Lab 3: Data Preparation

CPE232 Data Models

[1] Reviews on Pandas

- 1.1) Discover
 - methods to explore and understand your DataFrame

```
In [1]: import pandas as pd
          df = pd.read_csv('nss15.csv')
In [2]: # see the shape of the dataframe
          print(df.shape)
         (334839, 12)
In [3]: # seeing the summary of the dataframe
          print(df.info())
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 334839 entries, 0 to 334838
        Data columns (total 12 columns):
         # Column Non-Null Count Dtype
         ___
                                _____
             caseNumber 334839 non-null int64
         1 treatmentDate 334839 non-null object
         2 statWeight 334839 non-null float64
3 stratum 334839 non-null object
4 age 334839 non-null int64
5 sex 334837 non-null object
6 race 205014 non-null object
7 diagnosis 334839 non-null int64
8 bodyPart 334839 non-null int64
         9 disposition 334839 non-null int64
         10 location 334839 non-null int64
11 product 334839 non-null int64
        dtypes: float64(1), int64(7), object(4)
        memory usage: 30.7+ MB
        None
In [4]: # seeing the stats of the column in dataframe
          print(df.describe())
```

caseNumber

```
statWeight
                                                                  diagnosis
                                                         age
       count
               3.348390e+05
                              334839.000000
                                              334839.000000
                                                              334839.000000
               1.510271e+08
                                  39.343028
                                                  31.385451
                                                                  60.154591
       mean
       std
               1.720330e+06
                                  34.142933
                                                  26.105098
                                                                   6.170699
       min
               1.501032e+08
                                   4.965500
                                                   0.000000
                                                                  41.000000
       25%
               1.504405e+08
                                  15.059100
                                                  10.000000
                                                                  57.000000
       50%
               1.507358e+08
                                  15.776200
                                                  23.000000
                                                                  59.000000
       75%
               1.510231e+08
                                  74.881300
                                                  51.000000
                                                                  64.000000
       max
               1.603418e+08
                                  97.923900
                                                 107.000000
                                                                  74.000000
                    bodyPart
                                 disposition
                                                                     product
                                                    location
       count
               334839,000000
                               334839.000000
                                               334839.000000
                                                              334839.000000
                   64.374192
                                                                 2098.900854
       mean
                                    1.307930
                                                    2.485451
       std
                   24.002331
                                    0.977627
                                                    3.217617
                                                                 1332.222670
       min
                    0.000000
                                    1.000000
                                                    0.000000
                                                                  106.000000
       25%
                   35.000000
                                    1.000000
                                                    0.000000
                                                                 1211.000000
       50%
                   75.000000
                                    1.000000
                                                    1.000000
                                                                 1807.000000
       75%
                   82.000000
                                    1.000000
                                                    5.000000
                                                                 3265.000000
                   94.000000
                                    9.000000
                                                    9.000000
                                                                 5555.000000
       max
In [5]: # seeing the first 5 rows of the dataframe
         print(df.head())
           caseNumber treatmentDate statWeight stratum
                                                            age
                                                                    sex
                                                                           race
       0
           150733174
                          7/11/2015
                                          15.7762
                                                        V
                                                             5
                                                                   Male
                                                                            NaN
       1
           150734723
                            7/6/2015
                                          83.2157
                                                         S
                                                             36
                                                                   Male
                                                                         White
       2
           150817487
                            8/2/2015
                                          74.8813
                                                         L
                                                             20
                                                                 Female
                                                                            NaN
       3
            150717776
                           6/26/2015
                                          15.7762
                                                         ٧
                                                             61
                                                                   Male
                                                                            NaN
       4
                           7/4/2015
                                          74.8813
                                                             88
                                                                 Female Other
           150721694
                                                         L
          diagnosis
                      bodyPart
                                 disposition
                                               location
                                                         product
       0
                                                      9
                  57
                             33
                                            1
                                                             1267
       1
                  57
                                            1
                             34
                                                      1
                                                             1439
       2
                  71
                             94
                                            1
                                                      0
                                                             3274
       3
                             35
                  71
                                            1
                                                      0
                                                              611
       4
                             75
                  62
                                            1
                                                      a
                                                             1893
In [6]: # seeing the last 5 rows of the dataframe
         print(df.tail())
                caseNumber treatmentDate
                                           statWeight stratum
                                                                 age
                                                                          sex
                                                                                race
       334834
                 150739278
                                5/31/2015
                                               15.0591
                                                              ٧
                                                                   7
                                                                         Male
                                                                                 NaN
                                                              C
                                                                               Black
       334835
                 150733393
                                7/11/2015
                                                5.6748
                                                                   3
                                                                      Female
       334836
                 150819286
                                7/24/2015
                                               15.7762
                                                              ٧
                                                                  38
                                                                         Male
                                                                                 NaN
                                               97.9239
                                                              Μ
                                                                               White
       334837
                 150823002
                                 8/8/2015
                                                                  38
                                                                      Female
       334838
                 150723074
                                               49.2646
                                                                      Female
                                                                               White
                                6/20/2015
                                      disposition
                            bodyPart
                diagnosis
                                                    location
                                                               product
                       59
                                  76
                                                            1
       334834
                                                 1
                                                                  1864
       334835
                       68
                                  85
                                                 1
                                                            0
                                                                  1931
                                  79
       334836
                       71
                                                 1
                                                            0
                                                                  3250
       334837
                       59
                                  82
                                                 1
                                                            1
                                                                   464
                       57
                                                            9
       334838
                                  34
                                                 1
                                                                  3273
In [7]: # seeing the list of columns in the dataframe
         print(df.columns)
       Index(['caseNumber', 'treatmentDate', 'statWeight', 'stratum', 'age', 'sex',
               'race', 'diagnosis', 'bodyPart', 'disposition', 'location', 'product'],
              dtype='object')
```

1.2) Selecting variables

• select specific columns from the DataFrame to create a new DataFrame with only those columns

```
In [8]: df['age']
Out[8]: 0
                     5
                    36
                    20
          2
          3
                    61
          4
                    88
                    . .
          334834
                   7
          334835
                   3
          334836
                   38
                    38
          334837
          334838
          Name: age, Length: 334839, dtype: int64
 In [9]: df['age'].head()
 Out[9]: 0
               5
          1
               36
          2
              20
          3
              61
               88
          Name: age, dtype: int64
In [10]: df[['caseNumber', 'age']]
Out[10]:
                  caseNumber age
               0
                    150733174
                                 5
                    150734723
                                36
               2
                    150817487
                                20
               3
                    150717776
                                61
               4
                    150721694
                                88
          334834
                    150739278
                                 7
          334835
                    150733393
                                 3
          334836
                    150819286
                                38
          334837
                    150823002
                                38
          334838
                   150723074
                                 5
         334839 rows × 2 columns
In [11]: # select columns based on the data type
         df.select_dtypes(include=['number'])
```

