pandas example

November 8, 2019

```
[63]: import pandas as pd
     import matplotlib.pyplot as plt
     import numpy as np
[13]: world = {'Team':['west indies','west_
      →indies','India','Australia','Pakistan','Sri

      →lanka', 'Australia', 'Australia', 'Australia', 'India', 'Australia'],
               'Ranks': [7,7,2,1,6,4,1,1,1,2,1],
                'year': [1975,1979,1983,1987,1992,1996,1999,2003,2007,2011,2015]}
[14]: world
[14]: {'Team': ['west indies',
       'west indies',
       'India',
       'Australia',
       'Pakistan',
       'Sri lanka',
       'Australia',
       'Australia',
       'Australia',
       'India',
       'Australia'],
      'Ranks': [7, 7, 2, 1, 6, 4, 1, 1, 1, 2, 1],
      'year': [1975, 1979, 1983, 1987, 1992, 1996, 1999, 2003, 2007, 2011, 2015]}
[15]: df = pd.DataFrame(world)
     df
[15]:
                Team Ranks
                              year
                           7
                              1975
         west indies
     1
         west indies
                           7
                              1979
     2
               India
                             1983
     3
           Australia
                           1
                             1987
     4
            Pakistan
                             1992
     5
           Sri lanka
                           4
                             1996
     6
           Australia
                           1
                             1999
     7
           Australia
                              2003
           Australia
                           1 2007
```

