyplot as plt
```

```
[112]: #read the adult_income_data. dataset
```

```
adult_df=pd.read_csv("adult_income_data.csv")
```

```
[113]: # read the first 10 rows
```

```
adult_df.head(10)
```

```
[113]:    39      State-gov    77516    Bachelors    13      Never-married \
0  50    Self-emp-not-inc    83311    Bachelors    13      Married-civ-spouse
1  38      Private    215646      HS-grad     9      Divorced
2  53      Private    234721      11th      7      Married-civ-spouse
3  28      Private    338409    Bachelors    13      Married-civ-spouse
4  37      Private    284582    Masters    14      Married-civ-spouse
5  49      Private    160187      9th      5      Married-spouse-absent
6  52    Self-emp-not-inc    209642      HS-grad     9      Married-civ-spouse
7  31      Private    45781    Masters    14      Never-married
8  42      Private    159449    Bachelors    13      Married-civ-spouse
9  37      Private    280464    Some-college    10      Married-civ-spouse

      Adm-clerical    Not-in-family    Male    2174    0    40    United-States \
0    Exec-managerial      Husband    Male      0    0    13    United-States
1    Handlers-cleaners    Not-in-family    Male      0    0    40    United-States
2    Handlers-cleaners      Husband    Male      0    0    40    United-States
3    Prof-specialty      Wife    Female      0    0    40      Cuba
4    Exec-managerial      Wife    Female      0    0    40    United-States
5    Other-service    Not-in-family    Female      0    0    16      Jamaica
6    Exec-managerial      Husband    Male      0    0    45    United-States
7    Prof-specialty    Not-in-family    Female    14084    0    50    United-States
8    Exec-managerial      Husband    Male     5178    0    40    United-States
9    Exec-managerial      Husband    Male      0    0    80    United-States

    <=50K
0    <=50K
1    <=50K
2    <=50K
3    <=50K
4    <=50K
5    <=50K
6    >50K
7    >50K
8    >50K
9    >50K
```



```
[114]: # using the .txt file create a empty list, open the .txt file loop thru each
        ↪line in .txt seperated by ':'.
        # append to the headers list

headers = [] # initializes an empty list called headers to store the extracted
        ↪headers.
with open('adult_income_names.txt','r') as b:
    for line in b:
        b.readline()
        var=line.split(":")[0]
        headers.append(var)
```

```
[115]: # Output the headers list values
```

```
headers
```

```
[115]: ['age',
        'workclass',
        'fnlwgt',
        'education',
        'education-num',
        'marital-status',
        'occupation',
        'relationship',
        'sex',
        'capital-gain',
        'capital-loss',
        'hours-per-week',
        'native-country']
```

```
[116]: # append the 'Income' name to the response dataset
```

```
headers.append('Income')
```

```
[117]: # update the adult_df with the new names
```

```
adult_df=pd.read_csv("adult_income_data.csv",names=headers)
```

```
# Output the updated file for the new column
```

```
adult_df.head(10)
```

```
[117]:
```

	age	workclass	fnlwgt	education	education-num	\
0	39	State-gov	77516	Bachelors	13	
1	50	Self-emp-not-inc	83311	Bachelors	13	
2	38	Private	215646	HS-grad	9	
3	53	Private	234721	11th	7	

4	28	Private	338409	Bachelors	13
5	37	Private	284582	Masters	14
6	49	Private	160187	9th	5
7	52	Self-emp-not-inc	209642	HS-grad	9
8	31	Private	45781	Masters	14
9	42	Private	159449	Bachelors	13

	marital-status	occupation	relationship	sex \
0	Never-married	Adm-clerical	Not-in-family	Male
1	Married-civ-spouse	Exec-managerial	Husband	Male
2	Divorced	Handlers-cleaners	Not-in-family	Male
3	Married-civ-spouse	Handlers-cleaners	Husband	Male
4	Married-civ-spouse	Prof-specialty	Wife	Female
5	Married-civ-spouse	Exec-managerial	Wife	Female
6	Married-spouse-absent	Other-service	Not-in-family	Female
7	Married-civ-spouse	Exec-managerial	Husband	Male
8	Never-married	Prof-specialty	Not-in-family	Female
9	Married-civ-spouse	Exec-managerial	Husband	Male

	capital-gain	capital-loss	hours-per-week	native-country	Income
0	2174	0	40	United-States	<=50K
1	0	0	13	United-States	<=50K
2	0	0	40	United-States	<=50K
3	0	0	40	United-States	<=50K
4	0	0	40	Cuba	<=50K
5	0	0	40	United-States	<=50K
6	0	0	16	Jamaica	<=50K
7	0	0	45	United-States	>50K
8	14084	0	50	United-States	>50K
9	5178	0	40	United-States	>50K

```
[118]: # Missing data for adult_income_data DF
```

```
adult_df.isnull().sum()
```

```
[118]: age          0
workclass      0
fnlwgt         0
education      0
education-num  0
marital-status 0
occupation     0
relationship   0
sex            0
capital-gain   0
capital-loss   0
hours-per-week 0
```

```
native-country    0
Income            0
dtype: int64
```

```
[119]: # Create a subset df with only 'age', 'education', and 'occupation'.
```

```
adult_sub=adult_df[['age','education','occupation']]
```

```
[120]: # Head the new subset df to see changes
```

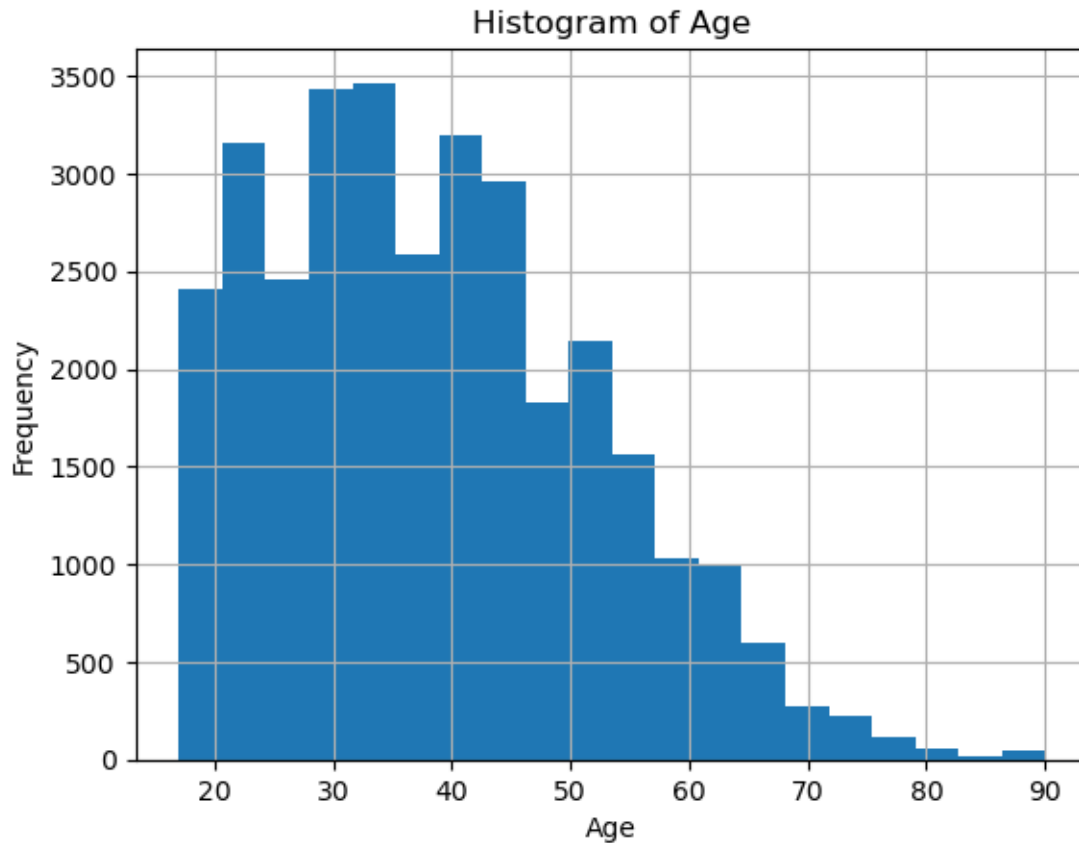
```
adult_sub.head(10)
```

```
[120]:
```

	age	education	occupation
0	39	Bachelors	Adm-clerical
1	50	Bachelors	Exec-managerial
2	38	HS-grad	Handlers-cleaners
3	53	11th	Handlers-cleaners
4	28	Bachelors	Prof-specialty
5	37	Masters	Exec-managerial
6	49	9th	Other-service
7	52	HS-grad	Exec-managerial
8	31	Masters	Prof-specialty
9	42	Bachelors	Exec-managerial

```
[121]: # Plot a histogram of age with a bin size of 20
```

```
adult_sub['age'].hist(bins=20)
plt.title('Histogram of Age')
plt.xlabel('Age')
plt.ylabel('Frequency')
plt.show()
```



[122]: *# Function to strip whitespace characters using the python strip method command*

```
def stp_wht_spc(c):
    return c.strip()
```

[123]: *# Use the apply method to apply the stp_wht_spc function to the adult_sub df*
↪ string columns

Education columns

```
adult_sub['ed_strip']=adult_df['education'].apply(stp_wht_spc)
adult_sub['education']=adult_sub['ed_strip']
adult_sub.drop(labels=['ed_strip'],axis=1,inplace=True)
```

Occupation column

```
adult_sub['oc_strip']=adult_df['occupation'].apply(stp_wht_spc)
adult_sub['occupation']=adult_sub['oc_strip']
adult_sub.drop(labels=['oc_strip'],axis=1,inplace=True)
```

```
/var/folders/yd/8fns7t4j1n9gv77y0dvs5zdc0000gn/T/ipykernel_27766/1840534740.py:5
: SettingWithCopyWarning:
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/user_guide/indexing.html#returning-a-view-versus-a-copy

```
adult_sub['ed_strip']=adult_df['education'].apply(stp_wht_spc)
/var/folders/yd/8fns7t4j1n9gv77y0dvs5zdc0000gn/T/ipykernel_27766/1840534740.py:6
: SettingWithCopyWarning:
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/user_guide/indexing.html#returning-a-view-versus-a-copy

```
adult_sub['education']=adult_sub['ed_strip']
/var/folders/yd/8fns7t4j1n9gv77y0dvs5zdc0000gn/T/ipykernel_27766/1840534740.py:7
: 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

```
adult_sub.drop(labels=['ed_strip'],axis=1,inplace=True)
/var/folders/yd/8fns7t4j1n9gv77y0dvs5zdc0000gn/T/ipykernel_27766/1840534740.py:1
1: SettingWithCopyWarning:
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/user_guide/indexing.html#returning-a-view-versus-a-copy

```
adult_sub['oc_strip']=adult_df['occupation'].apply(stp_wht_spc)
/var/folders/yd/8fns7t4j1n9gv77y0dvs5zdc0000gn/T/ipykernel_27766/1840534740.py:1
2: SettingWithCopyWarning:
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/user_guide/indexing.html#returning-a-view-versus-a-copy

```
adult_sub['occupation']=adult_sub['oc_strip']
/var/folders/yd/8fns7t4j1n9gv77y0dvs5zdc0000gn/T/ipykernel_27766/1840534740.py:1
3: 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

```
adult_sub.drop(labels=['oc_strip'],axis=1,inplace=True)
```

```
[124]: # Head the subset df
```

```
adult_sub.head(10)
```

```
[124]:
```

	age	education	occupation
0	39	Bachelors	Adm-clerical
1	50	Bachelors	Exec-managerial
2	38	HS-grad	Handlers-cleaners
3	53	11th	Handlers-cleaners
4	28	Bachelors	Prof-specialty
5	37	Masters	Exec-managerial
6	49	9th	Other-service
7	52	HS-grad	Exec-managerial
8	31	Masters	Prof-specialty
9	42	Bachelors	Exec-managerial

```
[125]: # Find the number of people who are aged between 30 and 50
```

```
adult_ages=adult_sub[(adult_sub['age']>=30) & (adult_sub['age']<=50)]
```

```
[126]: # Head the filter to show the results
```

```
adult_ages.head(10)
```

```
[126]:
```

	age	education	occupation
0	39	Bachelors	Adm-clerical
1	50	Bachelors	Exec-managerial
2	38	HS-grad	Handlers-cleaners
5	37	Masters	Exec-managerial
6	49	9th	Other-service
8	31	Masters	Prof-specialty
9	42	Bachelors	Exec-managerial
10	37	Some-college	Exec-managerial
11	30	Bachelors	Prof-specialty
13	32	Assoc-acdm	Sales

```
[127]: # Create a subset df with only 'age', 'education' to use for the group by below
```

```
adult_sub1=adult_sub[['education','age']]
```

```
[128]: # Create a subset df people who are aged between 30 and 50
```

```
adult_sub2=adult_sub1[(adult_sub1['age']>=30) & (adult_sub1['age']<=50)]
```

```
[129]: # head the new DF to see results
```

```
adult_sub2.head(10)
```

```
[129]:
```

	education	age
0	Bachelors	39
1	Bachelors	50
2	HS-grad	38
5	Masters	37
6	9th	49
8	Masters	31
9	Bachelors	42
10	Some-college	37
11	Bachelors	30
13	Assoc-acdm	32

```
[130]: # Group the record based on age and education to find the mean age
# Since the find the number between 30 and 50 was the question before this I
↳ assumed this was the ask
# If my assumption was wrong I can change, just let me know

adult_sub2.groupby(['education']).mean()
```

```
[130]:
```

education	age
10th	39.055921
11th	38.777188
12th	38.362319
1st-4th	39.384615
5th-6th	40.422222
7th-8th	40.617021
9th	38.894737
Assoc-acdm	38.825816
Assoc-voc	38.803851
Bachelors	39.179910
Doctorate	40.889831
HS-grad	39.049671
Masters	40.960360
Preschool	38.428571
Prof-school	40.492147
Some-college	39.224719

```
[131]: # Group by occupation and show summary stats for age

result=adult_sub.groupby('occupation').describe()['age']

# Sort by mean age in descending order
result_sorted = result.sort_values(by='mean', ascending=False)

# Display the sorted result
```

```
print(result_sorted)
```

	count	mean	std	min	25%	50%	75%	max
occupation								
Exec-managerial	4066.0	42.169208	11.974548	17.0	33.0	41.0	50.0	90.0
Priv-house-serv	149.0	41.724832	18.633688	17.0	24.0	40.0	57.0	81.0
Farming-fishing	994.0	41.211268	15.070283	17.0	29.0	39.0	52.0	90.0
?	1843.0	40.882800	20.336350	17.0	21.0	35.0	61.0	90.0
Prof-specialty	4140.0	40.517633	12.016676	17.0	31.0	40.0	48.0	90.0
Transport-moving	1597.0	40.197871	12.450792	17.0	30.0	39.0	49.0	90.0
Craft-repair	4099.0	39.031471	11.606436	17.0	30.0	38.0	47.0	90.0
Protective-serv	649.0	38.953775	12.822062	17.0	29.0	36.0	47.0	90.0
Machine-op-inspct	2002.0	37.715285	12.068266	17.0	28.0	36.0	46.0	90.0
Sales	3650.0	37.353973	14.186352	17.0	25.0	35.0	47.0	90.0
Tech-support	928.0	37.022629	11.316594	17.0	28.0	36.0	44.0	73.0
Adm-clerical	3770.0	36.964456	13.362998	17.0	26.0	35.0	46.0	90.0
Other-service	3295.0	34.949621	14.521508	17.0	22.0	32.0	45.0	90.0
Handlers-cleaners	1370.0	32.165693	12.372635	17.0	23.0	29.0	39.0	90.0
Armed-Forces	9.0	30.222222	8.089774	23.0	24.0	29.0	34.0	46.0

The profession with the oldest workers on average is “Exec-managerial”. The profession that has the largest share above 75% is the “?” profession.

[132]: *# Use subset and groupby to find the outliers*

```
out_stat=adult_sub.groupby('occupation').describe()['age']
```

```
# Sort by count in descending order
```

```
out_sorted = out_stat.sort_values(by='count', ascending=False)
```

```
# Display the sorted result
```

```
print(out_sorted)
```

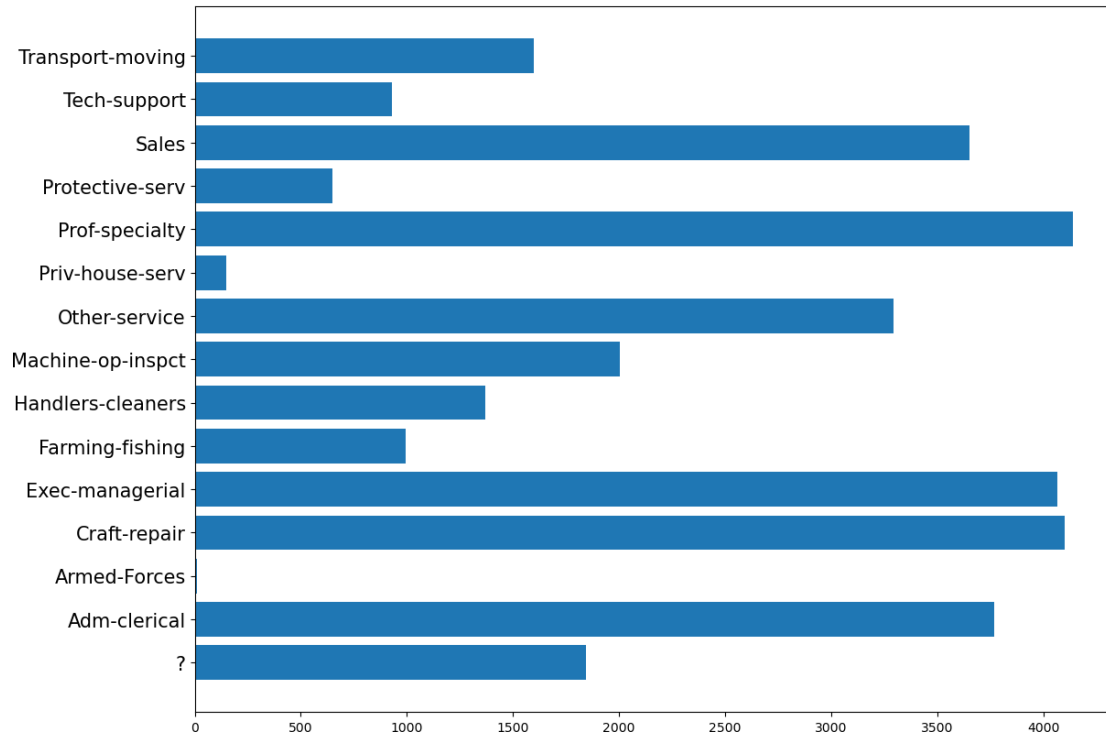
	count	mean	std	min	25%	50%	75%	max
occupation								
Prof-specialty	4140.0	40.517633	12.016676	17.0	31.0	40.0	48.0	90.0
Craft-repair	4099.0	39.031471	11.606436	17.0	30.0	38.0	47.0	90.0
Exec-managerial	4066.0	42.169208	11.974548	17.0	33.0	41.0	50.0	90.0
Adm-clerical	3770.0	36.964456	13.362998	17.0	26.0	35.0	46.0	90.0
Sales	3650.0	37.353973	14.186352	17.0	25.0	35.0	47.0	90.0
Other-service	3295.0	34.949621	14.521508	17.0	22.0	32.0	45.0	90.0
Machine-op-inspct	2002.0	37.715285	12.068266	17.0	28.0	36.0	46.0	90.0
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Handlers-cleaners	1370.0	32.165693	12.372635	17.0	23.0	29.0	39.0	90.0
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Priv-house-serv	149.0	41.724832	18.633688	17.0	24.0	40.0	57.0	81.0

Armed-Forces 9.0 30.222222 8.089774 23.0 24.0 29.0 34.0 46.0

The Armed-Forces occupation appears to be an outlier.

```
[133]: # Plot the values of the bar chart

plt.figure(figsize=(13,10))
plt.barh(y=out_stat.index,width=out_stat['count'])
plt.yticks(fontsize=15)
plt.show()
```



```
[134]: # Merge data using common keys.
# create new df by subset of columns from orig df then sample 10 rows and seed
↳ the rand # gen

merge_adult=adult_df[['age','workclass', 'occupation']].
↳ sample(10,random_state=101)
```

```
[135]: # head the new df

merge_adult.head()
```

```
[135]:      age workclass      occupation
22357   51   Private  Machine-op-inspct
```

26009	19	Private	Sales
20734	40	Private	Exec-managerial
17695	17	Private	Handlers-cleaners
27908	61	Private	Craft-repair

```
[136]: # Merge data using common keys.
# create new df by subset of columns from orig df then sample 10 rows and seed
↳ the rand # gen

merge2_adult=adult_df[['education','sex','occupation']].
↳ sample(10,random_state=101)
```

```
[137]: # head the new df

merge2_adult.head()
```

	education	sex	occupation
22357	HS-grad	Female	Machine-op-inspct
26009	11th	Male	Sales
20734	HS-grad	Male	Exec-managerial
17695	10th	Male	Handlers-cleaners
27908	7th-8th	Male	Craft-repair

```
[138]: # Merge data using common keys.
# create new df merge from two new dfs on occupation and an inner join dropping
↳ the dups.

final_adult=pd.merge(merge_adult, merge2_adult, on ='occupation',how ='inner').
↳ drop_duplicates()
```

```
[139]: # results

final_adult
```

	age	workclass	occupation	education	sex
0	51	Private	Machine-op-inspct	HS-grad	Female
1	19	Private	Sales	11th	Male
2	19	Private	Sales	HS-grad	Male
3	22	Private	Sales	11th	Male
4	22	Private	Sales	HS-grad	Male
5	40	Private	Exec-managerial	HS-grad	Male
6	17	Private	Handlers-cleaners	10th	Male
7	61	Private	Craft-repair	7th-8th	Male
8	58	?	?	Some-college	Female
10	26	?	?	Some-college	Female
12	37	Local-gov	Other-service	HS-grad	Male
13	22	Private	Adm-clerical	Assoc-voc	Female

0.2 Exercise 3

Create a series and practice basic arithmetic steps

```
[140]: # Create a series and practice basic arithmetic steps

import pandas as pd #import library

# Creating Series 1
series1_data = [7.3, -2.5, 3.4, 1.5]
series1_index = ['a', 'c', 'd', 'e']
series1 = pd.Series(series1_data, index=series1_index) # create the pandas
↳series

# Creating Series 2
series2_data = [-2.1, 3.6, -1.5, 4, 3.1]
series2_index = ['a', 'c', 'e', 'f', 'g']
series2 = pd.Series(series2_data, index=series2_index) # create the pandas
↳series

# Adding Series 1 and Series 2
addition_result = series1 + series2
print("Addition Result:")
print(addition_result)

# Subtracting Series 1 from Series 2
subtraction_result = series2 - series1
print("\nSubtraction Result:")
print(subtraction_result)
```

Addition Result:

```
a    5.2
c    1.1
d    NaN
e    0.0
f    NaN
g    NaN
dtype: float64
```

Subtraction Result:

```
a   -9.4
c    6.1
d    NaN
e   -3.0
f    NaN
g    NaN
dtype: float64
```