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|----------|-----------|---------------|------------|-----|-----------------|----------|-------------|----------|------|
| Out[11]: | | caseNumber | statWeight | age | diagnosis | bodyPart | disposition | location | proc |
| | 0 | 150733174 | 15.7762 | 5 | 57 | 33 | 1 | 9 | 1 |
| | 1 | 150734723 | 83.2157 | 36 | 57 | 34 | 1 | 1 | 1 |
| | 2 | 150817487 | 74.8813 | 20 | 71 | 94 | 1 | 0 | 3 |
| | 3 | 150717776 | 15.7762 | 61 | 71 | 35 | 1 | 0 | |
| | 4 | 150721694 | 74.8813 | 88 | 62 | 75 | 1 | 0 | 1 |
| | ••• | | | | ··· | | | | |
| | 334834 | 150739278 | 15.0591 | 7 | 59 | 76 | 1 | 1 | 1 |
| | 334835 | 150733393 | 5.6748 | 3 | 68 | 85 | 1 | 0 | 1 |
| | 334836 | 150819286 | 15.7762 | 38 | 71 | 79 | 1 | 0 | 3 |
| | 334837 | 150823002 | 97.9239 | 38 | 59 | 82 | 1 | 1 | |
| | 334838 | 150723074 | 49.2646 | 5 | 57 | 34 | 1 | 9 | 3 |
| | 334839 rd | ows × 8 colum | ns | | | | | | |
| | 4 | | | | | | | | • |
| Tn [12]: | # select | t row by .loc | | | | | | | |

```
In [12]: # select row by .loc
         df.loc[0]
Out[12]: caseNumber
                          150733174
         treatmentDate
                         7/11/2015
                          15.7762
         statWeight
         stratum
                                 ٧
                                 5
         age
                              Male
         sex
         race
                              NaN
                                57
         diagnosis
         bodyPart
                                33
         disposition
                                 1
         location
                                 9
         product
                              1267
         Name: 0, dtype: object
In [13]: # select column by .loc
         df.loc[:6,'treatmentDate':'diagnosis']
```

| Out[13]: | | treatmentDate | statWeight | stratum | age | sex | race | diagnosis |
|----------|---|---------------|------------|---------|-----|--------|-------|-----------|
| | 0 | 7/11/2015 | 15.7762 | V | 5 | Male | NaN | 57 |
| | 1 | 7/6/2015 | 83.2157 | S | 36 | Male | White | 57 |
| | 2 | 8/2/2015 | 74.8813 | L | 20 | Female | NaN | 71 |
| | 3 | 6/26/2015 | 15.7762 | V | 61 | Male | NaN | 71 |
| | 4 | 7/4/2015 | 74.8813 | L | 88 | Female | Other | 62 |
| | 5 | 7/2/2015 | 5.6748 | С | 1 | Female | White | 71 |
| | 6 | 6/8/2015 | 15.7762 | V | 25 | Male | Black | 51 |

In [14]: df.loc[df['age']>80, ['treatmentDate', 'age']]

| \cap | | + | Γ | 1 | Λ | ٦ | |
|--------|---|---|---|---|---|---|---|
| U | и | L | L | - | + | J | - |

| | treatmentDate | age |
|--------|---------------|-----|
| 4 | 7/4/2015 | 88 |
| 8 | 7/16/2015 | 98 |
| 39 | 5/3/2015 | 88 |
| 46 | 4/15/2015 | 91 |
| 63 | 1/12/2015 | 97 |
| ••• | | |
| 334701 | 4/27/2015 | 86 |
| 334784 | 7/7/2015 | 82 |
| 334785 | 7/11/2015 | 86 |
| 334815 | 10/28/2015 | 85 |
| 334819 | 1/13/2015 | 85 |

20422 rows × 2 columns

In [15]: # select row by .iloc df.iloc[0:5]

Out[15]:

| | caseNumber | treatmentDate | statWeight | stratum | age | sex | race | diagnosis | b |
|---|------------|---------------|------------|---------|-----|--------|-------|-----------|---|
| 0 | 150733174 | 7/11/2015 | 15.7762 | V | 5 | Male | NaN | 57 | |
| 1 | 150734723 | 7/6/2015 | 83.2157 | S | 36 | Male | White | 57 | |
| 2 | 150817487 | 8/2/2015 | 74.8813 | L | 20 | Female | NaN | 71 | |
| 3 | 150717776 | 6/26/2015 | 15.7762 | V | 61 | Male | NaN | 71 | |
| 4 | 150721694 | 7/4/2015 | 74.8813 | L | 88 | Female | Other | 62 | |
| | | | | | | | | | |

In [16]: # select column by .iloc
df.iloc[:,[0,1,2,3,4]]

Out[16]:

| | caseNumber | treatmentDate | statWeight | stratum | age |
|--------|------------|---------------|------------|---------|-----|
| 0 | 150733174 | 7/11/2015 | 15.7762 | V | 5 |
| 1 | 150734723 | 7/6/2015 | 83.2157 | S | 36 |
| 2 | 150817487 | 8/2/2015 | 74.8813 | L | 20 |
| 3 | 150717776 | 6/26/2015 | 15.7762 | V | 61 |
| 4 | 150721694 | 7/4/2015 | 74.8813 | L | 88 |
| ••• | | | | | |
| 334834 | 150739278 | 5/31/2015 | 15.0591 | V | 7 |
| 334835 | 150733393 | 7/11/2015 | 5.6748 | С | 3 |
| 334836 | 150819286 | 7/24/2015 | 15.7762 | V | 38 |
| 334837 | 150823002 | 8/8/2015 | 97.9239 | М | 38 |
| 334838 | 150723074 | 6/20/2015 | 49.2646 | М | 5 |

334839 rows × 5 columns

1.3) Filtering the data

In [17]: # filter rows based on the condition
df[df['age'] > 50]

Out[17]:

| | caseNumber | treatmentDate | statWeight | stratum | age | sex | race | diagno |
|-------|--------------------|---------------|------------|---------|-----|--------|-------|--------|
| | 3 150717776 | 6/26/2015 | 15.7762 | V | 61 | Male | NaN | |
| | 4 150721694 | 7/4/2015 | 74.8813 | L | 88 | Female | Other | |
| | 7 150704114 | 6/14/2015 | 83.2157 | S | 53 | Male | White | |
| : | 3 150736558 | 7/16/2015 | 83.2157 | S | 98 | Male | Black | |
| 10 | 6 150901411 | 8/27/2015 | 83.2157 | S | 65 | Female | White | |
| • | | | | | | | | |
| 33481 | 1 150702215 | 6/27/2015 | 15.7762 | V | 51 | Female | NaN | |
| 33481 | 5 151100368 | 10/28/2015 | 83.2157 | S | 85 | Female | NaN | |
| 33481 | 9 150528367 | 1/13/2015 | 49.2646 | М | 85 | Female | NaN | |
| 33482 | 6 150648619 | 6/17/2015 | 15.7762 | V | 52 | Female | White | |
| 33482 | 9 150633526 | 4/4/2015 | 49.2646 | М | 51 | Female | NaN | |