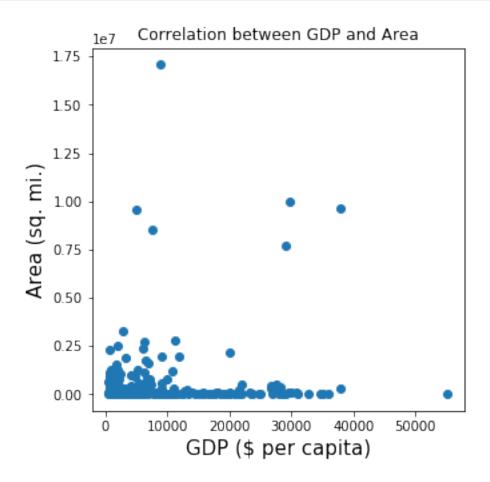
```
9
               India
                          2 2011
     10
           Australia
                          1 2015
[16]: b = df.groupby('Team').groups
[16]: {'Australia': Int64Index([3, 6, 7, 8, 10], dtype='int64'),
      'India': Int64Index([2, 9], dtype='int64'),
      'Pakistan': Int64Index([4], dtype='int64'),
      'Sri lanka': Int64Index([5], dtype='int64'),
      'west indies': Int64Index([0, 1], dtype='int64')}
[17]: c = df.groupby('Team').mean()
[17]:
                  Ranks
                           year
     Team
     Australia
                    1.0 2002.2
     India
                    2.0 1997.0
     Pakistan
                    6.0 1992.0
     Sri lanka
                    4.0 1996.0
                    7.0 1977.0
     west indies
[18]: print(df.groupby(['Team', 'Ranks']).groups)
    {('Australia', 1): Int64Index([3, 6, 7, 8, 10], dtype='int64'), ('India', 2):
    Int64Index([2, 9], dtype='int64'), ('Pakistan', 6): Int64Index([4],
    dtype='int64'), ('Sri lanka', 4): Int64Index([5], dtype='int64'), ('west
    indies', 7): Int64Index([0, 1], dtype='int64')}
[19]: df.groupby('Team').groups
[19]: {'Australia': Int64Index([3, 6, 7, 8, 10], dtype='int64'),
      'India': Int64Index([2, 9], dtype='int64'),
      'Pakistan': Int64Index([4], dtype='int64'),
      'Sri lanka': Int64Index([5], dtype='int64'),
      'west indies': Int64Index([0, 1], dtype='int64')}
[20]: df.groupby('Team').count()
[20]:
                  Ranks year
     Team
     Australia
                      5
                            5
     India
                      2
                            2
     Pakistan
                      1
                            1
     Sri lanka
                      1
                            1
     west indies
 []: # Grouping for India
     Team1 = df.groupby('Team')
     Team1
```

```
print(Team1.get_group('India'))
[]: ##loc method to retrieve the particular column
   df.loc[(df['Team'] =='India')]
[]: d = \{ 'odd' : np.arange(1,100,2), 
          'even': np.arange(0,100,2)}
   d
[]: d['odd']
[]: d['even']
[]: d.keys()
[]: d.values()
[]: df1 = pd.DataFrame(d)
   df1.head(5)
[]: print(df1.groupby('odd').groups)
[]: df = pd.DataFrame(np.random.randn(5,4), columns=['col1','col2','col3','col4'])
   df
[]: # Concantation
   chockers = {
        'Team' : ['South Africa', 'New Zealand', 'Zimbabwe'],
        'Rank' : [1,5,9],
        'points': [895,764,656]
[]: worldcric = {'Team':['west indies','west_
    →indies','India','Australia','Pakistan'],
             'Rank': [7,7,2,1,6],
             'year': [1975,1979,1983,1987,1992],
             'points': [895,764,656,673,844]}
[]: df11 = pd.DataFrame(chockers)
   df22 = pd.DataFrame(worldcric)
[]: df11
[]: df22
[]: print(pd.concat([df11,df22]))
[]: # Merging all data in the data frames
   pd.merge(df11, df22, on= "Team", how= "outer")
[]: # Includes all data in the data frame on left
   pd.merge(df11, df22, on= "Team", how= "left")
```

```
[]: # Includes all data in the data frame on right
     pd.merge(df11, df22, on= "Team", how= "right")
 []: pd.merge(df11, df22, on= "Team", how= "inner")
 []: left = pd.DataFrame({'Key':['k0','K1','K2','K3'],
                            ' A '
                                  :['AO','A1','A2','A3'],
                                   :['B0','B1','B2','B3']})
     left
 []: right = pd.DataFrame({'Key':['k0','K1','K2','K3'],
                               'C':['C0','C1','C2','C3'],
                               'D': ['D0', 'D1', 'D2', 'D3',]})
     right
 []: print(pd.concat([left,right], axis=1))
 []: print(pd.concat([left,right],axis=0))
[83]: country = pd.read_csv(r'Desktop/Countries.csv')
     country.head(5)
[83]:
                Country
                                                         Region Population \
                                ASIA (EX. NEAR EAST)
                                                                    31056997
     0
           Afghanistan
     1
               Albania
                          EASTERN EUROPE
                                                                    3581655
     2
               Algeria
                          NORTHERN AFRICA
                                                                    32930091
     3
       American Samoa
                          OCEANIA
                                                                       57794
                          WESTERN EUROPE
     4
               Andorra
                                                                       71201
        Area (sq. mi.) Pop. Density (per sq. mi.) Coastline (coast/area ratio)
     0
                647500
                                               48,0
                                                                              0,00
     1
                  28748
                                              124,6
                                                                              1,26
     2
               2381740
                                               13,8
                                                                              0,04
     3
                    199
                                              290,4
                                                                             58,29
                    468
                                              152,1
                                                                              0,00
       Net migration Infant mortality (per 1000 births)
                                                            GDP ($ per capita) \
               23,06
     0
                                                    163,07
                                                                          700.0
     1
               -4,93
                                                     21,52
                                                                         4500.0
     2
               -0,39
                                                        31
                                                                         6000.0
     3
              -20,71
                                                      9,27
                                                                         8000.0
                 6,6
                                                      4,05
                                                                        19000.0
       Literacy (%) Phones (per 1000) Arable (%) Crops (%) Other (%) Climate
                                                         0,22
                                                                  87,65
     0
               36,0
                                   3,2
                                             12,13
                                                                               1
               86,5
                                  71,2
                                             21,09
                                                         4,42
                                                                  74,49
                                                                               3
     1
     2
                                  78,1
                                                         0,25
               70,0
                                              3,22
                                                                  96,53
                                                                               1
     3
                                                           15
                                                                     75
                                                                               2
               97,0
                                 259,5
                                                10
              100,0
                                 497,2
                                              2,22
                                                                  97,78
```

```
Birthrate Deathrate Agriculture Industry Service
            46,6
                     20,34
     0
                                  0,38
                                            0,24
                                                    0,38
     1
           15,11
                      5,22
                                 0,232
                                           0,188
                                                   0,579
     2
           17,14
                      4,61
                                 0,101
                                             0,6
                                                   0,298
     3
           22,46
                      3,27
                                   NaN
                                             NaN
                                                     NaN
           8,71
                      6,25
                                   NaN
                                             NaN
                                                     NaN
 []: country.shape
 []: country.dtypes
 country.columns
 country.head(5)
 []: # Describing the columns with dtype as int and float
     country.describe()
 []: a =country.dtypes[country.dtypes == 'object'].index
     country[a].head(5)
 []: | # Describing the objects
     country[a].describe()
 []: country['Population'].head(5)
 []: # Displaying all values Whose population are greater than 500000
     df1 =country.loc[country['Population']>500000]
     df1.head(5)
 ]: df1.shape
 []: # Filtering Countries, whose population is >500000
     df2=country.loc[country['Population']>500000,['Country','Population']]
     df2.head(10)
 []: country.isnull()
[37]: # Correlation between GDP and Area
     import matplotlib.pyplot as plt
     data = country.loc[:,['Country','GDP ($ per capita)','Area (sq. mi.)']]
     plt.figure(figsize=(5,5))
     x = np.array(data['GDP ($ per capita)'])
     y = np.array(data['Area (sq. mi.)'])
     plt.title("Correlation between GDP and Area");
     plt.xlabel('GDP ($ per capita)', fontsize=15)
     plt.ylabel('Area (sq. mi.)', fontsize=15)
```

