Link Github: paulbboone/DataMining ThucHanh (github.com)

HOMEWORK LAB 02

In [2]:	<pre>%matplotlib inline import numpy as np import pandas as pd df =pd.read_csv("PastHires.csv") df.head()</pre>								
	0	10	Υ						
	1	0	N	0	BS	Y	Υ	Υ	
	2	7	N Y		BS MS	N Y	N N	N	
	3							Υ	
	4	20	N	2	PhD	Υ	N	N	
	[n [3]:	df.h	ead(10)						
Out[3]:	Y	ears Experience	Employed?	Previous employers	Level of Education	Top-tier school	Interned	Hired	
	0	10	Υ	4	BS	N	N	Υ	
	1	0	N	0	BS	Y	Υ	Υ	
	2	7	N	6	BS	N	N	N	
	3	2	Υ	1	MS	Υ	N	Υ	
	4	20	N	2	PhD	Y	N.	N	
	5	0	N	0	PhD	Y	Υ	Υ	
	6	5	Υ	2	MS	N	Υ	Υ	
	7	3	N	1	BS	N	Y	Υ	
	8	15	Υ	5	BS	N	N	Υ	
	9	0	N	0	BS	N	N	N	
[n [4]:	df.t	ail(4)							
Out[4]:	8.9	Years Experience	e Employed?	Previous employers	Level of Education	Top-tier school	Interned	Hired	
	9	(0 1	1 0	BS	N	l N	N	
	10		1 1	. 1	PhD	Y	N		
	11		4)			N	I Y	Υ	
	12	(0 1		PhD	. Y	N	Y	
[n [5]:	df.sl	nape							
Out[5]:	(13,	7)							
[n [6]:	20.0	170							

```
ourfol. AT
In [7]: len(df)
Out[7]: 13
In [9]: df.columns
dtype='object')
In [12]: df['Hired']
Out[12]: 0
          N
Y
N
Y
      3 4 5
      6 7 8
      10
11
       12
      Name: Hired, dtype: object
In [13]: df['Hired'][:5]
Out[13]: 0 Y
      2
      2 N
3 Y
4 N
      Name: Hired, dtype: object
In [17]: df['Hired'][5]
Out[17]: 'Y'
In [18]: df[['Years Experience','Hired']]
Out[18]: Years Experience Hired
       0 10 Y
       1
                  0
       2 7 N
        3
                  2
```

```
In [19]: df[['Years Experience', 'Hired']][:5]
Out[19]: Years Experience Hired
           10 Y
        0
        1
                    0
                  7 N
        2
        3
                    2
              20 N
In [20]: df.sort_values(['Years Experience'])
Out [20]: Years Experience Employed? Previous employers Level of Education Top-tier school Interned Hired
                        N
        1
                    0
                                          0
         5
                     0
                             N
                                           0
                                                      PhD
                                                    BS
        9
                   0 N
                                          0
                                                                  N
                                                                      N N
        12
                     0
                             N
                                           0
                                                      PhD
                                                                         N
        10
                                                                   Y N N
                                                      PhD
                     3
                                                       BS
                                                                  N
        11
                     4
                                                                  N
                                                       BS
                                           2
         6
                     5
                                                       MS
                                                                  N
                                           6
         2
                             N
                                                       BS
                                                                   N
         0
                    10
                             Υ
                                           4
                                                       BS
         8
                     15
                                           5
                                                       BS
                                                                  N
                                                                         N
         4
                    20
                            N
                                           2
                                                      PhD
                                                                         N N
In [22]: degree_counts=df['Level of Education'].value_counts()
        degree_counts
Out[22]: BS
        PhD 4
        MS
             2
        Name: Level of Education, dtype: int64
In [23]: degree_counts.plot(kind='bar')
Out[23]: <AxesSubplot:>
```

```
In [23]: degree_counts.plot(kind='bar')
Out[23]: <AxesSubplot:>
                                                                                               63
             7 -
             6
             5
             4
             3 -
             2
             1 -
             0 -
                             BS
In [25]: # Series:
In [26]: labels=['a','b','c']
my_list=[10,20,30]
arr=np.array([10,20,30])
d={'a':10,'b':20,'c':30}
In [27]: pd.Series(data=my_list)
Out[27]: 0
                10
           1 20
2 30
dtype: int64
In [28]: pd.Series(data=my_list,index=labels)
Out[28]: a 10
b 20
c 30
dtype: int64
In [29]: pd.Series(my_list, labels)
```

```
In [30]: pd.Series(arr)
Out[30]: 0
                 10
                 20
                30
           dtype: int32
In [31]: pd.Series(arr,labels)
Out[31]: a 10
b 20
c 30
           dtype: int32
In [32]: pd.Series(d)
Out[32]: a 10
b 20
c 30
dtype: int64
In [33]: pd.Series(data=labels)
Out[33]: 0
           1
              b
c
           dtype: object
In [34]: #Even functions (although unlikely that you will use this)
pd.Series([sum,print,len])
           0 <built-in function sum>
1 <built-in function print>
2 <built-in function len>
dtype: object
Out[34]: 0
In [36]: ser1=pd.Series([1,2,3,4],index=['USA','Germany','USSR','Japan'])
          ser1
Out[36]: USA
           Germany
           USSR
                       3 4
           Japan
           dtype: int64
In [37]: ser2=pd.Series([1,2,5,4],index=['USA','Germany','Italy','Japan'])
           ser2
Out[37]: USA
                       1
                      2 5
           Germany
           Italy
           Japan
           dtype: int64
```

```
Out[38]: 1
In [39]: ser1+ser2
Out[39]: Germany
                    4.0
         Italy
                    NaN
                    8.0
         Japan
USA
         USSR
                    NaN
         dtype: float64
In [40]: # DataFrames
In [42]: df=pd.DataFrame(randn(5,4),index='A B C D E'.split(),columns='W X Y Z'.split())
In [43]: df
Out[43]:
                 W
                          X Y
          A 2.706850 0.628133 0.907969 0.503826
          B 0.651118 -0.319318 -0.848077 0.605965
          C -2.018168 0.740122 0.528813 -0.589001
          D 0.188695 -0.758872 -0.933237 0.955057
         E 0.190794 1.978757 2.605967 0.683509
In [44]: # Selection and Indexing
In [45]: df['W']
Out[45]: A 2.706850

8 0.651118

C -2.018168

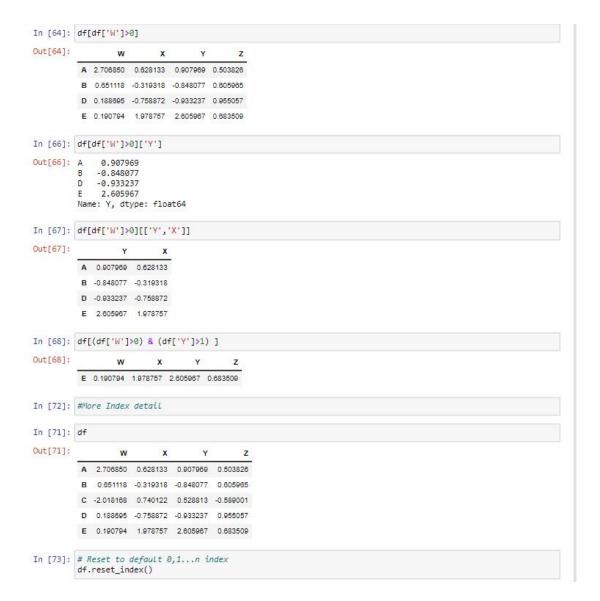
D 0.188695

E 0.190794
         Name: W, dtype: float64
In [48]: #Pass a list of column names
df[['W','Z']]
Out[48]:
         A 2.706850 0.503826
```

```
In [49]: # SQL syntax (NOT RECOMMENDED!) df.W
Out[49]: A 2.706850
         B 0.651118
C -2.018168
D 0.188695
E 0.190794
         Name: W, dtype: float64
In [50]: type(df['W'])
Out[50]: pandas.core.series.Series
In [51]: df['new']=df['W']+df['Y']
Out[51]: W X Y Z
         A 2.706850 0.628133 0.907969 0.503826 3.614819
         B 0.651118 -0.319318 -0.848077 0.605965 -0.196959
         C -2.018168 0.740122 0.528813 -0.589001 -1.489355
         D 0.188695 -0.758872 -0.933237 0.955057 -0.744542
         E 0.190794 1.978757 2.805967 0.683509 2.796762
In [52]: #Return a new Dataframe with the 'new' colum dropped
         df.drop('new',axis=1)
          w x y z
Out[52]:
         A 2.706850 0.628133 0.907969 0.503826
          B 0.651118 -0.319318 -0.848077 0.605965
         C -2.018168 0.740122 0.528813 -0.589001
          D 0.188695 -0.758872 -0.933237 0.955057
         E 0.190794 1.978757 2.605967 0.683509
In [53]: #Not inplace unless specified!
         df
Out[53]:
         W X Y Z
         A 2.706850 0.628133 0.907969 0.503826 3.614819
         B 0.651118 -0.319318 -0.848077 0.605965 -0.196959
         C -2.018168 0.740122 0.528813 -0.589001 -1.489355
         D 0.188895 -0.758872 -0.933237 0.955057 -0.744542
         E 0.190794 1.978757 2.605967 0.683509 2.796762
```

```
In [54]: # Drop the 'new' column of Dataframe itself
df.drop('new',axis=1,inplace=True)
         df
Out[54]: w x y z
         A 2.706850 0.628133 0.907969 0.503826
          B 0.651118 -0.319318 -0.848077 0.605965
         C -2.018168 0.740122 0.528813 -0.589001
         D 0.188695 -0.758872 -0.933237 0.955057
         E 0.190794 1.978757 2.605967 0.683509
In [55]: df.drop('E',axis=0)
Out[55]: w x y z
         A 2.706850 0.628133 0.907969 0.503826
         B 0.651118 -0.319318 -0.848077 0.605965
         C -2.018168 0.740122 0.528813 -0.589001
         D 0.188695 -0.758872 -0.933237 0.955057
In [56]: df.loc['A']
Out[56]: W 2.706850
         X 0.628133
Y 0.907969
Z 0.503826
         Name: A, dtype: float64
In [57]: df.iloc[2]
Out[57]: W -2.018168
         X 0.740122
Y 0.528813
Z -0.589001
         Name: C, dtype: float64
In [58]: df.loc['B','Y']
Out[58]: -0.8480769834036315
In [60]: df.loc[['A','B'],['W','Y']]
Out[60]: w Y
         A 2.706850 0.907969
```

```
In [60]: df.loc[['A','B'],['W','Y']]
Out[60]: w Y
        A 2.706850 0.907969
        B 0.651118 -0.848077
In [61]: # Conditional Selection
Out[61]: w x y z
        A 2.706850 0.628133 0.907969 0.503826
        B 0.651118 -0.319318 -0.848077 0.605965
        C -2.018168 0.740122 0.528813 -0.589001
        D 0.188695 -0.758872 -0.933237 0.955057
        E 0.190794 1.978757 2.605967 0.683509
In [62]: df>0
Out[62]: w x y z
        A True True True True
        B True False False True
        C False True True False
        D True False False True
        E True True True True
In [63]: df[df>0]
Out[63]: w x y z
        A 2.706850 0.628133 0.907969 0.503826
        B 0.651118 NaN NaN 0.605965
        C NaN 0.740122 0.528813 NaN
        D 0.188695 NaN NaN 0.955057
        E 0.190794 1.978757 2.605967 0.683509
```



```
ni i ezer Tilnev()
Out[73]: index W X Y Z
        0 A 2.706850 0.628133 0.907969 0.503826
            B 0.651118 -0.319318 -0.848077 0.605985
        2 C -2.018168 0.740122 0.528813 -0.589001
         3 D 0.188695 -0.758872 -0.933237 0.955057
         4 E 0.190794 1.978757 2.605967 0.683509
In [74]: newind='CA NY WY OR CO'.split()
In [75]: df['States']=newind
Out[75]: W X Y Z States
        A 2.706850 0.628133 0.907969 0.503826 CA
        B 0.651118 -0.319318 -0.848077 0.605965 NY
        C -2.018168 0.740122 0.528813 -0.589001 WY
         D 0.188695 -0.758872 -0.933237 0.955057 OR
         E 0.190794 1.978757 2.605967 0.683509 CO
In [76]: df.set_index('States')
Out[76]:
                 w x y z
        States
        CA 2.706850 0.628133 0.907969 0.503826
          NY 0.651118 -0.319318 -0.848077 0.605965
        WY -2.018168 0.740122 0.528813 -0.589001
           OR 0.188695 -0.758872 -0.933237 0.955057
        CO 0.190794 1.978757 2.605967 0.683509
In [77]: df
Out[77]: W X Y Z States
        A 2.706850 0.628133 0.907969 0.503826 CA
         B 0.651118 -0.319318 -0.848077 0.605965 NY
        C -2.018168 0.740122 0.528813 -0.589001 WY
        D 0.188695 -0.758872 -0.933237 0.955057 OR
         E 0.190794 1.978757 2.605967 0.683509 CO
In [78]: df.set_index('States',inplace=True)
       df
```

```
In [78]: df.set_index('States',inplace=True)
Out[78]:
            w x y z
            States
           CA 2.706850 0.628133 0.907969 0.503826
             NY 0.651118 -0.319318 -0.848077 0.605965
            WY -2.018168 0.740122 0.528813 -0.589001
              OR 0.188695 -0.758872 -0.933237 0.955057
           CO 0.190794 1.978757 2.605967 0.683509
In [80]: # Multi_Index and Index Hierarchy
In [82]: #Index Levels
          #Index Levets
outside=['G1','G1','G1','G2','G2','G2']
inside=[1,2,3,1,2,3]
hier_index = list(zip(outside,inside))
hier_index = pd.MultiIndex.from_tuples(hier_index)
In [83]: hier_index
Out[83]: MultiIndex([('G1', 1),
('G1', 2),
('G1', 3),
('G2', 1),
('G2', 2),
('G2', 3)],
\label{eq:columns} In~[84]:~df=pd.DataFrame(np.random.randn(6,2),index=hier\_index,~columns=['A','B'])
           df
Out[84]:
            A B
              1 0.302665 1.693723
           G1 2 -1.708086 -1.159119
            3 -0.134841 0.390528
              1 0.168905 0.184502
            G2 2 0.807706 0.072960
              3 0.638787 0.329646
In [85]: df.loc['G1']
             Α
Out[85]:
                              В
           1 0.302665 1.693723
           2 -1.706086 -1.159119
           9 0 194941 0 900690
```

```
in [oo]. disindexindmes-[ droup ; num ]
In [89]: df
Out[89]:
                            A
           Group Num
           1 0.302665 1.693723
                 2 -1.706086 -1.159119
                3 -0.134841 0.390528
                    1 0.166905 0.184502
             G2 2 0.807708 0.072960
                    3 0.638787 0.329846
In [90]: df.xs('G1')
Out[90]:
          1 0.302665 1.693723
            2 -1.708088 -1.159119
          3 -0.134841 0.390528
In [92]: df.xs(['G1',1])
          C:\Users\My My\AppData\Local\Temp\ipykernel_27220\580597333.py:1: FutureWarning: Passing lists as key for xs is deprecated and will be removed in a future version. Pass key as a tuple instead.