85235 rows × 12 columns



```
In [18]: # filter coloum based on column name
    df.filter(like='age')
```

| Out[18]: | | age |
|----------|--------|-----|
| | 0 | 5 |
| | 1 | 36 |
| | 2 | 20 |
| | 3 | 61 |
| | 4 | 88 |
| | ••• | |
| | 334834 | 7 |
| | 334835 | 3 |
| | 334836 | 38 |
| | 334837 | 38 |
| | 334838 | 5 |

334839 rows × 1 columns

- 1.4) Sorting
 - Sort the DataFrame by its index based on column

```
In [19]: # sort the dataframe based on column name and ascending order
df.sort_values(by='statWeight', ascending=False)
```

| \cap | urt | ٠г | 1 | a | 7 | |
|--------|-----|----|---|---|---|---|
| \cup | uч | ٠L | _ | J | J | ۰ |

| | caseNumber | treatmentDate | statWeight | stratum | age | sex | race | diagno |
|--------|------------|---------------|------------|---------|-----|--------|-------|--------|
| 67072 | 150533084 | 5/15/2015 | 97.9239 | М | 89 | Male | NaN | |
| 313846 | 150521217 | 4/18/2015 | 97.9239 | М | 36 | Female | NaN | |
| 230135 | 150857760 | 8/25/2015 | 97.9239 | М | 14 | Male | White | |
| 141323 | 151039262 | 10/11/2015 | 97.9239 | М | 39 | Female | White | |
| 230141 | 150662453 | 6/5/2015 | 97.9239 | М | 11 | Female | White | |
| ••• | | | | | | | | |
| 122009 | 151146792 | 11/15/2015 | 4.9655 | С | 2 | Female | White | |
| 211090 | 151253201 | 12/15/2015 | 4.9655 | С | 2 | Male | White | |
| 317625 | 160106638 | 12/25/2015 | 4.9655 | С | 1 | Male | White | |
| 33679 | 151256307 | 12/20/2015 | 4.9655 | С | 9 | Female | Black | |
| 229596 | 160148171 | 12/4/2015 | 4.9655 | С | 16 | Female | Other | |

334839 rows × 12 columns

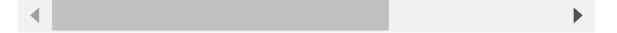


In [20]: # sort the index of the dataframe
df.sort_index()

Out[20]:

| | caseNumber | treatmentDate | statWeight | stratum | age | sex | race | diagno |
|--------|------------|---------------|------------|---------|-----|--------|-------|--------|
| 0 | 150733174 | 7/11/2015 | 15.7762 | V | 5 | Male | NaN | |
| 1 | 150734723 | 7/6/2015 | 83.2157 | S | 36 | Male | White | |
| 2 | 150817487 | 8/2/2015 | 74.8813 | L | 20 | Female | NaN | |
| 3 | 150717776 | 6/26/2015 | 15.7762 | V | 61 | Male | NaN | |
| 4 | 150721694 | 7/4/2015 | 74.8813 | L | 88 | Female | Other | |
| | | | | | | | | |
| 334834 | 150739278 | 5/31/2015 | 15.0591 | V | 7 | Male | NaN | |
| 334835 | 150733393 | 7/11/2015 | 5.6748 | С | 3 | Female | Black | |
| 334836 | 150819286 | 7/24/2015 | 15.7762 | V | 38 | Male | NaN | |
| 334837 | 150823002 | 8/8/2015 | 97.9239 | М | 38 | Female | White | |
| 334838 | 150723074 | 6/20/2015 | 49.2646 | М | 5 | Female | White | |

334839 rows × 12 columns



1.5) Add/Remove

• This section shows how to manipulate the DataFrame's structure

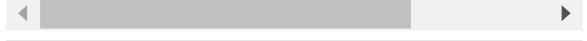
In [20]: # Dropping the column
df.drop(columns=['disposition'])

| Out[20]: | | caseNumber | treatmentDate | statWeight | stratum | age | sex | race | diagno |
|----------|--------|------------|---------------|------------|---------|-----|--------|-------|--------|
| | 0 | 150733174 | 7/11/2015 | 15.7762 | V | 5 | Male | NaN | |
| | 1 | 150734723 | 7/6/2015 | 83.2157 | S | 36 | Male | White | |
| | 2 | 150817487 | 8/2/2015 | 74.8813 | L | 20 | Female | NaN | |
| | 3 | 150717776 | 6/26/2015 | 15.7762 | V | 61 | Male | NaN | |
| | 4 | 150721694 | 7/4/2015 | 74.8813 | L | 88 | Female | Other | |
| | ••• | | | | | | | | |
| | 334834 | 150739278 | 5/31/2015 | 15.0591 | V | 7 | Male | NaN | |
| | 334835 | 150733393 | 7/11/2015 | 5.6748 | С | 3 | Female | Black | |
| | 334836 | 150819286 | 7/24/2015 | 15.7762 | V | 38 | Male | NaN | |
| | 334837 | 150823002 | 8/8/2015 | 97.9239 | М | 38 | Female | White | |
| | | | | | | | | | |

6/20/2015

334839 rows × 11 columns

334838 150723074



49.2646

M 5 Female White

In [22]: # Adding column and create into a new column
df.assign(new_column=df['diagnosis'] + df['bodyPart'])

| Out[22]: | | caseNumber | treatmentDate | statWeight | stratum | age | sex | race | diagno |
|----------|--------|------------|---------------|------------|---------|-----|--------|-------|--------|
| | 0 | 150733174 | 7/11/2015 | 15.7762 | V | 5 | Male | NaN | |
| | 1 | 150734723 | 7/6/2015 | 83.2157 | S | 36 | Male | White | |
| | 2 | 150817487 | 8/2/2015 | 74.8813 | L | 20 | Female | NaN | |
| | 3 | 150717776 | 6/26/2015 | 15.7762 | V | 61 | Male | NaN | |
| | 4 | 150721694 | 7/4/2015 | 74.8813 | L | 88 | Female | Other | |
| | ••• | | | | | | | | |
| | 334834 | 150739278 | 5/31/2015 | 15.0591 | V | 7 | Male | NaN | |
| | 334835 | 150733393 | 7/11/2015 | 5.6748 | С | 3 | Female | Black | |
| | 334836 | 150819286 | 7/24/2015 | 15.7762 | V | 38 | Male | NaN | |
| | 334837 | 150823002 | 8/8/2015 | 97.9239 | М | 38 | Female | White | |
| | 334838 | 150723074 | 6/20/2015 | 49.2646 | М | 5 | Female | White | |

334839 rows × 13 columns



In [21]: # Removing the column and assigning it to a new variable
ages = df.pop('age')

- 1.6) Clean missing
 - to remove rows with missing values or replace missing values with a specified value
- In [22]: # replaceing the missing values with a specified value
 df.fillna(value=0)

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|----------|-----------|----------------|---------------|------------------|----------|--------|-------|-----------|---|
| Out[22]: | | caseNumber | treatmentDate | statWeight | stratum | sex | race | diagnosis | b |
| | 0 | 150733174 | 7/11/2015 | 15.7762 | V | Male | 0 | 57 | |
| | 1 | 150734723 | 7/6/2015 | 83.2157 | S | Male | White | 57 | |
| | 2 | 150817487 | 8/2/2015 | 74.8813 | L | Female | 0 | 71 | |
| | 3 | 150717776 | 6/26/2015 | 15.7762 | V | Male | 0 | 71 | |
| | 4 | 150721694 | 7/4/2015 | 74.8813 | L | Female | Other | 62 | |
| | ••• | | | | | | | | |
| | 334834 | 150739278 | 5/31/2015 | 15.0591 | V | Male | 0 | 59 | |
| | 334835 | 150733393 | 7/11/2015 | 5.6748 | С | Female | Black | 68 | |
| | 334836 | 150819286 | 7/24/2015 | 15.7762 | V | Male | 0 | 71 | |
| | 334837 | 150823002 | 8/8/2015 | 97.9239 | М | Female | White | 59 | |
| | 334838 | 150723074 | 6/20/2015 | 49.2646 | М | Female | White | 57 | |
| | 334839 rd | ows × 11 colun | nns | | | | | | |
| | 4 | | | | | | | • | • |



| Out[23]: | | caseNumber | treatmentDate | statWeight | stratum | sex | race | diagnosis | b |
|----------|--------|------------|---------------|------------|---------|--------|-------|-----------|---|
| | 1 | 150734723 | 7/6/2015 | 83.2157 | S | Male | White | 57 | |
| | 4 | 150721694 | 7/4/2015 | 74.8813 | L | Female | Other | 62 | |
| | 5 | 150721815 | 7/2/2015 | 5.6748 | С | Female | White | 71 | |
| | 6 | 150713483 | 6/8/2015 | 15.7762 | V | Male | Black | 51 | |
| | 7 | 150704114 | 6/14/2015 | 83.2157 | S | Male | White | 57 | |
| | ••• | | | | | | | | |
| | 334830 | 150628863 | 6/8/2015 | 15.7762 | V | Female | White | 64 | |
| | 334831 | 150607637 | 5/22/2015 | 5.6748 | С | Female | Black | 59 | |
| | 334835 | 150733393 | 7/11/2015 | 5.6748 | С | Female | Black | 68 | |
| | 334837 | 150823002 | 8/8/2015 | 97.9239 | М | Female | White | 59 | |
| | 334838 | 150723074 | 6/20/2015 | 49.2646 | М | Female | White | 57 | |

205014 rows × 11 columns



[2] Data Cleaning and Preparation

.isnull, .dropna, .fillna

2.1) checking

```
In [24]:
        df.columns
Out[24]: Index(['caseNumber', 'treatmentDate', 'statWeight', 'stratum', 'sex', 'race',
                 'diagnosis', 'bodyPart', 'disposition', 'location', 'product'],
               dtype='object')
In [25]: # isnull checking
         df.isnull().sum()
Out[25]: caseNumber
                               0
         treatmentDate
                               0
         statWeight
         stratum
         sex
                         129825
         race
         diagnosis
                               0
         bodyPart
                               0
         disposition
                               0
         location
                               0
         product
         dtype: int64
In [26]: # percentage of missing values for the race
         df.race.isnull().sum()/df.shape[0]*100
Out[26]: 38.772365226272925
In [27]:
         df.shape[0]
Out[27]: 334839
         2.2) Drop column
In [28]: # remove column by using
         df = df.drop(columns=['race'])
In [29]: df.head()
```

| Out[29]: | | caseNumber | treatmentDate | statWeight | stratum | sex | diagnosis | bodyPart | disp |
|----------|---|------------|---------------|------------|---------|--------|-----------|----------|------|
| | 0 | 150733174 | 7/11/2015 | 15.7762 | V | Male | 57 | 33 | |
| | 1 | 150734723 | 7/6/2015 | 83.2157 | S | Male | 57 | 34 | |
| | 2 | 150817487 | 8/2/2015 | 74.8813 | L | Female | 71 | 94 | |
| | 3 | 150717776 | 6/26/2015 | 15.7762 | V | Male | 71 | 35 | |
| | 4 | 150721694 | 7/4/2015 | 74.8813 | L | Female | 62 | 75 | |
| | 4 | | | | | | | | • |

2.3) Data imputation

```
In [30]: # fillna
df['age'] = df['age'].fillna(df['age'].median())
```

```
KeyError
                                          Traceback (most recent call last)
File ~\AppData\Local\Packages\PythonSoftwareFoundation.Python.3.11_qbz5n2kfra8p0
\LocalCache\local-packages\Python311\site-packages\pandas\core\indexes\base.py:38
05, in Index.get_loc(self, key)
  3804 try:
-> 3805
            return self._engine.get_loc(casted_key)
   3806 except KeyError as err:
File index.pyx:167, in pandas._libs.index.IndexEngine.get_loc()
File index.pyx:196, in pandas._libs.index.IndexEngine.get_loc()
File pandas\\_libs\\hashtable_class_helper.pxi:7081, in pandas._libs.hashtable.Py
ObjectHashTable.get_item()
File pandas\\_libs\\hashtable_class_helper.pxi:7089, in pandas._libs.hashtable.Py
ObjectHashTable.get_item()
KeyError: 'age'
The above exception was the direct cause of the following exception:
                                          Traceback (most recent call last)
KeyError
Cell In[30], line 2
      1 # fillna
----> 2 df['age'] = df['age'].fillna(df['age'].median())
File ~\AppData\Local\Packages\PythonSoftwareFoundation.Python.3.11_qbz5n2kfra8p0
\LocalCache\local-packages\Python311\site-packages\pandas\core\frame.py:4102, in
DataFrame.__getitem__(self, key)
  4100 if self.columns.nlevels > 1:
            return self._getitem_multilevel(key)
-> 4102 indexer = self.columns.get_loc(key)
  4103 if is integer(indexer):
            indexer = [indexer]
   4104
File ~\AppData\Local\Packages\PythonSoftwareFoundation.Python.3.11_qbz5n2kfra8p0
\LocalCache\local-packages\Python311\site-packages\pandas\core\indexes\base.py:38
12, in Index.get loc(self, key)
           if isinstance(casted_key, slice) or (
   3807
   3808
                isinstance(casted key, abc.Iterable)
                and any(isinstance(x, slice) for x in casted_key)
   3809
   3810
          ):
  3811
                raise InvalidIndexError(key)
-> 3812
           raise KeyError(key) from err
  3813 except TypeError:
   3814
          # If we have a listlike key, _check_indexing_error will raise
  3815
            # InvalidIndexError. Otherwise we fall through and re-raise
          # the TypeError.
  3816
   3817
           self. check indexing error(key)
KeyError: 'age'
```

[Q1] From the above cell, Why it showing an error?

Ans: ก่อนหน้านี้ใช้ ages = df.pop('age') ทำให้ตอนนี้ไม่มี column age แล้ว ดังนั้นเรียกใช้ column age ไม่ได้

[Q2] Fix the error from Q1 problem.

```
In [31]: # [Q2]

# hint: see the cell that run `df.pop()

# Ans :
df["age"] = ages

# fillna again
df['age'] = df['age'].fillna(df['age'].median())

df.head()
```

Out[31]: caseNumber treatmentDate statWeight stratum sex diagnosis bodyPart disp 0 150733174 ٧ 57 33 7/11/2015 15.7762 Male 1 150734723 7/6/2015 83.2157 S Male 57 34 2 71 150817487 94 8/2/2015 74.8813 L Female 3 150717776 6/26/2015 Male 71 35 15.7762 75 62 4 150721694 7/4/2015 74.8813 L Female

2.4) Drop row that have missing value

```
In [32]: # remove column by using .dropna()
         df = df.dropna()
In [33]:
         df.isnull().sum()
Out[33]: caseNumber
                           0
          treatmentDate
                           0
          statWeight
                           0
          stratum
                           0
          sex
          diagnosis
          bodyPart
                           0
          disposition
          location
                           0
          product
                           0
          age
          dtype: int64
```

Datetime

2.5) Working with the datetime format

```
In [34]: df["treatmentDate"] = pd.to_datetime(df["treatmentDate"], format="%m/%d/%Y")
```

```
C:\Users\punch\AppData\Local\Temp\ipykernel_15332\3208943844.py:1: SettingWithCop
        yWarning:
        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/stabl
        e/user_guide/indexing.html#returning-a-view-versus-a-copy
          df["treatmentDate"] = pd.to_datetime(df["treatmentDate"], format="%m/%d/%Y")
In [35]: df.info()
        <class 'pandas.core.frame.DataFrame'>
        Index: 334837 entries, 0 to 334838
        Data columns (total 11 columns):
                     Non-Null Count Dtype
         # Column
        --- -----
                            -----
         0 caseNumber
                           334837 non-null int64
             treatmentDate 334837 non-null datetime64[ns]
         1
         2 statWeight 334837 non-null float64
3 stratum 334837 non-null object
4 sex 334837 non-null object
5 diagnosis 334837 non-null int64
6 bodyPart 334837 non-null int64
         7 disposition 334837 non-null int64
         8 location
                           334837 non-null int64
         9
             product
                            334837 non-null int64
         10 age
                            334837 non-null int64
        dtypes: datetime64[ns](1), float64(1), int64(7), object(2)
        memory usage: 30.7+ MB
In [36]: df['Year'] = df['treatmentDate'].dt.year
        C:\Users\punch\AppData\Local\Temp\ipykernel_15332\1686165144.py:1: SettingWithCop
        yWarning:
        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/stabl
        e/user guide/indexing.html#returning-a-view-versus-a-copy
          df['Year'] = df['treatmentDate'].dt.year
In [37]: df['Month'] = df['treatmentDate'].dt.month
        C:\Users\punch\AppData\Local\Temp\ipykernel 15332\404848564.py:1: SettingWithCopy
        Warning:
        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/stabl
        e/user guide/indexing.html#returning-a-view-versus-a-copy
         df['Month'] = df['treatmentDate'].dt.month
In [38]: df.head()
```

| Out[38]: | | caseNumber | treatmentDate | statWeight | stratum | sex | diagnosis | bodyPart | disp |
|----------|---|------------|---------------|------------|---------|--------|-----------|----------|------|
| | 0 | 150733174 | 2015-07-11 | 15.7762 | V | Male | 57 | 33 | |
| | 1 | 150734723 | 2015-07-06 | 83.2157 | S | Male | 57 | 34 | |
| | 2 | 150817487 | 2015-08-02 | 74.8813 | L | Female | 71 | 94 | |
| | 3 | 150717776 | 2015-06-26 | 15.7762 | V | Male | 71 | 35 | |
| | 4 | 150721694 | 2015-07-04 | 74.8813 | L | Female | 62 | 75 | |
| | | | | | | | | | • |

[Q3] Can you change the format to DD/MM/YYYY? Show your work.

```
In [39]: # write your code here
         df["treatmentDate"] = df["treatmentDate"].dt.strftime("%d/%m/%Y")
         df.head()
        C:\Users\punch\AppData\Local\Temp\ipykernel_15332\366907535.py:2: SettingWithCopy
        Warning:
        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/stabl
        e/user_guide/indexing.html#returning-a-view-versus-a-copy
          df["treatmentDate"] = df["treatmentDate"].dt.strftime("%d/%m/%Y")
Out[39]:
             caseNumber treatmentDate statWeight stratum
                                                                sex diagnosis
                                                                               bodyPart disp
               150733174
                                            15.7762
          0
                             11/07/2015
                                                               Male
                                                                           57
                                                                                     33
          1
               150734723
                             06/07/2015
                                            83.2157
                                                               Male
                                                                           57
                                                                                     34
          2
               150817487
                             02/08/2015
                                            74.8813
                                                          L Female
                                                                           71
                                                                                     94
          3
                                                                                     35
               150717776
                             26/06/2015
                                            15.7762
                                                               Male
                                                                           71
               150721694
                             04/07/2015
                                                                                     75
                                            74.8813
                                                          L Female
                                                                           62
```

Combine Dataframe by .merge and .concat

2.6 Merge

| Out[41]: | | Customer ID | Returned |
|----------|-----|-------------|----------|
| | 0 | ZD-21925 | Yes |
| | 3 | TB-21055 | Yes |
| | 10 | JS-15685 | Yes |
| | 13 | LC-16885 | Yes |
| | 20 | BS-11755 | Yes |
| | ••• | | |
| | 688 | ED-13885 | Yes |
| | 689 | TS-21205 | Yes |
| | 696 | MF-17665 | Yes |
| | 702 | SH-19975 | Yes |
| | 705 | RB-19435 | Yes |

222 rows × 2 columns

[Q4] What does the argument how="inner" do?

Ans: how="inner" จะรวมเฉพาะข้อมูลที่มีค่าตรงกันในคอลัมน์ที่ใช้เชื่อมจากทั้งสอง DataFrame เท่านั้น

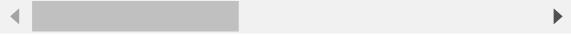
[Q5] In your opinion, what information that the result above conveys?

Ans: ได้ข้อมูลการคืนสินค้าจาก Customer ID ที่คืนแล้วโดยไม่มีข้อมูลซ้ำ

More merging...

```
In [42]: superstore_order.merge(superstore_return,
    on="Order ID" ,
    how="inner")
```

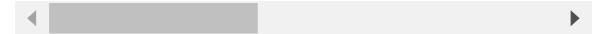
| Out[42]: | | Row ID | Order ID | Order Date | Ship Date | Ship Mode | Customer ID | Customer Name | Segment |
|----------|--------|-----------|------------------------|---------------|------------|-------------------|----------------|-----------------------|-----------|
| | 0 | 19 | CA- 2014- 143336 | 27/08/2014 | 01/09/2014 | Second Class | ZD-21925 | Zuschuss Donatelli | Consumer |
| | 1 | 20 | CA- 2014- 143336 | 27/08/2014 | 01/09/2014 | Second Class | ZD-21925 | Zuschuss Donatelli | Consumer |
| | 2 | 21 | CA- 2014- 143336 | 27/08/2014 | 01/09/2014 | Second Class | ZD-21925 | Zuschuss Donatelli | Consumer |
| | 3 | 56 | CA- 2016- 111682 | 17/06/2016 | 18/06/2016 | First Class | TB-21055 | Ted Butterfield | Consumer |
| | 4 | 57 | CA- 2016- 111682 | 17/06/2016 | 18/06/2016 | First Class | TB-21055 | Ted Butterfield | Consumer |
| | ••• | | | | | | | | |
| | 702 | 8870 | CA- 2017- 101805 | 01/12/2017 | 06/12/2017 | Standard Class | SH-19975 | Sally Hughsby | Corporate |
| | 703 | 8871 | CA- 2017- 101805 | 01/12/2017 | 06/12/2017 | Standard Class | SH-19975 | Sally Hughsby | Corporate |
| | 704 | 8872 | CA- 2017- 101805 | 01/12/2017 | 06/12/2017 | Standard Class | SH-19975 | Sally Hughsby | Corporate |
| | 705 | 8873 | US- 2014- 105137 | 10/10/2014 | 10/10/2014 | Same Day | RB-19435 | Richard Bierner | Consumer |
| | 706 | 8874 | US- 2014- 105137 | 10/10/2014 | 10/10/2014 | Same Day | RB-19435 | Richard Bierner | Consumer |
| | 707 rd | ows × 2 | 22 columi | ns | | | | | |
| | 4 | | | | | | | | • |



2.7) Concatenate

| In [43]: | pd | · conca | <pre>concat([superstore_order, superstore_people], axis=1, join='inner')</pre> | | | | | | | | | |
|----------|----|-----------|--|---------------|------------|-------------------|----------------|--------------------|-----------|--------|--|--|
| Out[43]: | | Row ID | Order ID | Order Date | Ship Date | Ship Mode | Customer ID | Customer Name | Segment | Co | | |
| | 0 | 1 | CA- 2016- 152156 | 08/11/2016 | 11/11/2016 | Second Class | CG-12520 | Claire Gute | Consumer | ر : | | |
| | 1 | 2 | CA- 2016- 152156 | 08/11/2016 | 11/11/2016 | Second Class | CG-12520 | Claire Gute | Consumer | L : | | |
| | 2 | 3 | CA- 2016- 138688 | 12/06/2016 | 16/06/2016 | Second Class | DV-13045 | Darrin Van Huff | Corporate | L ! | | |
| | 3 | 4 | US- 2015- 108966 | 11/10/2015 | 18/10/2015 | Standard Class | SO-20335 | Sean ODonnell | Consumer | L : | | |

4 rows × 23 columns



[Q6] What is the difference between inner and outer on parameter join in pd.concat ?

ANS: inner : จะเก็บเฉพาะคอลัมน์หรือแถวที่มีอยู่ในทั้ง DataFrame ที่ทำการเชื่อมต่อ ถ้า DataFrame ใด ขาดคอลัมน์หรือแถวที่ต้องการ จะไม่ถูกนำมารวมในผลลัพธ์ outer : จะเก็บทุกคอลัมน์หรือแถวจาก DataFrame ทั้งหมด และจะเติมค่า NaN ให้ในกรณีที่มีคอลัมน์หรือแถวใดขาดหายไป ทำการรวมคอลัมน์ หรือแถวที่มีจากทุก DataFrame

Groupby

In [44]: superstore_order.groupby(['Segment','Ship Mode'])[['Sales','Quantity','Discount'

Out[44]:

| | | Sales | Quantity | Discount | Profit |
|-------------|----------------|-------------|----------|----------|------------|
| Segment | Ship Mode | | | | |
| Consumer | First Class | 138594.9328 | 2455 | 110.29 | 18953.7264 |
| | Same Day | 53660.6340 | 1001 | 43.85 | 8555.7193 |
| | Second Class | 203605.6822 | 3489 | 127.29 | 24701.9148 |
| | Standard Class | 627061.3262 | 10430 | 443.05 | 68864.9892 |
| Corporate | First Class | 97720.1209 | 1670 | 73.07 | 12660.2526 |
| | Same Day | 41716.5550 | 366 | 14.50 | 1120.9222 |
| | Second Class | 130759.9288 | 2027 | 71.47 | 15582.1762 |
| | Standard Class | 359359.2109 | 6203 | 262.82 | 49832.6780 |
| Home Office | First Class | 76743.8674 | 924 | 39.82 | 11829.8821 |
| | Same Day | 20968.5170 | 343 | 12.50 | 3909.3442 |
| | Second Class | 77175.1080 | 1148 | 37.80 | 12785.8953 |
| | | | | | |

[Q7] Describe an information that the result above conveys?

Standard Class 218325.9795

Ans: แสดงข้อมูลการผลรวมของ Sales, Quantity, Discount, Profit ตาม Segment และ Ship Mode

3595

142.14 27298.5786

```
In [45]: superstore_order["Profit Ratio"] = superstore_order["Profit"]/superstore_order["
In [62]: superstore_order.groupby(["Category", "Sub-Category"]).agg(mean_profit_ratio = (
```

Out[62]:

mean_profit_ratio

| Category | Sub-Category | |
|-----------------|--------------|-----------|
| Furniture | Bookcases | -0.127756 |
| | Chairs | 0.045028 |
| | Furnishings | 0.140782 |
| | Tables | -0.147916 |
| Office Supplies | Appliances | -0.145513 |
| | Art | 0.251678 |
| | Binders | -0.191641 |
| | Envelopes | 0.421913 |
| | Fasteners | 0.301157 |
| | Labels | 0.429984 |
| | Paper | 0.425586 |
| | Storage | 0.092382 |
| | Supplies | 0.104970 |
| Technology | Accessories | 0.219012 |
| | Copiers | 0.317826 |
| | Machines | -0.059535 |
| | Phones | 0.118926 |

[Q8] Describe an information that the result above conveys?

Ans: แสดงค่าค่าเฉลี่ย Profit Ratio ในแต่ละ Category และ Sub-Category

Pivot and Melt

Pivot

In [46]: superstore_order.pivot_table(index="State", columns="Ship Mode", values="Order I

| Out[46]: | Ship Mode | First Class | Same Day | Second Class | Standard Class |
|----------|----------------------|-------------|----------|---------------------|----------------|
| | State | | | | |
| | Alabama | 9.0 | 1.0 | 18.0 | 30.0 |
| | Arizona | 42.0 | 15.0 | 22.0 | 123.0 |
| | Arkansas | 10.0 | 2.0 | 8.0 | 35.0 |
| | California | 302.0 | 106.0 | 346.0 | 1000.0 |
| | Colorado | 43.0 | 5.0 | 32.0 | 95.0 |
| | Connecticut | 19.0 | 8.0 | 11.0 | 39.0 |
| | Delaware | 16.0 | 2.0 | 13.0 | 55.0 |
| | District of Columbia | 0.0 | 0.0 | 3.0 | 7.0 |
| | Florida | 47.0 | 25.0 | 57.0 | 210.0 |
| | Georgia | 19.0 | 15.0 | 31.0 | 108.0 |

In [47]: pivot_table_result = superstore_order.pivot_table(index="State", columns="Ship M
 print(pivot_table_result)

| Ship Mode | First Class | Same Day | Second Class | Standard Class |
|----------------------|-------------|----------|--------------|----------------|
| State | | | | |
| Alabama | 9.0 | 1.0 | 18.0 | 30.0 |
| Arizona | 42.0 | 15.0 | 22.0 | 123.0 |
| Arkansas | 10.0 | 2.0 | 8.0 | 35.0 |
| California | 302.0 | 106.0 | 346.0 | 1000.0 |
| Colorado | 43.0 | 5.0 | 32.0 | 95.0 |
| Connecticut | 19.0 | 8.0 | 11.0 | 39.0 |
| Delaware | 16.0 | 2.0 | 13.0 | 55.0 |
| District of Columbia | 0.0 | 0.0 | 3.0 | 7.0 |
| Florida | 47.0 | 25.0 | 57.0 | 210.0 |
| Georgia | 19.0 | 15.0 | 31.0 | 108.0 |
| Idaho | 3.0 | 0.0 | 2.0 | 13.0 |
| Illinois | 58.0 | 24.0 | 96.0 | 249.0 |
| Indiana | 13.0 | 3.0 | 30.0 | 79.0 |
| Iowa | 1.0 | 1.0 | 4.0 | 17.0 |
| Kansas | 6.0 | 1.0 | 2.0 | 15.0 |
| Kentucky | 12.0 | 5.0 | 49.0 | 62.0 |
| Louisiana | 7.0 | 2.0 | 14.0 | 15.0 |
| Maine | 0.0 | 0.0 | 0.0 | 5.0 |
| Maryland | 18.0 | 7.0 | 12.0 | 63.0 |
| Massachusetts | 14.0 | 4.0 | 35.0 | 71.0 |
| Michigan | 20.0 | 16.0 | 43.0 | 151.0 |
| Minnesota | 9.0 | 4.0 | 13.0 | 59.0 |
| Mississippi | 3.0 | 4.0 | 7.0 | 36.0 |
| Missouri | 7.0 | 2.0 | 20.0 | 24.0 |
| Montana | 1.0 | 1.0 | 0.0 | 13.0 |
| Nebraska | 6.0 | 3.0 | 6.0 | 20.0 |
| Nevada | 4.0 | 1.0 | 12.0 | 17.0 |
| New Hampshire | 2.0 | 0.0 | 10.0 | 13.0 |
| New Jersey | 5.0 | 1.0 | 20.0 | 87.0 |
| New Mexico | 1.0 | 0.0 | 9.0 | 22.0 |
| New York | 155.0 | 57.0 | 183.0 | 606.0 |
| North Carolina | 36.0 | 14.0 | 40.0 | 139.0 |
| North Dakota | 0.0 | 0.0 | 5.0 | 2.0 |
| Ohio | 66.0 | 47.0 | 84.0 | 199.0 |
| Oklahoma | 5.0 | 6.0 | 7.0 | 44.0 |
| Oregon | 20.0 | 0.0 | 15.0 | 81.0 |
| Pennsylvania | 103.0 | 9.0 | 78.0 | 341.0 |
| Rhode Island | 16.0 | 0.0 | 21.0 | 16.0 |
| South Carolina | 3.0 | 5.0 | 18.0 | 16.0 |
| South Dakota | 2.0 | 0.0 | 0.0 | 9.0 |
| Tennessee | 21.0 | 2.0 | 24.0 | 118.0 |
| Texas | 125.0 | 37.0 | 161.0 | 537.0 |
| Utah | 4.0 | 2.0 | 19.0 | 28.0 |
| Vermont | 0.0 | 0.0 | 1.0 | 2.0 |
| Virginia | 39.0 | 4.0 | 33.0 | 115.0 |
| Washington | 56.0 | 34.0 | 97.0 | 265.0 |
| West Virginia | 0.0 | 0.0 | 0.0 | 3.0 |
| Wisconsin | 12.0 | 3.0 | 10.0 | 66.0 |
| Wyoming | 0.0 | 0.0 | 0.0 | 1.0 |
| | | | | |

Melt

In [48]: melted_result = pd.melt(pivot_table_result.reset_index(), id_vars=["State"], var
print(melted_result)

| | State | Ship Mode | Order Count |
|-----|---------------|----------------|-------------|
| 0 | Alabama | First Class | 9.0 |
| 1 | Arizona | First Class | 42.0 |
| 2 | Arkansas | First Class | 10.0 |
| 3 | California | First Class | 302.0 |
| 4 | Colorado | First Class | 43.0 |
| • • | • • • | | |
| 191 | Virginia | Standard Class | 115.0 |
| 192 | Washington | Standard Class | 265.0 |
| 193 | West Virginia | Standard Class | 3.0 |
| 194 | Wisconsin | Standard Class | 66.0 |
| 195 | Wyoming | Standard Class | 1.0 |

[196 rows x 3 columns]

[Q9] What is the advantage of using melt?

ANS : เปลี่ยนรูปแบบข้อมูลจาก wide format เป็น long format ซึ่งช่วยให้การจัดการและการวิเคราะห์ ข้อมูลทำได้ง่ายขึ้น

[Q10] From the superstore_order, display the ascending order considering values in the 'Profit' column to group the 'Category'.

[Q11] Create a new column that calculates the total price (sale*quantity) before discount then group by 'product id' and 'category', then show the mean of the total price

```
In [52]: #enter your code here
    superstore_order['Total Price'] = superstore_order['Sales'] * superstore_order['
    result = superstore_order.groupby(['Product ID', 'Category'])['Total Price'].mea
    result
```

```
Out[52]: Product ID
                       Category
         FUR-BO-10000112 Furniture
                                      7426.566000
         FUR-BO-10000330 Furniture
                                    1258.192000
                                    1726.898000
         FUR-BO-10000362 Furniture
         FUR-BO-10000468 Furniture
                                     426.532400
         FUR-BO-10000711 Furniture
                                    3194.100000
                                        . . .
         TEC-PH-10004912 Technology
                                     747.320000
         TEC-PH-10004922 Technology
                                     673.249500
         TEC-PH-10004924 Technology
                                      57.149333
         TEC-PH-10004959 Technology
                                      412.009000
         TEC-PH-10004977 Technology
                                      2441.475429
         Name: Total Price, Length: 1846, dtype: float64
```

[Q12] Complete the function to apply ratio column that calculates from First Class and Standard Class columns on pivot_table_result

```
In [54]: # [Q12] Complete the function to apply `ratio` column that calculates from `Firs

# function to transform the ratio
def get_class_ratio(row):

# get the first class column
first_class = row['First Class']

# get the standard class column
standard_class = row['Standard Class']

# calculate the ratio
ratio = first_class/standard_class

return ratio

pivot_table_result["ratio"] = pivot_table_result.apply(get_class_ratio, axis=1)
pivot_table_result.head()
```

Out [54]: Ship Mode First Class Same Day Second Class Standard Class ratio

| - | Alabama | 9.0 | 1.0 | 18.0 | 30.0 | 0.300000 |
|----|-----------|-------|-------|-------|--------|----------|
| | Arizona | 42.0 | 15.0 | 22.0 | 123.0 | 0.341463 |
| A | Arkansas | 10.0 | 2.0 | 8.0 | 35.0 | 0.285714 |
| Ca | alifornia | 302.0 | 106.0 | 346.0 | 1000.0 | 0.302000 |
| C | olorado | 43.0 | 5.0 | 32.0 | 95.0 | 0.452632 |

[Q13] After complete Q12, What does the apply function do?

ANS : พึงก์ชัน apply ใช้ get_class_ratio ไปยังแต่ละแถวของ DataFrame

[Q14] Create a new column(short_ratio) that works the same as Q12 but with lambda function

```
In [56]: # [Q13] Create a new column(`short_ratio`) that works the same as Q11 but with `
    pivot_table_result["short_ratio"] = pivot_table_result.apply(lambda row: row['Fi
    pivot_table_result.head()
```

| Out[56]: | Ship Mode | First Class | Same Day | Second Class | Standard Class | ratio | short_ratio |
|----------|--------------|----------------|-------------|-----------------|-------------------|----------|-------------|
| | State | | | | | | |
| | Alabama | 9.0 | 1.0 | 18.0 | 30.0 | 0.300000 | 0.300000 |
| | Arizona | 42.0 | 15.0 | 22.0 | 123.0 | 0.341463 | 0.341463 |
| | Arkansas | 10.0 | 2.0 | 8.0 | 35.0 | 0.285714 | 0.285714 |
| | California | 302.0 | 106.0 | 346.0 | 1000.0 | 0.302000 | 0.302000 |
| | Colorado | 43.0 | 5.0 | 32.0 | 95.0 | 0.452632 | 0.452632 |

[Q15] What is the difference between using function in apply and lambda function? give 2 examples use case.

ANS:

Regular Function: ใช้พึงก์ชันมีตรรกะที่ซับซ้อนหรือมีหลายขั้นตอน และต้องการให้พึงก์ชันนั้นสามารถนำ ไปใช้ซ้ำในหลายๆครั้ง

Lambda Function: ใช้พึงก์ชันมีความง่ายและรวดเร็วในการคำนวณเพียงแค่หนึ่งบรรทัด โดยไม่ต้องการการ อธิบายหรือการใช้งานซ้ำ

| Out[61]: | | caseNumber | treatmentDate | statWeight | stratum | sex | diagnosis | bodyPart | disp |
|----------|---|-----------------------|----------------|-------------|-----------|---------|-------------------|-----------|------|
| | 0 | 150733174 | 11/07/2015 | 15.7762 | V | Male | 57 | 33 | |
| | 1 | 150734723 | 06/07/2015 | 83.2157 | S | Male | 57 | 34 | |
| | 2 | 150817487 | 02/08/2015 | 74.8813 | L | Female | 71 | 94 | |
| | 3 | 150717776 | 26/06/2015 | 15.7762 | V | Male | 71 | 35 | |
| | 4 | 150721694 | 04/07/2015 | 74.8813 | L | Female | 62 | 75 | |
| | 4 | | | | | | | | • |
| In [64]: | _ | 'Discount'] head() | = df.apply(lam | bda row: ro | w['statWe | eight'] | * 0.2 if r | ow['sex'] | == ' |
| Out[64]: | | caseNumber | treatmentDate | statWeight | stratum | sex | diagnosis | bodyPart | disp |
| | 0 | 150733174 | 11/07/2015 | 15.7762 | V | Male | 57 | 33 | |
| | 1 | 150734723 | 06/07/2015 | 83.2157 | S | Male | 57 | 34 | |
| | 2 | 150817487 | 02/08/2015 | 74.8813 | L | Female | 71 | 94 | |
| | 3 | 150717776 | 26/06/2015 | 15.7762 | V | Male | 71 | 35 | |
| | 4 | 150721694 | 04/07/2015 | 74.8813 | L | Female | 62 | 75 | |
| | 4 | | | | | | | | • |

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