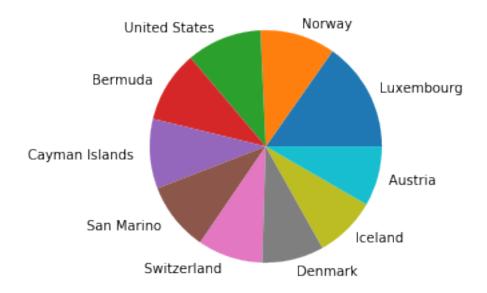
```
plt.scatter(x,y)
plt.show()
```



```
[118]: # 10 Richest country in the world

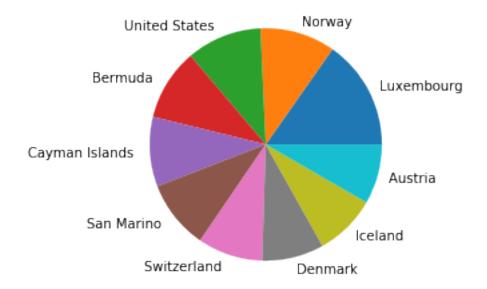
df3=country.sort_values('GDP ($ per capita)',ascending = False)
    df4=df3.loc[:,['Country','GDP ($ per capita)']]
    df5 = df4.head(10)

[119]: plt.pie(df5['GDP ($ per capita)'], labels=df5['Country'])
    plt.figure(figsize=(500,500))
    plt.show()
```



<Figure size 36000x36000 with 0 Axes>

```
[109]: # Another way
      select = country.loc[:,['GDP ($ per capita)','Country']]
      sort1 = select.sort_values('GDP ($ per capita)', ascending = False)
      top10 = sort1.iloc[:10]
      top10
[109]:
           GDP ($ per capita)
                                        Country
                      55100.0
      121
                                    Luxembourg
      154
                      37800.0
                                        Norway
                      37800.0
                                United States
      214
      22
                      36000.0
                                       Bermuda
                      35000.0 Cayman Islands
      38
      177
                      34600.0
                                    San Marino
      196
                      32700.0
                                   Switzerland
      54
                      31100.0
                                       Denmark
                                       Iceland
      93
                      30900.0
                      30000.0
                                       Austria
[110]: plt.pie(top10['GDP ($ per capita)'], labels=top10['Country'])
      plt.figure(figsize=(500,500))
      plt.show()
```



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[]: