

互评作业1：数据探索性分析与数据预处理：Chicago Building Violations

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1. 数据集：wine-reviews

1个csv文件：

- building-violations.csv

读取数据：

```
In [18]:
```

```
%matplotlib inline
import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
import seaborn as sns
```

```
In [19]:
```

```
dirpath_cbv = "ChicagoBuildingViolations-dat
```

In [21]:

```
data_cbv = pd.read_csv(dirpath_cbv)
data_cbv.head()
```

	ID	VIOLATION LAST MODIFIED DATE	VIOLATION DATE	
0	6392482	2019-12-04T12:40:09.000	2019-12-04T00:00:00.000	(
1	6392480	2019-12-04T12:40:09.000	2019-12-04T00:00:00.000	(
2	6392335	2019-12-04T14:00:12.000	2019-12-04T00:00:00.000	(
3	6391883	2019-12-04T08:32:01.000	2019-12-04T00:00:00.000	(
4	6392369	2019-12-04T14:14:24.000	2019-12-04T00:00:00.000	(
5 rows × 32 columns				

显示数据信息

In [22]:

```
data_cbv = data_cbv.drop(columns=['ID', 'VIO
```

```
data_cbv.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1677788 entries, 0 to 1677787
Data columns (total 29 columns):
#   Column                                     Non-Null Count  Dtype
---  -
0   VIOLATION LAST MODIFIED DATE             1677788 non-null object
1   VIOLATION DATE                           1677788 non-null object
2   VIOLATION CODE                           1677788 non-null object
3   VIOLATION STATUS                         1677788 non-null object
4   VIOLATION STATUS DATE                    641589 non-null object
5   VIOLATION LOCATION                       780506 non-null object
6   VIOLATION ORDINANCE                      1630207 non-null object
7   INSPECTOR ID                            1677788 non-null object
8   INSPECTION NUMBER                        1677788 non-null int64
9   INSPECTION STATUS                        1677772 non-null object
10  INSPECTION WAIVED                        1677788 non-null object
11  INSPECTION CATEGORY                      1677788 non-null object
12  DEPARTMENT BUREAU                       1677788 non-null object
13  ADDRESS                                  1677788 non-null object
```

```

14  STREET NUMBER          167
7788 non-null  int64
15  STREET DIRECTION      167
7788 non-null  object
16  STREET NAME           167
7788 non-null  object
17  STREET TYPE           166
4247 non-null  object
18  PROPERTY GROUP        167
7788 non-null  int64
19  SSA                   321
521 non-null   float64
20  LATITUDE              167
6278 non-null  float64
21  LONGITUDE             167
6278 non-null  float64
22  LOCATION              167
6278 non-null  object
23  Community Areas       167
5509 non-null  float64
24  Zip Codes             167
6278 non-null  float64
25  Boundaries - ZIP Codes 167
5509 non-null  float64
26  Census Tracts         167
6243 non-null  float64
27  Wards                 167
5509 non-null  float64
28  Historical Wards 2003-2015 167
5509 non-null  float64
dtypes: float64(9), int64(3), object(17)
memory usage: 371.2+ MB

```

2. 数据分析

2.1 数据可视化和摘要

2.1.1 标称属性

```
In [23]:
```

```
title = ['VIOLATION LAST MODIFIED DATE', 'VI
```

```
In [24]:
```

```
# 由于本数据集属性太多，这里仅显示violation statu  
title[3]
```

```
'VIOLATION STATUS'
```

```
In [25]:
```

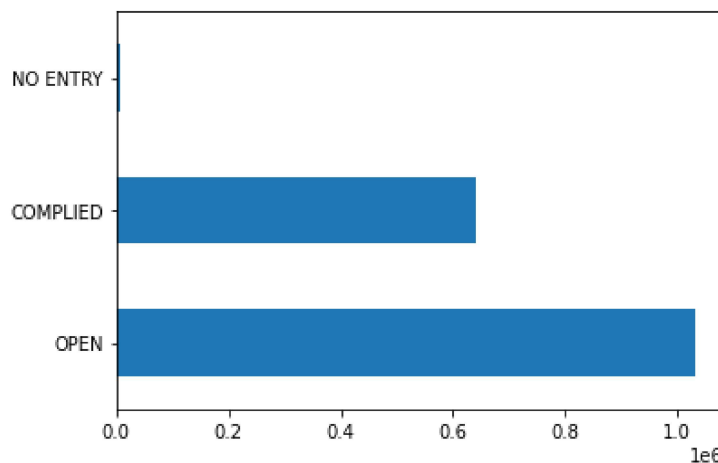
```
getattr(data_cbv, title[3]).value_counts()
```

```
OPEN          1030958  
COMPLIED      641247  
NO ENTRY       5583  
Name: VIOLATION STATUS, dtype: int64
```

```
In [26]:
```

```
data_cbv[title[3]].value_counts().head(10).p
```

<AxesSubplot:>



2.2.2 数值属性

直接调用describe函数给出数据的基本统计量

```
In [27]:  
  
data_cbv.describe()
```

	INSPECTION NUMBER	STREET NUMBER	PROPER GROU
count	1.677788e+06	1.677788e+06	1.677788e+
mean	8.049798e+06	4.150382e+03	2.020547e+
std	4.555757e+06	2.893493e+03	1.862796e+
min	2.655750e+05	1.000000e+00	1.000000e+
25%	2.304416e+06	1.648000e+03	2.056000e+
50%	1.041875e+07	3.747000e+03	1.543230e+
75%	1.168728e+07	6.228000e+03	3.669840e+
max	1.305092e+07	1.377000e+04	6.779750e+

数据可视化

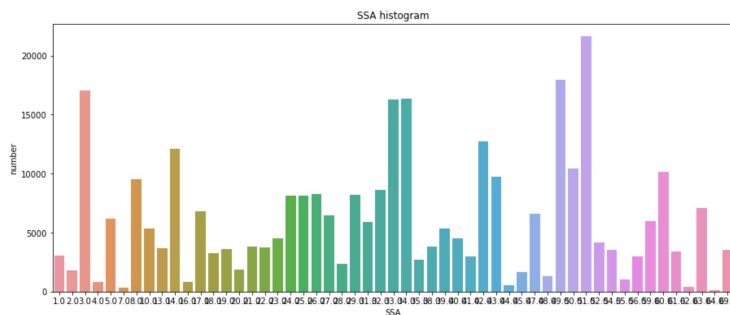
直方图：

In [28]:

```
def histogram(data, x, y ,title):
    plt.figure(figsize = (15,6))
    plt.title(title)
    sns.set_color_codes("pastel")
    sns.barplot(x=x, y=y, data=df)
    locs, labels = plt.xticks()
    plt.show()

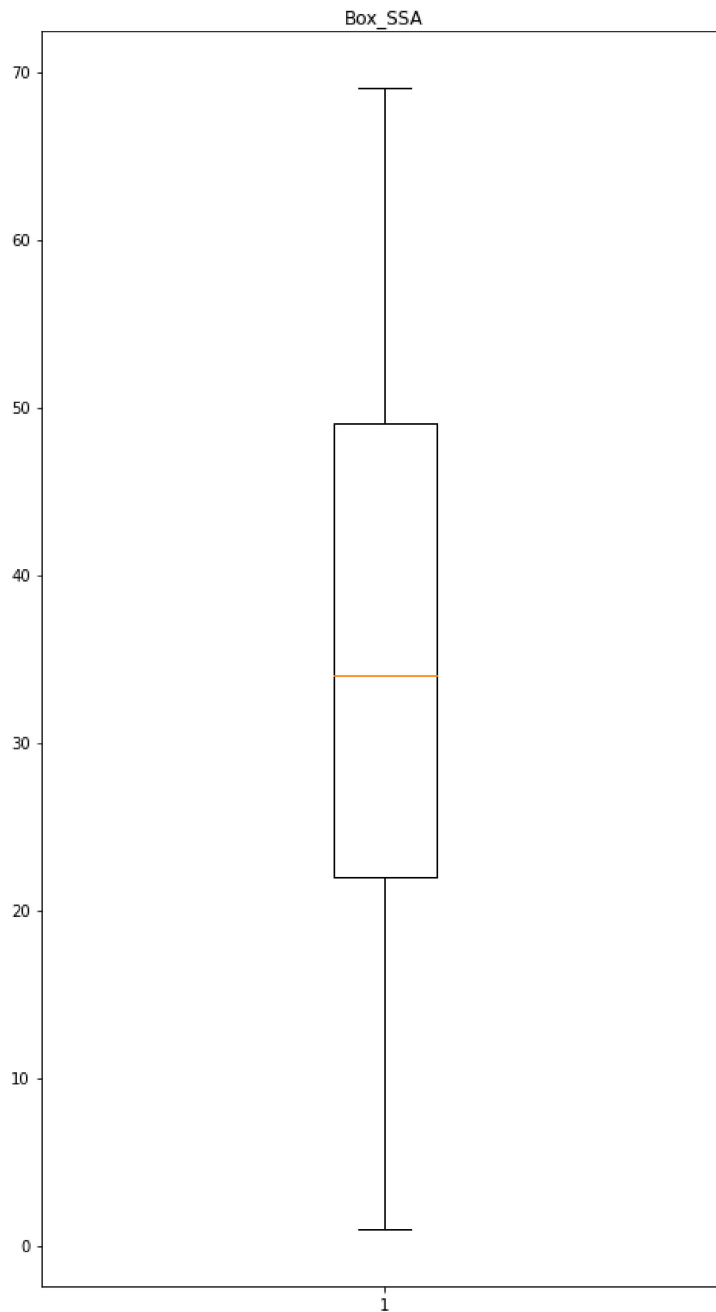
#SSA
temp = data_cbv['SSA'].value_counts()
df = pd.DataFrame({'SSA':temp.index, 'number'
                    :temp.values})

histogram(df, 'SSA', 'number', 'SSA histogram')
```



盒图:


```
In [30]:  
  
fig = plt.figure(figsize=(8, 15))  
plt.boxplot(data_cbv['SSA'].loc[data_cbv['SS  
t = plt.title('Box_SSA')
```



