lab3

November 29, 2019

```
[2]: import pandas as pd
    import matplotlib.pyplot as plt
    from IPython.display import display
    import seaborn as sns
[3]: pd_city = pd.read_csv("./city_data.csv")
    pd_city.head(5)
[3]:
               city
                     driver_count
                                     type
        Richardfort
                                38
                                   Urban
    1 Williamsstad
                                59
                                   Urban
    2
       Port Angela
                                67
                                   Urban
    3
         Rodneyfort
                                34
                                   Urban
        West Robert
                                39
                                   Urban
[4]: pd_city.dtypes
[4]: city
                    object
    driver_count
                     int64
    type
                    object
    dtype: object
[5]: pd_ride = pd.read_csv("./ride_data.csv")
    pd_ride.head(5)
[5]:
                     city
                                           date
                                                  fare
                                                               ride_id
    0 Lake Jonathanshire 2018-01-14 10:14:22
                                                 13.83 5739410935873
    1 South Michelleport 2018-03-04 18:24:09
                                                 30.24 2343912425577
    2 Port Samanthamouth 2018-02-24 04:29:00
                                                 33.44
                                                        2005065760003
    3
               Rodneyfort 2018-02-10 23:22:03
                                                 23.44 5149245426178
    4
               South Jack 2018-03-06 04:28:35
                                                 34.58 3908451377344
[6]: pd_ride.dtypes
[6]: city
                object
                object
    date
    fare
               float64
                 int64
    ride_id
    dtype: object
[7]: pd_join = pd_city.join(pd_ride.set_index("city"),on="city")
    pd_join['type'].unique()
```

```
[7]: array(['Urban', 'Suburban', 'Rural'], dtype=object)
[8]: pd_join[pd_join['city'] == 'Amandaburgh']
[8]:
                                                                   fare \
               city
                      driver_count
                                                            date
                                      type
                                            2018-03-05 02:15:38
                                                                  26.28
    61
        Amandaburgh
                                12
                                    Urban
        Amandaburgh
                                12
                                            2018-02-24 23:10:49
                                                                  43.66
    61
                                    Urban
        Amandaburgh
                                12
                                    Urban
                                            2018-02-10 20:42:46
                                                                  36.17
    61
        Amandaburgh
                                12
                                    Urban
                                            2018-01-11 02:22:07
                                                                  29.24
        Amandaburgh
                                12 Urban
                                                                   9.26
    61
                                            2018-01-21 04:12:54
    61
        Amandaburgh
                                12
                                    Urban
                                            2018-04-19 16:30:12
                                                                   6.27
                                            2018-03-20 07:40:33
                                                                  27.45
    61
        Amandaburgh
                                12
                                    Urban
                                    Urban
                                            2018-04-01 09:24:21
                                                                  24.29
    61
        Amandaburgh
                                12
                                                                  16.27
        Amandaburgh
                                12
                                    Urban
                                            2018-04-20 02:16:07
    61
                                    Urban
                                                                  13.88
    61
        Amandaburgh
                                12
                                            2018-03-13 12:52:31
                                                                  42.52
    61
        Amandaburgh
                                12
                                    Urban
                                            2018-04-22 21:34:17
        Amandaburgh
                                12 Urban
                                            2018-02-06 10:02:30
                                                                  11.93
    61
    61
        Amandaburgh
                                12
                                    Urban
                                            2018-04-24 08:02:27
                                                                  14.55
        Amandaburgh
                                12 Urban
                                            2018-04-05 10:22:33
                                                                  25.55
    61
        Amandaburgh
                                12 Urban
                                            2018-03-07 02:26:33
                                                                  18.76
    61
    61
        Amandaburgh
                                12 Urban
                                            2018-01-02 09:57:04
                                                                  33.06
    61
        Amandaburgh
                                12
                                    Urban
                                            2018-01-13 16:04:10
                                                                  23.35
        Amandaburgh
    61
                                    Urban
                                            2018-01-29 23:28:12
                                                                  41.06
              ride_id
    61
         906850928986
    61
        6573820412437
        6455620849753
    61
    61
        7279902884763
    61
        5528427024492
        4400632718421
        3701008274871
    61
        1995462170530
    61
        3513123734716
    61
        6222134922674
    61
        1901157522591
    61
        7550325158038
    61
        7836117055007
        8581415267582
        3419454549176
        6330658179518
    61
        9975084532253
    61
        4296858665195
    61
[9]: pd_group = pd.DataFrame(data=list(pd_join.groupby(by='city').
     →mean()['fare']),columns=['Average Fare'])
    type(pd_join.groupby(by='city').mean()['fare'])
```

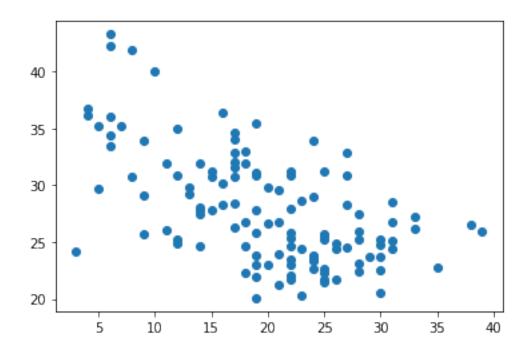
[9]:		Average Fare	Number of Drivers	Number of Rides	
	city	<u>o</u>			
	Amandaburgh	24.641667	18	18	
	Barajasview	25.332273	22	22	
	Barronchester	36.422500	16	16	
	Bethanyland	32.956111	18	18	
	Bradshawfurt	40.064000	10	10	
	Brandonfort	35.437368	19	19	
	Carriemouth	28.314444	27	27	
	Christopherfurt	24.501852	27	27	
	Colemanland	30.894545	22	22	
	Davidfurt	31.995882	17	17	
	Deanville	25.842632	19	19	
	East Aaronbury	25.661111	9	9	
	East Danielview	31.560588	17	17	
	East Kaylahaven	23.757931	29	29	
	East Kentstad	29.823077	13	13	
	East Marymouth	30.835185	27	27	
	Erikaland	24.906667	12	12	
	Garzaport	24.123333	3	3	
	Grahamburgh	25.221200	25	25	
	Grayville	27.763333	15	15	
	Harringtonfort	33.470000	6	6	
	Huntermouth	28.993750	24	24	
	Hurleymouth	25.891429	28	28	
	Jerryton	25.649200	25	25	
	Jessicaport	36.013333	6	6	
	Johnton	26.785714	21	21	
	Joneschester	22.289600	25	25	
	Josephside	32.858148	27	27	
	Justinberg	23.694333	30	30	
	Karenberg	26.340000	17	17	
	 South Evanton	26.726129	31	31	
	South Jack	22.965263	19	19	
	South Jennifer	35.264286	7	7	
	South Karenland	26.535526	38	38	
	South Latoya	20.093158	19	19	

South Marychester	41.870000	8	8
South Michelleport	24.451613	31	31
South Phillip	28.571290	31	31
South Saramouth	36.160000	4	4
South Teresa	31.220455	22	22
Taylorhaven	42.263333	6	6
Valentineton	24.636364	22	22
Veronicaberg	32.828235	17	17
Victoriaport	27.780000	14	14
West Angela	25.990000	39	39
West Anthony	24.736667	30	30
West Christopherberg	24.421154	26	26
West Ericstad	22.347222	18	18
West Gabriel	20.346087	23	23
West Hannah	29.547619	21	21
West Heather	33.890000	9	9
West Heidi	23.133929	28	28
West Josephberg	21.720385	26	26
West Kimmouth	29.871500	20	20
West Patrickchester	28.233125	16	16
West Robert	25.123871	31	31
West Samuelburgh	21.767600	25	25
Williamsonville	31.875000	14	14
Williamsstad	24.362174	23	23
Williamsview	26.599000	20	20

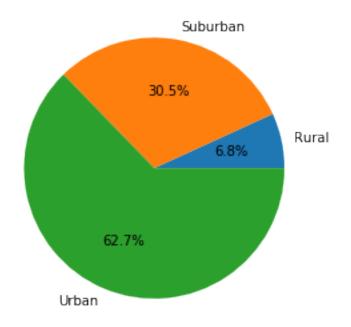
[120 rows x 3 columns]

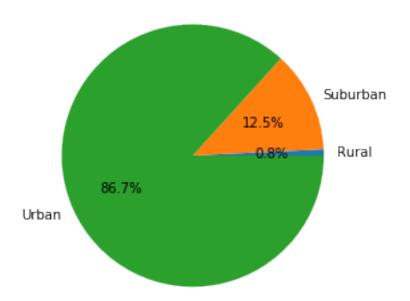
[10]: plt.scatter(list(pd_group['Number of Drivers']), list(pd_group['Average Fare']))

[10]: <matplotlib.collections.PathCollection at 0x268db365710>



```
[11]: pd_join.groupby(by='type').sum()
[11]:
               driver_count
                                 fare
                                                ride_id
     type
     Rural
                        537
                              4327.93
                                        580968240341287
     Suburban
                       8570
                             19356.33
                                       3106884522576766
    Urban
                      59602
                             39854.38
                                       7919412664056093
[12]: fig,axe1 = plt.subplots()
     total_fare_index = pd_join.groupby(by='type').sum()['fare'].index
     total_fare_value = list(pd_join.groupby(by='type').sum()['fare'])
     axe1.pie(total_fare_value,labels=total_fare_index,autopct='%1.1f%%')
     axe1.axis('equal')
     total_ride_index = pd_join.groupby(by='type').sum()['driver_count'].index
     total_ride_value = list(pd_join.groupby(by='type').sum()['driver_count'])
     fig,axe2 = plt.subplots()
     axe2.pie(total_ride_value,labels=total_ride_index,autopct='%1.1f%%')
     axe2.axis('equal')
     plt.show()
```



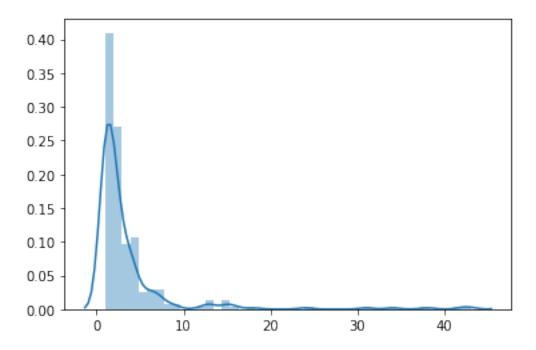


```
[30]: #t = sns.load_dataset("titanic") why "" not working
t = sns.load_dataset('titanic')
#len(t.index)
t.head(10)
```

[30]:	survived	pclass	sex	age	sibsp	parch	fare	embarked	class	\
C	0	3	male	22.0	1	0	7.2500	S	Third	
1	l 1	1	female	38.0	1	0	71.2833	C	First	

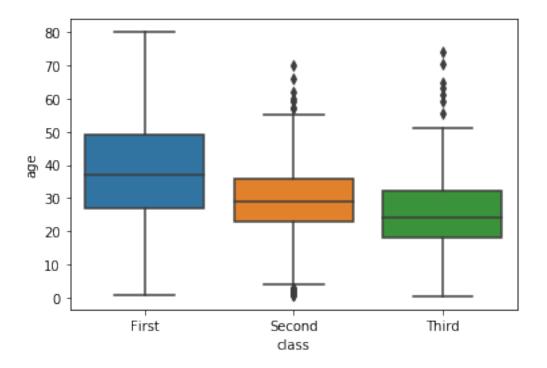
```
2
                1
                           female
                                    26.0
                                               0
                                                           7.9250
                                                                          S
                                                                              Third
     3
                1
                                    35.0
                                                         53.1000
                                                                          S
                        1
                           female
                                               1
                                                                              First
     4
                                                                          S
                0
                        3
                              male
                                    35.0
                                               0
                                                           8.0500
                                                                              Third
     5
                0
                        3
                              male
                                               0
                                                                          Q
                                                                              Third
                                     NaN
                                                           8.4583
     6
                0
                        1
                              male 54.0
                                               0
                                                          51.8625
                                                                          S
                                                                              First
     7
                0
                        3
                                               3
                                                                          S
                                                                              Third
                              male
                                     2.0
                                                       1
                                                          21.0750
                                                                              Third
     8
                1
                        3
                           female 27.0
                                               0
                                                       2
                                                          11.1333
                                                                          S
     9
                1
                        2
                            female
                                               1
                                                          30.0708
                                                                             Second
                                    14.0
                adult_male deck
                                  embark_town alive
                                                       alone
     0
                      True
                             NaN
                                  Southampton
                                                       False
          man
                                                  no
     1
        woman
                     False
                               C
                                    Cherbourg
                                                       False
                                                 yes
     2
        woman
                     False
                            {\tt NaN}
                                  Southampton
                                                 yes
                                                        True
     3
        woman
                     False
                               C
                                  Southampton
                                                 yes
                                                       False
     4
                      True
                                  Southampton
          man
                             NaN
                                                        True
                                                  no
     5
          man
                      True
                             NaN
                                   Queenstown
                                                  no
                                                        True
                                  Southampton
     6
          man
                      True
                               Ε
                                                        True
                                                  no
     7
        child
                     False
                                  Southampton
                                                       False
                             NaN
                                                  no
        woman
                     False
                             NaN
                                  Southampton
                                                 yes False
        child
                     False
                             NaN
                                    Cherbourg
                                                 yes False
[49]: fare_pd = pd.DataFrame()
     fare_pd['fare'] = t.groupby('fare').indices
     fare_pd['count'] = t.groupby('fare').count()['sex']
     fare_pd.set_index('fare',inplace=True,drop=True)
     #type(fare_pd)
     #fare_pd
     #t.groupby('fare').count().head(3)
     sns.distplot(fare_pd)
```

[49]: <matplotlib.axes._subplots.AxesSubplot at 0x268e653a978>



[51]: sns.boxplot(x='class',y='age',data=t)

[51]: <matplotlib.axes._subplots.AxesSubplot at 0x268e6679c18>



```
[76]: male_pd = t[t['sex']=='male']
  male_pd.head(5)
  male_pd.dropna(inplace=True)
  male_pd['age']
  #male_pd.groupby('age').count()
  #male_pd['age']<=80
  sns.distplot(male_pd['age'])</pre>
```

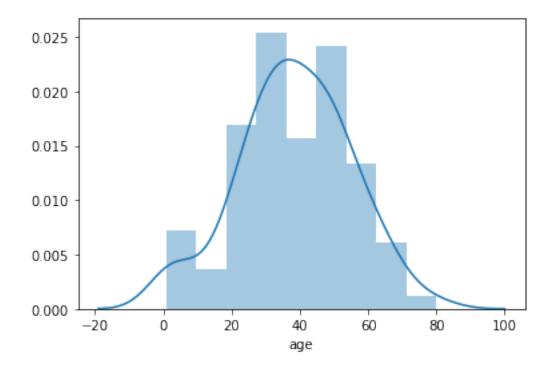
E:\software\anaconda\lib\site-packages\ipykernel_launcher.py:3:
SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

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

This is separate from the ipykernel package so we can avoid doing imports until

[76]: <matplotlib.axes._subplots.AxesSubplot at 0x268e6981c50>



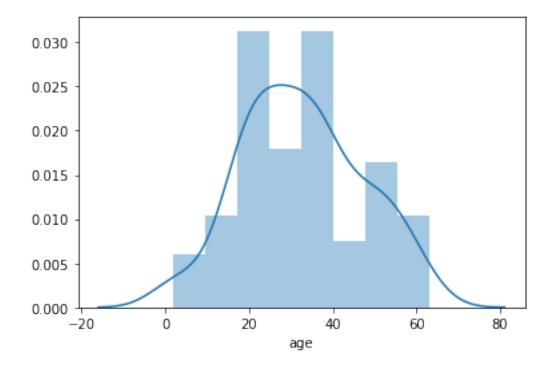
```
[79]: female_pd = t[t['sex'] == 'female']
female_pd.dropna(inplace=True)
#female_pd['age']
#male_pd.groupby('age').count()
#male_pd['age'] <= 80
sns.distplot(female_pd['age'])
```

E:\software\anaconda\lib\site-packages\ipykernel_launcher.py:2: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

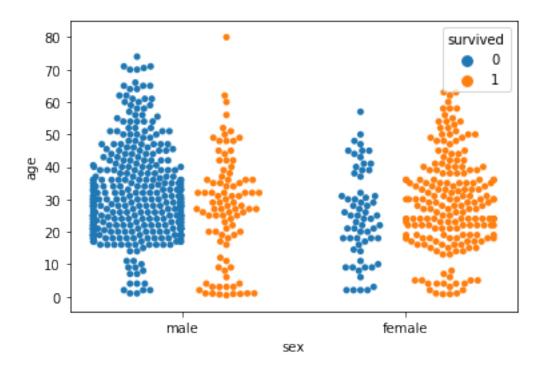
See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.html#indexing-view-versus-copy

[79]: <matplotlib.axes._subplots.AxesSubplot at 0x268e6a06438>



[82]: sns.swarmplot(x='sex',y='age',hue='survived',dodge=True,data=t)

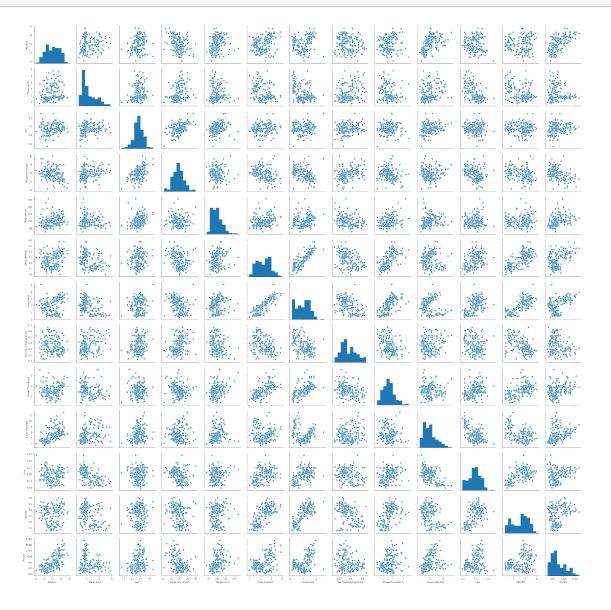
[82]: <matplotlib.axes._subplots.AxesSubplot at 0x268e6beb400>



```
[78]: wine_pd = pd.read_csv("./wine_data.csv")
     #wine_pd.head()
     wine_pd['Label'].unique()
     labels = wine_pd['Label']
     labels
     wine_pd = wine_pd.drop(axis=1,columns='Label')
     wine_pd.head(5)
[78]:
        Alcohol Malic acid
                              Ash
                                    Alcalinity of ash Magnesium
                                                                   Total phenols \
          14.23
                       1.71
                             2.43
                                                 15.6
                                                                            2.80
                                                              127
     0
          13.20
                                                 11.2
     1
                       1.78 2.14
                                                              100
                                                                            2.65
          13.16
                                                                            2.80
     2
                       2.36
                             2.67
                                                 18.6
                                                              101
          14.37
                                                 16.8
                                                                            3.85
     3
                       1.95
                             2.50
                                                              113
     4
          13.24
                       2.59 2.87
                                                 21.0
                                                              118
                                                                            2.80
        Flavanoids Nonflavanoid phenols Proanthocyanins Color intensity
                                                                               Hue
     0
              3.06
                                     0.28
                                                      2.29
                                                                        5.64 1.04
              2.76
                                     0.26
                                                       1.28
     1
                                                                        4.38 1.05
     2
              3.24
                                     0.30
                                                      2.81
                                                                        5.68 1.03
                                     0.24
     3
              3.49
                                                      2.18
                                                                        7.80 0.86
              2.69
                                     0.39
                                                      1.82
                                                                        4.32 1.04
        OD280 Proline
         3.92
     0
                  1065
         3.40
                  1050
     1
     2
         3.17
                  1185
```

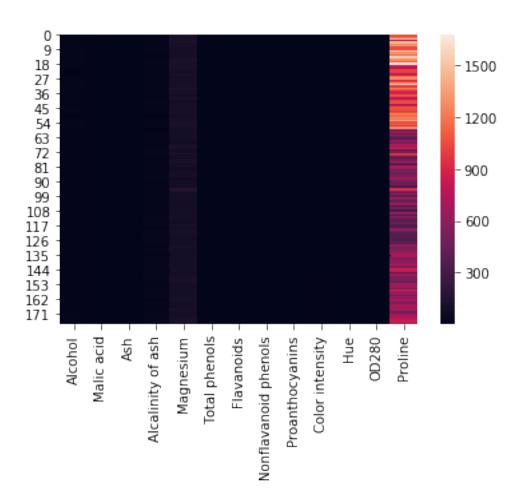
3 3.45 1480 4 2.93 735

[14]: sns_plot = sns.pairplot(wine_pd,diag_kind="hist")



[15]: #plt.subplots(figsize=(100,100))
sns.heatmap(wine_pd)

[15]: <matplotlib.axes._subplots.AxesSubplot at 0x268e1e7a080>



```
[16]: from sklearn import preprocessing
     from sklearn.cluster import KMeans
[17]: standardScaler = preprocessing.StandardScaler()
     standardScaler.fit(wine_pd)
     X_scaled_array = standardScaler.transform(wine_pd)
     X scaled array
     \#normalizedData = pd.DataFrame(X_scaled_array, columns = wine_pd.columns)
     #normalizedData.head(5)
     #len(normalizedData.index)
[17]: array([[ 1.51861254, -0.5622498 ,
                                       0.23205254, ...,
                                                          0.36217728,
             1.84791957, 1.01300893],
            [0.24628963, -0.49941338, -0.82799632, ..., 0.40605066,
             1.1134493 , 0.96524152],
            [0.19687903, 0.02123125, 1.10933436, ..., 0.31830389,
             0.78858745, 1.39514818],
            [ 0.33275817, 1.74474449, -0.38935541, ..., -1.61212515,
            -1.48544548, 0.28057537],
```

```
[0.20923168, 0.22769377, 0.01273209, ..., -1.56825176,
            -1.40069891, 0.29649784],
            [1.39508604, 1.58316512, 1.36520822, ..., -1.52437837,
            -1.42894777, -0.59516041]])
[18]: kMeansClustering = KMeans(n_clusters = 3)
     res = kMeansClustering.fit_predict(normalizedData)
     res
            NameError
                                                      Traceback (most recent call_
     →last)
            <ipython-input-18-6005a674d9fd> in <module>
              1 kMeansClustering = KMeans(n_clusters = 3)
        ----> 2 res = kMeansClustering.fit_predict(normalizedData)
              3 res
            NameError: name 'normalizedData' is not defined
 []: normalizedData['cluster'] = res
     normalizedData.head(3)
```

[]: sns_plot = sns.pairplot(normalizedData, hue = "cluster", diag_kind="hist")