Handling missing values in Python

September 25, 2019

0.1 Preparation

```
In [20]: import numpy as np
    import pandas as pd
    PREVIOUS_MAX_ROWS=pd.options.display.max_rows
    pd.options.display.max_rows=20
    np.random.seed(123)
    import matplotlib.pyplot as plt
    plt.rc('figure',figsize=(10,6))
    np.set_printoptions(precision=4, suppress=True)
```

0.2 Identify missing strings

```
In [21]: string_data=pd.Series(['annie','bob',np.nan,'donald'])
         string_data
         string_data.isnull()
Out[21]: 0
              False
         1
              False
         2
               True
         3
              False
         dtype: bool
In [22]: string_data[0]=None #change the first element to missing
         string_data.isnull()
Out[22]: 0
               True
              False
         2
               True
              False
         dtype: bool
```

0.3 Treating missing values

```
Out[23]: 0
            1.0
             3.0
        2
            5.0
        dtype: float64
In [24]: data[data.notnull()] #show nonmissing observations
Out[24]: 0
             1.0
             3.0
            5.0
        dtype: float64
In [25]: data=pd.DataFrame([ [1.,6.5,3.], [1.,NA,NA],
                         [NA,NA,NA], [NA,6.5,3.]])
        cleaned=data.dropna() #drops the entire row with missing observations
        data
Out[25]: 0
                 1
        0 1.0 6.5 3.0
        1 1.0 NaN NaN
        2 NaN NaN NaN
        3 NaN 6.5 3.0
In [26]: cleaned
Out[26]: 0 1 2
        0 1.0 6.5 3.0
In [27]: data.dropna(how='all') #do not show rows where all elements are NA
Out[27]: 0
               1
        0 1.0 6.5 3.0
        1 1.0 NaN NaN
        3 NaN 6.5 3.0
In [30]: data[4]=NA #add a fourth column with all missing values
        data
        data.dropna(axis=1,how='all')
Out[30]:
            0
                 1
        0 1.0 6.5 3.0
        1 1.0 NaN NaN
        2 NaN NaN NaN
        3 NaN 6.5 3.0
In [47]: df=pd.DataFrame(np.random.randn(7,3)) # a 7x3 matrix with random values
        df.iloc[:4,1]=NA
        df.iloc[:2,2]=NA
        df
```

```
Out[47]:
                            1
        0 -0.314758
                          NaN
                                    NaN
        1 -1.212523
                          NaN
                                    NaN
        2 1.150206
                          NaN 0.181035
        3 1.177862
                          NaN 1.031114
        4 -1.084568 -1.363472 0.379401
        5 -0.379176  0.642055 -1.977888
        6 0.712265 2.598304 -0.024626
In [48]: df.dropna() #drops rows with any missing value
Out[48]:
        4 -1.084568 -1.363472 0.379401
        5 -0.379176  0.642055 -1.977888
        6 0.712265 2.598304 -0.024626
In [49]: df.dropna(thresh=2) #drops rows with missing values in two columns
Out [49]:
                            1
        2 1.150206
                          NaN 0.181035
        3 1.177862
                          NaN 1.031114
        4 -1.084568 -1.363472 0.379401
        5 -0.379176  0.642055 -1.977888
        6 0.712265 2.598304 -0.024626
0.4 Interpolating missing data
In [46]: df.fillna(0) #change missing to zero
Out [46]:
        0 -1.326265 0.000000 0.000000
        1 0.045490 0.000000 0.000000
        2 0.199524 0.000000 -0.831155
        3 1.162204 0.000000 -2.123100
        4 1.039727 -0.403366 -0.126030
        5 -0.837517 -1.605963 1.255237
        6 -0.688869 1.660952 0.807308
In [37]: df.fillna({1:0.5,2:0}) #change column 1's missing to 0.5, column 2's missing to zero
Out [37]:
                            1
        0 -1.085631 0.500000 0.000000
        1 -1.506295 0.500000 0.000000
        2 -2.426679 0.500000 1.265936
        3 -0.866740 0.500000 -0.094709
        4 1.491390 -0.638902 -0.443982
        5 -0.434351 2.205930 2.186786
        6 1.004054 0.386186 0.737369
```

```
In [50]: _=df.fillna(0,inplace=True) #sets equal to zero
        df
Out [50]:
                  0
        0 -0.314758  0.000000  0.000000
         1 -1.212523 0.000000 0.000000
        2 1.150206 0.000000 0.181035
        3 1.177862 0.000000 1.031114
        4 -1.084568 -1.363472 0.379401
        5 -0.379176  0.642055 -1.977888
        6 0.712265 2.598304 -0.024626
In [51]: df=pd.DataFrame(np.random.randn(6,3)) #another random matrix
        df.iloc[2:,1]=NA
        df.iloc[4:,2]=NA
        df
Out[51]:
                   0
                             1
        0 0.034142 0.179549 -1.861976
        1 0.426147 -1.605410 -0.427680
        2 1.242870
                          NaN 0.501249
        3 1.012739
                          NaN -1.370948
        4 -0.332475
                          {\tt NaN}
                                     NaN
        5 -0.275786
                          NaN
                                     NaN
In [53]: df.fillna(method='ffill') # fill out the missing values using forward fill
Out [53]:
                             1
        0 0.034142 0.179549 -1.861976
         1 0.426147 -1.605410 -0.427680
        2 1.242870 -1.605410 0.501249
        3 1.012739 -1.605410 -1.370948
        4 -0.332475 -1.605410 -1.370948
        5 -0.275786 -1.605410 -1.370948
In [54]: df.fillna(method='ffill',limit=2) # forward fill up to two steps
Out [54]:
                             1
        0 0.034142 0.179549 -1.861976
         1 0.426147 -1.605410 -0.427680
        2 1.242870 -1.605410 0.501249
         3 1.012739 -1.605410 -1.370948
        4 -0.332475
                          NaN -1.370948
        5 -0.275786
                          NaN -1.370948
In [56]: data=pd.Series([1.,NA,3.5,NA,7]) # a vector with missing values
        data
Out [56]: 0
              1.0
         1
             NaN
```

```
2 3.5
3 NaN
4 7.0
dtype: float64
```

In [57]: data.fillna(data.mean()) # replace by mean