# 数据预处理-2

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## 作业3

### 一些额外的Trick:

#比如想要产生新的一列来统计同一时间内所有地方的PM指数和 df['sumOfPM']=df['PM\_US\_POST']+df['PM\_Nongzhanguan']+df['PM\_Dongsi']+df['PM\_Dongsihuan']

#比如想要按照某一列排序,升序 def top(df,n=5,colunm='PRES'): return df.sort\_values(by=column)[-n:]

#apply函数是 pandas 里面所有函数中自由度最高的函数。该函数如下: DataFrame.apply(func, axis=0, broadcast=False, raw=False...) 该函数最有用的是第一个参数,这个参数是函数,相当于C/C++的函数指针。

#这次预处理实验和上次预处理实验后,我明显感受到直接按列操作比遍历每一个csv中的cell效率高得多,减小了IO时间

In [1]: import numpy as np import pandas as pd import time import matplotlib pyplot as plt

df = pd.read\_csv("BeijingPM20100101\_20151231.csv",encoding='utf-8') df.describe()

Out[1]: PM\_Dongsi PM\_Dongsihuan PM\_Nongzhanguan PM\_US Post DEWP HUMI PRES TEMP lws precipitation 92.560806 88.643737 95.904241 2.074554 54.602421 1016.465442 12.587040 23.261829 19.25868 88.041166 91.643772 14.222059 25.991338 10.295070 12.098527 49.281706 4381.03553. 87.239267 88.027434 3.000000 3.000000 3.000000 1.000000 -40.000000 2.000000 991.000000 -19.000000 0.450000 0.00000 24.000000 28.000000 24.000000 27.000000 -10.000000 31.000000 1008.000000 2.000000 1.790000 0.00000 64 000000 68 000000 62 000000 69 000000 2 000000 55 000000 1016 000000 14 000000 4 920000 0.00000 124.000000 122.000000 132.000000 15.000000 78.000000 1025.000000 23.000000 21.020000 0.00000 127.000000 737.000000 672.000000 844.000000 994.000000 28.000000 100.000000 1046.000000 42.000000 585.600000 999990.00000

```
In [2] todo = [HUMI', 'PRES', 'TEMP']
#进行线性插值
for each in todo:
df[each]=df[each].interpolate(method="linear")
print("将dataframe其中一列抽出来是什么类型?{}".format(type(df['HUMI'])))
for each in todo:
mean = df[each].mean()
stdDoub = df[each].mean()
stdDoub = df[each].apply(lambda x : mean + stdDoub*2 if x>mean + stdDoub * 2
else ( mean-2 * stdDoub if x < mean - 2 * stdDoub else x))
```

将dataframe其中一列抽出来是什么类型?<class 'pandas.core.series.Series'>

```
处理前
    PM_Dongsi PM_Dongsihuan PM_Nongzhanguan
90 287.0 443.0 326.0
30590
30591
         495.0
                      508.0
                                   512.0
30592
                      513.0
                                   513.0
                     513.0
537.0
505.0
                                   501.0
530.0
508.0
30593
          485 0
30595
         507.0
30596
          487.0
                      443.0
                                   475.0
357.0
                      351.0
         345.0
30597
30598
          34 0
                                  28.0
          30.0
                      40.0
                                  24.0
处理后
    PM_Dongsi PM_Dongsihuan PM_Nongzhanguan
30590
                                   326.0
500.0
         287.0
                      443.0
          495.0
                      500.0
30591
                                   500.0
500.0
30592
         495.0
                      500.0
30593
30594
          500.0
                      500.0
                                   500.0
30596
         487 0
                      443 0
                                   475.0
30597
         345.0
                      351.0
                                   357.0
                                  28.0
30598
          34.0
                      39.0
30599
          30.0
                      40.0
                                  24.0
```

```
In [4]: #下面第一种方式比较直观但如果出现两个连续的cv,怎么处理?
# Exce里面也告诉我们的确出现了
# for i in range(len(df[cbwd]]):
# if df[cbwd]]] == cv:
# dfat[ii, cbwd] = df[cbwd][i+1]
print("对C处理前")
print("对C处理后")
#TODO采用的训替换方式
df[cbwd]"-df[cbwd] replace[cv| method="hfill")
        df['cbwd']=df['cbwd'].replace('cv',method='bfill')
print(df.iloc[20:25,8:15])
        对CV处理前
           PM_Nongzhanguan PM_US Post DEWP HUMI PRES TEMP cbwd
                   ggznanguan PM_UD Post DEWP HUMI PK
NAN NAN 1-77.0 38.0 1017.0 -5.0 cv
NaN NaN -17.0 38.0 1018.0 -5.0 NW
NaN NaN -17.0 38.0 1018.0 -5.0 NW
NaN 129.0 -17.0 41.0 1020.0 -5.0 cv
NaN 148.0 -16.0 38.0 1020.0 -4.0 SE
        20
21
        对CV处理后
          9CVSC地田市

PM_Nongzhanguan PM_US Post DEWP HUMI PRES TEMP cbwd

0 NaN NaN -17.0 38.0 1017.0 -5.0 NW

1 NaN NaN -17.0 38.0 1018.0 -5.0 NW

2 NaN NaN -17.0 38.0 1018.0 -5.0 NW

3 NaN 129. -17.0 41.0 1020.0 -5.0 SE

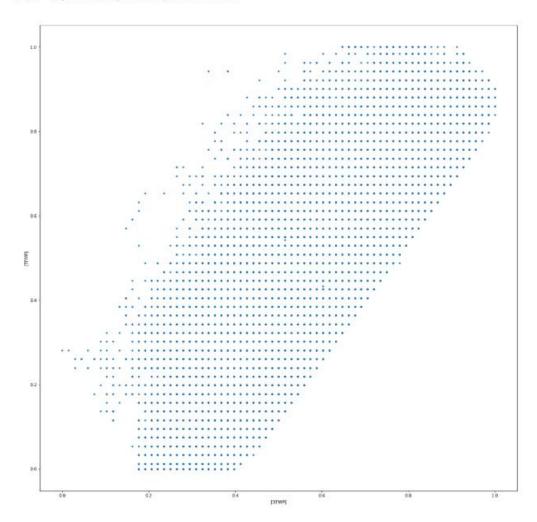
4 NaN 148.0 -16.0 38.0 1020.0 -4.0 SE
        21
        22
23
In [5]: df.describe()
Out[5]: PM_Dongsi PM_Dongsihuan PM_Nongzhanguan PM_US Post
                                                                                                                       PRES
                                                                                                                                       TEMP
                                                                                       DEWP
                                                                                                       HUMI
                                                                                                                                                        lws precipitat
        25052.000000 20508.000000 24931.000000 50387.000000 52579.000000 52584.000000 52584.000000 52584.000000 52579.000000 52100.0000
             88.909788
                               92.373464
                                                    88.409570
                                                                    95.904241
                                                                                   2.074554
                                                                                                   54.851158 1016.517506
                                                                                                                                   12.599689
                                                                                                                                                   23.261829
                                                                                                                                                                    19.2586
                          87.056715 86.760055 91.643772 14.222059
                                                                                                    26.095084 10.224785
        85.898308
                                                                                                                                   12.063515
                                                                                                                                                   49.281706 4381.035
              3.000000
                                3.000000
                                                      3.000000
                                                                      1.000000
                                                                                   -40.000000
                                                                                                     2.660485
                                                                                                                  995.931255
                                                                                                                                  -11.609537
                                                                                                                                                    0.450000
                                                                                                                                                                     0.0000
            24.000000 28.000000 24.000000 27.000000 -10.000000 32.000000 1008.000000 2.000000
                                                                                                                                                    1.790000
                                                                                                                                                                    0.0000
             64.000000
                               68.000000
                                                     62.000000
                                                                    69.000000
                                                                                    2 000000
                                                                                                    55.000000 1016.000000
                                                                                                                                   14 000000
                                                                                                                                                    4 920000
                                                                                                                                                                     0.0000
        124.000000 127.000000 122.000000 132.000000 15.000000 78.000000 125.000000 23.000000 21.020000 0.0000
                           500.000000
                                                   500.00000 994.00000 28.00000 100.00000 1037.147438 36.783754 585.60000 999990.0000
           500.000000
       4
                                                                                                                           In [6]: df.to_csv("result.csv")
```

比较前后的描述信息,可以较为明显地看出有关标准差采取的措施带来的改变

# 作业4

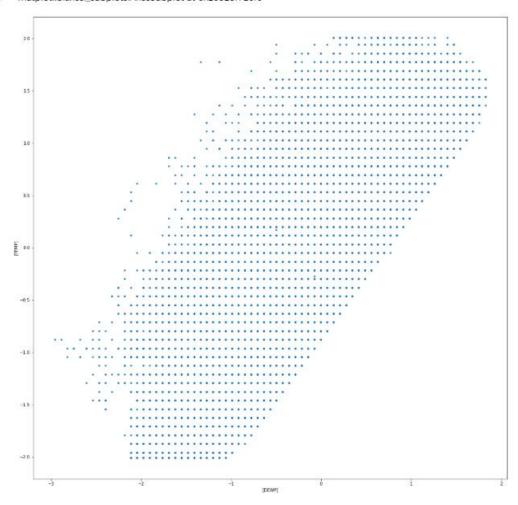
```
In [7]: from sklearn.preprocessing import MinMaxScaler scaler = MinMaxScaler() features = [DEWP, TEMP] df[features] = scaler.fit_transform(df[features]) df.head() df.plot.scatter(x=[DEWP].y=[TEMP].s=10.figsize = (20.20))
```

Out[7]: <matplotlib.axes\_subplots.AxesSubplot at 0x1e91b7291d0>



```
In [8]: from sklearn.preprocessing import StandardScaler ss = StandardScaler() features = ['DEWP', 'TEMP'] df[features] = ss.fit_transform(df[features]) df.head() df.plot.scatter(x=['DEWP'].y=['TEMP'].s=10.figsize = (20.20))
```

Out[8]: <matplotlib.axes\_subplots.AxesSubplot at 0x1e918f71ef0>



# In [9]: sections = [0.50.100.150.200,300.1200] #划分为不同长度的区间 section\_names=["green","yellow","orange","red","purple", "Brownish red"] #设置每个区间的标签 # df = df.fillna(df.mean()) result = pd.cut(df.PM\_Dongsi.sections.labels=section\_names) print(pd.value\_counts(result))

green 10576
yellow 6268
orange 3578
red 1942
purple 1910
Brownish red 778
Name PM Dongs i dwo

Name: PM\_Dongsi, dtype: int64