          df.xs(['G1',1])
Out[92]: A 0.302665
B 1.693723
          Name: (G1, 1), dtype: float64
In [93]: df.xs(1,level='Num')
Out[93]: A B
          Group
          G1 0.302685 1.693723
             G2 0.166905 0.184502
In [94]: # MISSING DATA
In [95]: df =pd.DataFrame({'A':[1,1,np.nan], 'B':[5,np.nan,np.nan], 'C':[1,2,3]})
In [96]: df
Out[96]: A B C
```

```
In [96]: df
Out[96]: A B C 0 1.0 5.0 1
        1 1.0 NaN 2
        2 NaN NaN 3
 In [97]: df.dropna()
Out[97]: A B C
        0 1.0 5.0 1
 In [98]: df.dropna(axis=1)
 Out[98]: c
        0 1
        1 2
        2 3
 In [99]: df.dropna(thresh=2)
 Out[99]: A B C
        0 1.0 5.0 1
        1 1.0 NaN 2
In [100]: df.fillna(value='FILL VALUE')
Out[100]: A
        0 1.0 5.0 1
        1 1.0 FILL VALUE 2
        2 FILL VALUE FILL VALUE 3
In [101]: df['A'].fillna(value=df['A'].mean())
Out[101]: 0 1.0
1 1.0
2 1.0
Name: A, dtype: float64
In [102]: #Group by
        In [103]: df=pd.DataFrame(data)
```

```
In [103]: df=pd.DataFrame(data)
In [104]: df
Out[104]: Company Person Sales
        0 GOOG Sam
        1 GOOG Charlie 120
        2 MSFT Amy 340
        3 MSFT vanessa 124
        4 FB Carl 243
         5 FB Sarah 350
In [105]: df.groupby('Company')
Out[105]: <pandas.core.groupby.generic.DataFrameGroupBy object at 0x0000001A9E5A3CE20>
In [107]: by_comp=df.groupby("Company")
In [108]: by_comp.mean()
Out[108]: Sales
        Company
        FB 296.5
         GOOG 160.0
        MSFT 232.0
In [109]: df.groupby('Company').mean()
Out[109]: Sales
        Company
        FB 296.5
          GOOG 160.0
        MSFT 232.0
In [110]: by_comp.std()
Out[110]:
                 Sales
        Company
        FB 75.660426
         GOOG 56.568542
        MSFT 152.735065
```

```
In [111]: by_comp.min()
Out[111]: Person Sales
       Company
FB Carl 243
         GOOG Charlie 120
        MSFT Amy 124
In [112]: by_comp.max()
Out[112]: Person Sales
        Company
        FB Sarah 350
         GOOG Sam 200
        MSFT vanessa 340
In [113]: by_comp.count()
Out[113]: Person Sales
        Company
       FB 2 2
         G00G 2 2
        MSFT 2 2
In [114]: by_comp.describe()
Out[114]:
          count mean std min 25% 50% 75% max
        Company
        FB 2.0 298.5 75.660426 243.0 269.75 296.5 323.25 350.0
         GOOG 2.0 160.0 56.568542 120.0 140.00 160.0 180.00 200.0
        MSFT 2.0 232.0 152.735085 124.0 178.00 232.0 286.00 340.0
In [115]: by_comp.describe().transpose()
```

```
Out[115]: Company FB GOOG MSFT
               count 2.000000 2.000000 2.000000
               mean 296.500000 160.000000 232.000000
              std 75.880428 58.588542 152.735085
                min 243.000000 120.000000 124.000000
              25% 269.750000 140.000000 178.000000
         Sales
               50% 298.500000 160.000000 232.000000
              75% 323.250000 180.000000 286.000000
                max 350.000000 200.000000 340.000000
 In [116]: by_comp.describe().transpose()['GOOG']
 Out[116]: Sales count
                      2.000000
                    160.000000
              mean
              std
                     56.568542
                    120,000000
              min
                    140.000000
              25%
                   160.000000
180.000000
200.000000
              50%
              75%
              max
         Name: GOOG, dtype: float64
 In [117]: # MERGING, JOINING and CONCATENATING
 In [122]: df1
 Out[122]: A B C D
         0 A0 B0 C0 D0
         1 A1 B1 C1 D1
         2 A2 B2 C2 D2
         3 A3 B3 C3 D3
```

```
In [123]: df2
Out[123]: A B C D
       4 A4 B4 C4 D4
       5 A5 B5 C5 D5
       6 A6 B6 C6 D6
       7 A7 B7 C7 D7
In [124]; df3
Out[124]: A B C D
        8 A8 B8 C8 D8
        9 A9 B9 C9 D9
       10 A10 B10 C10 D10
       11 A11 B11 C11 D11
In [125]: pd.concat([df1,df2,df3])
Out[125]: A B C D
       0 A0 B0 C0 D0
        1 A1 B1 C1 D1
       2 A2 B2 C2 D2
        3 A3 B3 C3 D3
       4 A4 B4 C4 D4
        5 A5 B5 C5 D5
       6 A6 B6 C6 D6
        7 A7 B7 C7 D7
       8 A8 B8 C8 D8
        9 A9 B9 C9 D9
       10 A10 B10 C10 D10
       11 A11 B11 C11 D11
In [126]: pd.concat([df1,df2,df3], axis=1)
```

```
Out[126]: A B C D A B C D A B C D
       0 AO BO CO DO NAN NAN NAN NAN NAN NAN NAN NAN
        1 A1 B1 C1 D1 NaN NaN NaN NaN NaN NaN NaN NaN NaN
       2 A2 B2 C2 D2 NaN NaN NaN NaN NaN NaN NaN NaN
        3 A3 B3 C3 D3 NaN NaN NaN NaN NaN NaN NaN NaN
       4 NaN NaN NaN NaN A4 B4 C4 D4 NaN NaN NaN NaN
        5 NaN NaN NaN NaN A5 B5 C5 D5 NaN NaN NaN NaN
       6 NaN NaN NaN NaN A6 B6 C6 D6 NaN NaN NaN NaN
        7 NaN NaN NaN NaN A7 B7 C7 D7 NaN NaN NaN NaN
       8 NaN NaN NaN NaN NaN NaN NaN A8 B8 C8 D8
        9 NaN NaN NaN NaN NaN NaN NaN A9 B9 C9 D9
       10 NaN NaN NaN NaN NaN NaN NaN A10 B10 C10 D10
        11 NaN NaN NaN NaN NaN NaN NaN A11 B11 C11 D11
In [127]: # Merging
 In [129]: left
Out[129]: key A B
       0 K0 A0 B0
       1 K1 A1 B1
       2 K2 A2 B2
       3 K3 A3 B3
In [130]: right
Out[130]: key C D
       0 K0 C0 D0
       1 K1 C1 D1
       2 K2 C2 D2
        3 K3 C3 D3
In [131]: pd.merge(left,right,how='inner',on='key')
```

```
Out[131]:
        key A B C D
        0 K0 A0 B0 C0 D0
        1 K1 A1 B1 C1 D1
        2 K2 A2 B2 C2 D2
         3 K3 A3 B3 C3 D3
  In [6]: pd.merge(left,right,on=['key1','key2'])
  Out[6]:
          key1 key2 A B C D
        0 K0 K0 A0 B0 C0 D0
        1 K1 K0 A2 B2 C1 D1
        2 K1 K0 A2 B2 C2 D2
  In [7]: pd.merge(left,right,how='outer', on=['key1','key2'])
  Out[7]:
                  A B C D
          key1 key2
        0 K0 K0 A0 B0 C0 D0
              K1 A1
                     B1 NaN NaN
        2
              K0 A2 B2 C1 D1
           K1
              K0 A2 B2 C2 D2
        4 K2 K1 A3 B3 NaN NaN
           K2 K0 NaN NaN C3 D3
  In [8]: pd.merge(left,right,how='right', on=['key1','key2'])
  Out[8]: key1 key2 A B C D
        0
           K0
              KO
                  A0
                     B0 C0 D0
        1 K1 K0 A2 B2 C1 D1
        2 K1 K0 A2 B2 C2 D2
        3 K2 K0 NaN NaN C3 D3
In [9]: pd.merge(left,right,how='left', on=['key1','key2'])
Out[9]: key1 key2 A B C D
       0 K0 K0 A0 B0 C0 D0
       1
         KO
             K1 A1 B1 NaN NaN
       2 K1 K0 A2 B2 C1 D1
       3
         K1 K0 A2 B2 C2 D2
       4 K2 K1 A3 B3 NaN NaN
In [10]: #Joining
right = pd.DataFrame()
```

```
In [127]: left.join(right)
          A B key1 key2 C D
Out[127]:
         KO AO BO NaN NaN NaN NaN
         K1 A1 B1 NaN NaN NaN NaN
         K2 A2 B2 NaN NaN NaN NaN
In [128]: left.join(right, how='outer')
           A B key1 key2 C D
Out[128]:
         KO AO BO NaN NaN NaN NaN
          K1 A1 B1 NaN NaN NaN NaN
         K2 A2 B2 NaN NaN NaN NaN
          0 NaN NaN KO KO CO DO
         1 NaN NaN K1 K0 C1 D1
          2 NaN NaN K1 K0 C2 D2
         3 NaN NaN K2 K0 C3 D3
In [134]:
import pandas as pd
    df = pd.DataFrame({'col1':[1,2,3,4],'col2':[444,555,666,444],'col3':['abc','def','ghi','xyz']})
    df.head()
Out[134]:
         col1 col2 col3
0 1 444 abc
         1 2 555 def
2 3 666 ghi
         3 4 444 xyz
In [135]: df['col2'].unique()
Out[135]: array([444, 555, 666], dtype=int64)
In [136]: df['col2'].nunique()
Out[136]: 3
In [137]: df['col2'].value_counts()
```

```
Out[137]: 444 2
555 1
666 1
          Name: col2, dtype: int64
In [142]: newdf = df[(df['col1']>2) & (df['col2']==444)]
In [143]: newdf
Out[143]: col1 col2 col3
          3 4 444 xyz
In [145]: def times2(x):
         return x*2
In [146]: df['col1'].apply(times2)
Out[146]: 0 2
          1 4
2 6
3 8
         Name: col1, dtype: int64
In [147]: df['col3'].apply(len)
Out[147]: 0 3
          1 3
2 3
3 3
          Name: col3, dtype: int64
In [148]: df['col1'].sum()
Out[148]: 10
In [149]: del df['col1']
In [150]: df
Out[150]: col2 col3
          0 444 abc
           1 555 def
          2 666 ghi
          3 444 xyz
```

```
In [151]: df.columns
Out[151]: Index(['co12', 'co13'], dtype='object')
In [152]: df.index
Out[152]: RangeIndex(start=0, stop=4, step=1)
In [153]: df
Out[153]: col2 col3
        0 444 abc
         1 555 def
         2 666 ghi
         3 444 xyz
In [154]: df.sort_values(by='col2')
Out[154]: col2 col3
         0 444 abc
         3 444 xyz
         1 555 def
         2 666 ghi
In [155]: df.isnull()
Out[155]: col2 col3
         0 False False
         1 False False
         2 False False
         3 False False
In [156]: df.dropna()
Out[156]: col2 col3
         0 444 abc
         1 555 def
         2 666 ghi
         3 444 xyz
```

```
df.head()
Out[158]: col1 col2 col3
        0 1.0 NaN abc
        1 2.0 555.0 def
        2 3.0 666.0 ghi
        3 NaN 444.0 xyz
In [159]: df.isnull()
Out[159]: col1 col2 col3
       0 False True False
        1 False False False
       2 False False False
        3 True False False
In [160]: df.dropna()
Out[160]: col1 col2 col3
       1 2.0 555.0 def
       2 3.0 668.0 ghi
In [161]: df.fillna('FILL')
Out[161]: col1 col2 col3
        0 1.0 FILL abc
        1 2.0 555.0 def
        2 3.0 668.0 ghi
        3 FILL 444.0 xyz
```

```
In [165]: df
Out[165]: A B C D
         0 foo one x 1
         1 foo one y 3
         2 foo two x 2
          3 bar two y 5
         4 bar one x 4
          5 bar one y 1
In [202]: df.pivot_table(values='D',index=['A','B'],columns=['C'])
Out[202]: C x y
         A B
          one 4.0 1.0
two NaN 5.0
          one 1.0 3.0
two 2.0 NaN
In [204]: import numpy as np
         import pandas as pd
In [205]: df = pd.read_csv('example.csv')
Out[205]: A B C D
         0 foo one x 1
          1 foo one y 3
         2 foo two x 2
          3 bar two y 5
         4 bar one x 4
         5 bar one y 1
In [190]: df.to_csv('example.csv',index=False)
```

```
In [7]: import pandas as pd
pd.read_excel('Book1.xlsx',sheet_name='Sheet1')
 Out[7]:
          a b c d
         0 0 1 2 3
         1 4 5 6 7
         2 8 9 10 11
         3 12 13 14 15
In [10]: df.to_excel('Book1.xlsx',sheet_name='Sheet1')
In [11]: from sqlalchemy import create_engine
In [13]: engine = create_engine('sqlite:///memory')
In [18]: df.to_sql('data', engine, if_exists='replace')
Out[18]: 4
In [19]: sql_df = pd.read_sql('data', con=engine)
In [20]: sql_df
Out[20]: index a b c d
         0 0 0 1 2 3
         1 1 4 5 6 7
         2 2 8 9 10 11
         3 3 12 13 14 15
```