3. 数据缺失处理

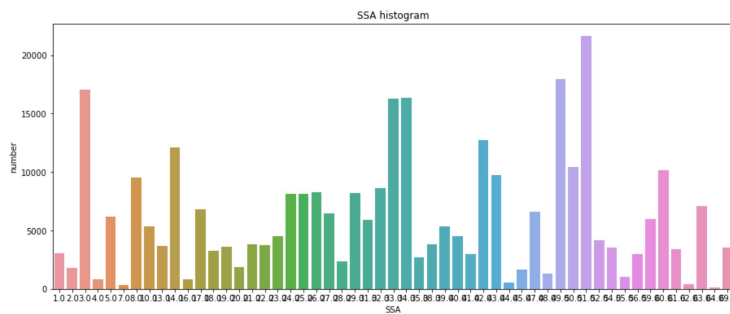
3.1 将缺失部分剔除

In [31]:

```
data_delete = data_cbv['SSA'].dropna()

temp = data_delete.value_counts()
df = pd.DataFrame({'SSA':temp.index, 'number':temp.values})

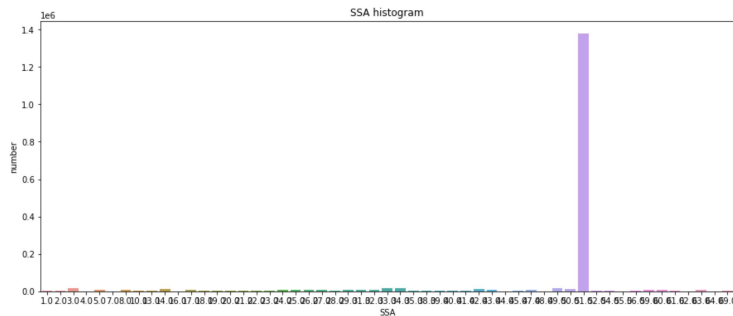
histogram(df, 'SSA', 'number', 'SSA histogram')
```



3.2 用最高频率来填补缺失值

In [34]:

```
data_most = data_cbv['SSA'].fillna(data_cbv[
temp = data_most.value_counts()
df = pd.DataFrame({'SSA':temp.index, 'number
histogram(df, 'SSA', 'number', 'SSA histogram
```



3.3 通过属性的相关关系来填补缺失值

使用纬度LATITUDE填补经度LONGITUDE

In [35]:

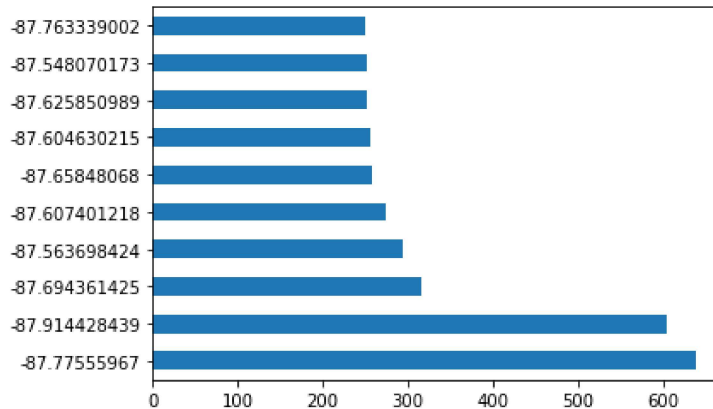
```
data_fill = pd.DataFrame(data_cbv, columns=[
data_fill.head(10)
```

	LATITUDE	LONGITUDE
0	41.749169	-87.602551
1	41.749169	-87.602551
2	41.711751	-87.537842
3	41.844521	-87.712416
4	41.753908	-87.562784
5	41.806815	-87.611539
6	41.753908	-87.562784
7	41.749169	-87.602551
8	41.711751	-87.537842
9	41.748732	-87.659904

In [36]:

```
data_fill['LONGITUDE'].value_counts().head(1)
```

<AxesSubplot:>



3.4 通过数据对象之间的相似性来填补缺失值

In [38]:

```
data_sim = data_cbv[['LATITUDE', 'LONGITUDE']]
point2price = {}
for row in data_sim.iterrows():
    if point2price.get(row[1]['LONGITUDE'],
                        if not pd.isnull(row[1]['LATITUDE'])
                        point2price[row[1]['LONGITUDE']]
                        point2price[row[1]['LONGITUDE']]
    else:
        if not pd.isnull(row[1]['LATITUDE'])
        point2price[row[1]['LONGITUDE']]
```

In [39]:

```
for k in point2price.keys():
    point2price[k][0] = round(point2price[k]
```

In [40]:

```
for row in data_sim.iterrows():
    if pd.isnull(row[1]['LATITUDE']):
        try:
            row[1]['LATITUDE'] = point2price
        except:
            continue
```

In [*]:

```
#对被填充后的 $price$ 画直方图
temp = data_sim['LATITUDE'].value_counts()
df = pd.DataFrame({'LATITUDE':temp.index, 'n':temp.values})

histogram(df, 'LATITUDE', 'number', 'LATITUDE')
```

In []: