A simple Data Interview Assignment_ Yiting Duan

You can find my data on Kaggle-yelp

```
Data Exploration && Data clearning
import numpy as np
import pandas as pd
data=pd.DataFrame(pd.read_excel("pandas处理json.xlsx"))
 import pandas as pd
 import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
sns.set_style('whitegrid')
 import warnings
 import gc
warnings.simplefilter(action='ignore', category=FutureWarning) warnings.simplefilter(action='ignore', category=DeprecationWarning) %matplotlib inline
 def null_values(df):
              il_values(df):
    mis_val = df. isnull().sum()
    mis_val_percent = 100 * df. isnull().sum() / len(df)
    mis_val_percent = 100 * df. isnull().sum() / len(df)
    mis_val_table = pd. concat([mis_val, mis_val_percent], axis=1)
    mis_val_table_ren_columns = mis_val_table_rename(
    columns = (0 : 'Missing Values', 1 : '% of Total Values'))
    mis_val_table_ren_columns = mis_val_table_ren_columns[
        mis_val_table_ren_columns : lioc[i,1] != 0]. sort_values(
'% of Total Values', ascending=False). round(1)
    print ('Dataframe has " + str(df. shape[1]) + " columns. \n"
        "There are " + str(mis_val_table_ren_columns. shape[0]) +
        " columns that have missing values.")
    return mis val table ren columns
               return mis_val_table_ren_columns
missing_rate = null_values(data.dropna())
      Dataframe has 16 columns.
      There are 0 columns that have missing values.
                                                                          8880 E
Via
                                                           Carlos
                                                                                                                                                                                                                                                {'GoodForKids': 
'ByAppointmentOr
 1 Yzvjg0SayhoZgCljUJRF9Q
                                                            Santo,
NMD
                                                                         Linda,
Ste 107
                                                                                        Scottsdale
                                                                                                          AZ
                                                                                                                               85258 33.569404 -111.890264 5.0
                                                                                                                                                                                                              4.0
                                                                                                                                                                                                                              1.0
                                                                        3554
Rue
Notre-
Dame O
 2 XNoUzKckATkOD1hP6vghZg
                                                                                          Montreal QC
                                                                                                                           H4C 1P4 45.479984 -73.580070 5.0
                                                                                                                                                                                                             5.0
                                                                                                                                                                                                                              1.0
                                                       Nevada
House of
Hose
                                                                             1015
                                                                                                                                                                                                                             0.0 {'BusinessAcceptsCreditC
                                                                                        North Las
Vegas
          6OAZjbxqM5ol29BuHsil3w
                                                                                                                              89030 36.219728 -115.127725 2.5
                                                                            Sharp
Cir
                                                       USE MY
                                                                          4827 E
 4 51M2Kk903DFYl6gnB5l6SQ SERVICES LLC
                                                                                                                                                                                                                              1.0 {'BusinessAcceptsCredit(
'True', 'By/
                                                                                                                               85205 33 428065 -111 726648 4 5
                                                                       Downing
Cir
                                                                                              Mesa A7
                                                                                                                                                                                                            26.0
```

```
data.duplicated()

0 False
1 False
2 False
3 False
4 False
----
1048570 True
1048571 True
1048572 True
1048573 True
1048574 True
1048574 True
Length: 1048575, dtype: bool

data.drop_duplicates()
```

data.isnull().sum()

839182 839183 847861 839184 839182 839691 839182 business_id name address address city state postal_code latitude longitude 839182 839182 839182 839182 839182 stars review_count review_count
is_open
attributes
categories
hours
number_categories
number_days
dtype: int64 839182 868227 839706 884025 839706 884025

data['address'].isnull().value_counts()

True 847861 False 200714 Name: address, dtype: int64

data['name']=data['name'].dropna()

data.isnull().sum()

business_id 839182 839183 847861 839184 839182 839691 839182 839182 839182 839182 839182 839182 839182 839182 839182 839182 839182 name address city state state
postal_code
latitude
longitude
stars
review_count
is_open
attributes
categories
hours
number_categories
number_days
dtype: int64 839706 884025

data.dropna(axis=0, subset = ["stars"])

	business_id	name	address	city	state	postal_code	latitude	longitude	stars	review_count	is_open	
0	f9NumwFMBDn751xgFiRbNA	The Range At Lake Norman	10913 Bailey Rd	Cornelius	NC	28031	35.462724	-80.852612	3.5	36.0	1.0	{'BusinessAcce
1	Yzvjg0SayhoZgCljUJRF9Q	Carlos Santo, NMD	8880 E Via Linda, Ste 107	Scottsdale	AZ	85258	33.569404	-111.890264	5.0	4.0	1.0	{'Gooc 'ByAppo
2	XNoUzKckATkOD1hP6vghZg	Felinus	3554 Rue Notre- Dame O	Montreal	QC	H4C 1P4	45.479984	-73.580070	5.0	5.0	1.0	
3	6OAZjbxqM5ol29BuHsil3w	Nevada House of Hose	1015 Sharp Cir	North Las Vegas	NV	89030	36.219728	-115.127725	2.5	3.0	0.0	{'BusinessAcce
4	51M2Kk903DFYl6gnB5l6SQ	USE MY GUY SERVICES LLC	4827 E Downing Cir	Mesa	AZ	85205	33.428065	-111.726648	4.5	26.0	1.0	("BusinessAcce
	***	(707)		V575V		5751	85558		511		:222	
209388	9Q0fPWAjUweoFDk0kafuzQ	Nishi Sushi	9750 Weston Road	Vaughan	ON	L4H 2P2	43.838555	-79.559823	4.0	5.0	0.0	{'Ambience': "{'r

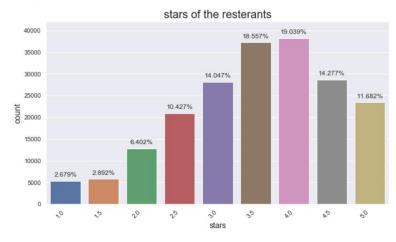
```
data.isnull().sum()
business_id
                          839182
                          839183
name
address
                          847861
839184
city
state
                          839182
postal_code
                          839691
latitude
longitude
                          839182
                          839182
stars
                          839182
839182
review_count
is_open
attributes
                          868227
839706
categories
                          884025
839706
number_categories
number_days
dtype: int64
                          884025
```

```
data['stars'].value_counts()

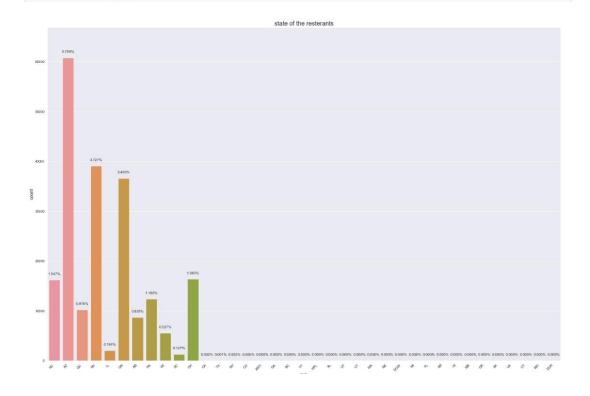
4.0 39199
3.5 38079
4.5 29940
3.0 28634
5.0 27080
2.5 21435
2.0 13124
1.5 6004
1.0 5898
Name: stars, dtype: int64
```

plot

(0, 42034.3)



```
| State | Stat
```

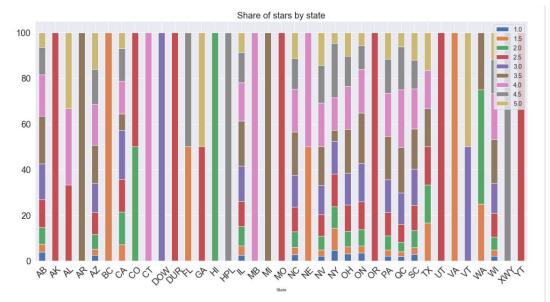


```
#crosstab
tab=pd.crosstab(data['state'],
data['stars'],
normalize='index') * 100
```

tab.head()

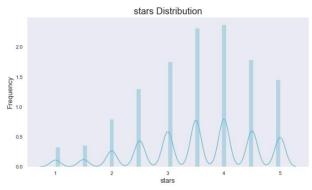
stars	1.0	1.5	2.0	2.5	3.0	3.5	4.0	4.5	5.0
state									
AB	3.844340	3.478774	7.334906	12.323113	15.648585	20.813679	18.054245	12.051887	6.450472
AK	0.000000	0.000000	0.000000	100.000000	0.000000	0.000000	0.000000	0.000000	0.000000
AL	0.000000	0.000000	0.000000	33.333333	0.000000	0.000000	33.333333	0.000000	33.333333
AR	0.000000	0.000000	0.000000	0.000000	0.000000	100.000000	0.000000	0.000000	0.000000
AZ	2.466786	2.715227	6.353737	9.708567	12.829052	16.492230	18.005779	15.294076	16.134546

```
fig, ax = plt.subplots(1, 1, figsize=(20, 10))
g = tab.plot(kind='bar', stacked=True, rot='45', ax=ax)
ax.set_xlabel('State', fontsize=10)
ax.sick_params(axis='both', labelsize=20)
ax.set_title('Share of stars by state', fontsize=20)
plt.legend(fontsize=14)
```



```
]: plt.figure(figsize=(11,6))
sns.set(style='dark')
g = sns.distplot(data["stars"], color='c')
g.set_ylabel("frequency", fontsize=15)
g.set_ylabel("frequency", fontsize=16)
g.set_title("stars Distribution", fontsize=20)
```

Text(0.5, 1.0, 'stars Distribution')



```
plt.figure(figsize=(11,6))

g2 = sns.boxplot(x="state", y="stars", data=data, order = data['state'].value_counts().index,color='r')

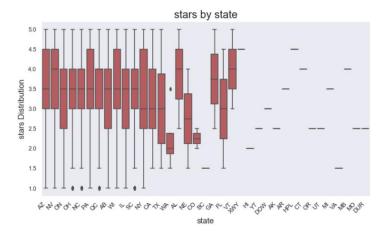
g2.set_xticklabels(g2.get_xticklabels(),fontdict=('horizontalalignment':'right'),rotation=45)

g2.set_xlabel("state", fontsize=15)

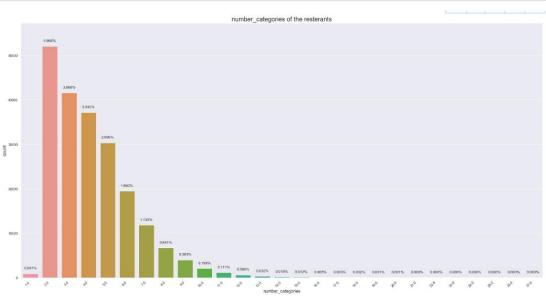
g2.set_ylabel("stars Distribution", fontsize=15)

g2.set_ylabel("stars by state", fontsize=20)

plt.show()
```



```
data.isnull().sum()
business_id
name
address
                                 1
0
1
0
city
state
postal_code
                               109
0
0
0
0
latitude
 longitude
stars
review_count
is_open
attributes
                           27215
496
categories
hours
number_categories
                           42719
496
number_days
dtype: int64
                           42719
```

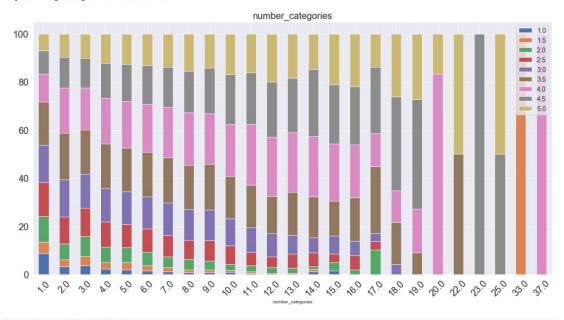


tab.head()

stars	1.0	1.5	2.0	2.5	3.0	3.5	4.0	4.5	5.0
number_categories									
1.0	9.330406	4.939627	10.208562	13.830955	14.928650	17.672887	11.416026	10.428101	7.244786
2.0	3.612029	2.974499	6.509717	10.982026	15.001152	18.943467	18.401951	12.623857	10.951302
3.0	3.715103	3.945945	8.024142	11.681535	13.872124	17.928680	17.207300	12.268257	11.356914
4.0	2.539546	2.687507	6.284300	10.139352	13.434843	18.212633	18.618853	14.535134	13.547832
5.0	2.130883	3.016830	6.053420	9.432533	13.177222	17.728815	19.121958	15.410203	13.928136

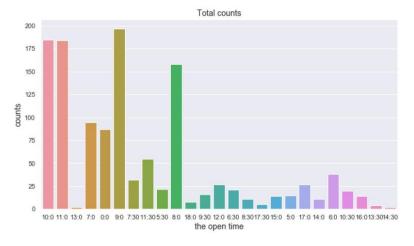
```
fig, ax = plt.subplots(1, 1, figsize=(20, 10))
g = tab.plot(kind='bar', stacked=True, rot='45', ax=ax)
ax.set_xlabel('number_categories', fontsize=10)
ax.tick_params(axis='both', labelsize=20)
ax.set_title('number_categories', fontsize=20)
plt.legend(fontsize=14)
```

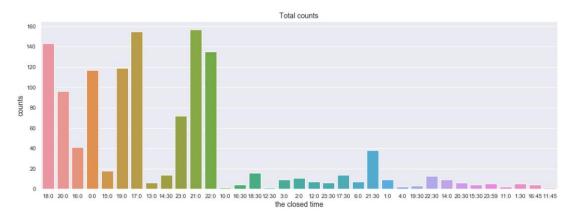
(matplotlib.legend.Legend at 0x1d02a2e6408>

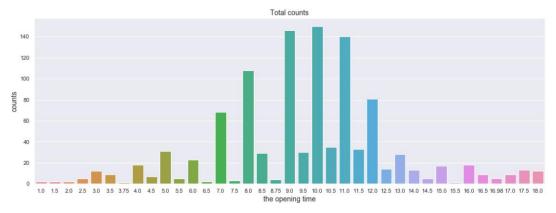


hours=data.loc[:,['hours']].values df=pd.DataFrame(hours)

0 {Monday': '10.0-18.0', 'Tuesday': '11.0-20.0'...
1 {Monday': '7.0-16.0', 'Tuesday': '7.0-16.0', ...
2 {Monday': '0.0-0.0', 'Tuesday': '9.0-16.0', .'..
3 {Monday': '7.0-18.0', 'Tuesday': '7.0-18.0', ...
4 {Monday': '7.0-19.0', 'Tuesday': '7.0-19.0', ...
...
141132 {Monday': '5:30-15.0', 'Tuesday': '5:30-15.0'...
141133 {Monday': '7.0-17.0', 'Tuesday': '7.0-17.0', ...
141134 {Monday': '10.30-0.0', 'Tuesday': '10.30-0.0'...
141135 {Monday': '11.0-22.0', 'Tuesday': '11.0-22.0'...
141136 {Monday': '11.0-22.0', 'Tuesday': '11.0-22.0'...



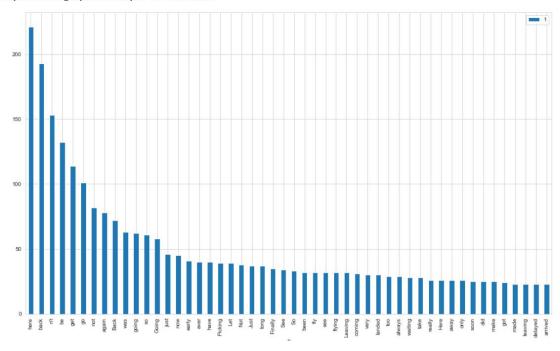




将列表中的元素按照頻率大小排序 x =pd.DataFrame(counter.most_common()) x

```
new=x.iloc[:50]
new.plot(0,kind='bar',figsize=(17,10))
```

(matplotlib.axes._subplots.AxesSubplot at 0x2c58e6d5d08>



Use IntensityAnalyzer & calculate the score

```
from n1tk.sentiment.vader import SentimentIntensityAnalyzer
```

sid=SentimentIntensityAnalyzer()

```
score=[]
for i in range(3679):
    sentence=r[i]
    ss=sid.polarity_scores(sentence)
    score.append(ss)
score=pd.DataFrame(score)
```

score

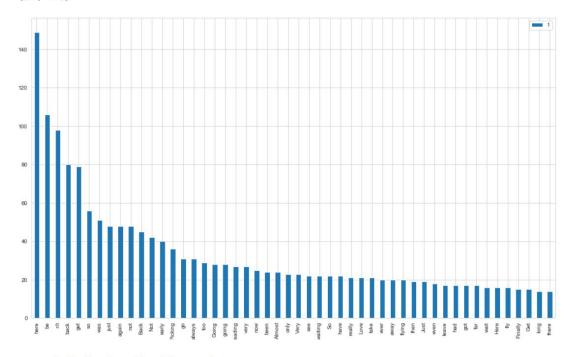
	neg	neu	pos	compound
0	0.000	1.000	0.000	0.0000
1	0.000	1.000	0.000	0.0000
2	0.000	0.882	0.118	0.3680
3	0.000	1.000	0.000	0.0000
4	0.000	0.590	0.410	0.7067
	100	100	100	1:22
3674	0.000	1.000	0.000	0.0000
3675	0.534	0.466	0.000	-0.7830
3676	0.000	0.626	0.374	0.5550
3677	0.000	1.000	0.000	0.0000

```
import numpy
from collections import Counter
def count_sentence_emotion(score):
list1 = score[ neg ].values.tolist()
list2=score[ neg ].values.tolist()
list3=score[ pos ].values.tolist()
list4=score[ pos ].values.tolist()
collection_words1 = Counter(list1)
most_counterNum1 = collection_words1.most_counterNum1)
collection_words2 = Counter(list2)
most_counterNum2 = collection_words2.most_common(1)
print(most_counterNum2)
collection_words3 = Counter(list2)
most_counterNum2 = collection_words3.most_common(1)
print(most_counterNum3)
collection_words3 = Counter(list3)
most_counterNum3 = collection_words3.most_common(1)
print(most_counterNum3)
collection_words4 = Counter(list4)
most_counterNum3 = Counter(list4)
most_counterNum3 = Counter(list4)
most_counterNum4)
count_sentence_emotion(election_words4.most_common(1)
print(most_counterNum4)
count_sentence_emotion(score)

[(0.0, 2006)]
```

[(0.0, 2906) [(1.0, 1811) [(0.0, 2293)

[(0.0, 2048)] [(1.0, 1200)] [(0.0, 1467)] [(0.0, 1204)]



use statistical method to analyse

```
: text_new=''
for i in range(20000): #殷前2000条进行分析
text_new=text_new+' '+text.iloc[i]
```

```
words=n1tk.word_tokenize(text_new)
 from nltk.stem import SnowballStemmer
 snowball_stemmer=SnowballStemmer("english")
 words_new=[]
 for i in words:
           words_new.append(snowball_stemmer.stem(i))
 from nltk.corpus import stopwords #差傳用词
 words_new=[word for word in words_new if word not in stopwords.words('english')]
                                                                                             #去掉一些不能提供情感分析的符号
 words_refresh=[]
 import re
 for list in words_new:
          string = re.sub("[\s+\.\!\/_$%^*(+\"\']+|[+---!, . ?:) `` ; ? ~ @#\Y.......&* () ]+", " ", list) if string!=' ':
                    words_refresh.append(string)
 import enchant
  def is_english_word(word):
    d_en = enchant.Dict("en_US")
            return d_en.check(word)
  words_fin=[]
                                                                                  #对拼写进行检查
  for i in words_refresh:
             if is_english_word(i) == True:
                      words_fin.append(i)
freq_dist = nltk.FreqDist(words_fin)
for k,v in freq_dist.items():
    print(str(k)+':'+str(v))
print(freq_dist.most_common(50)) #最常见的50个词
  from wordcloud import WordCloud
  width=1500
                      height=960,
                       margin=10
                       ).generate(str(words_fin))
 plt.imshow(wordcloud)
plt.axis("off")
plt.show()
  wordcloud.to_file('my_test2.png')
          eat ...drink!pizzal new! back check' come
                            rink pizza newy front od to the pizza newy food to the pizza newy food to the pizza new 
                                                                         place
 <wordcloud.wordcloud.WordCloud at 0x2821da5dcc8>
```

Use same method to analyze reviews

```
import pandas as pd
 ## Bulid up a chunsize to load the data
{\tt df=pd.\,read\_json('yelp\_academic\_dataset\_review.\,json', lines=True, chunksize=20000, nrows=20000)}
print(chunksize)
                                                                                                                           review id
                                                                                                                                                                                                                                                                                                                                                                                               business id \
                                                       xQY8N XvtGbearT5X4QrvQ OwiRMXRC0KvPrI1ciaXeFQ -MhfebM0QIsKt87iDN-FNw

        UmfNZ8PyXTTY2Gewssf0YA
        nJJD_7ZXHq-FX8byPM0xHQ
        brUsStG39Dfr-Qlfn5rna

        L62ZaY10gpr2DK_90y1Mv
        V34qe_jxNsCocgD8C0HXv-Q
        bt128ZHwzERHafhTraQvNQ

        6faWDKywebjoTki zeMce8A
        UgW8bLEOQMJDCkQ1Ax5Mg
        154cv902ykd8vj1TR0N3-A

              | 19995 | uFBmIMdpnJC4Tq7n3gWOJs | vBOAoIDR7b6r3v3pAqMtLA | sAFFh7e7tY-W2gWWSbIAWQ | 19996 | G8xO5iym3dxYAqtANdWTJw | Hprxt1x2u8SbKJPYCrGsbQ | SFcmwdfsyjsgmu-b9L91Ww | 19997 | OsmdUzukFBleozjpHuIXcw | HvQeFvu7lzDrKUZV17rZzkg | 1jfwt_DkxQgTZzlcwhl4dg | 19998 | 8jgmirikW7--w9fd4dyis | TBiAkd8UB1VytvOpBOAtAQ | JmWNVYqkXPFXbCxHsiuQ | 19999 | CTzPiVuhrgsFPHKZXk_CFQ | XuXdIy1Z6OrldbeBQ4HpWQ | uoZwJJap1LsktVrRvGJiQw | UoZwJJ
                                                     stars useful funny cool \
              0
              4
                                                                                                                     0
                                                                                                                                                            0
                                                                                                                                                                                               0
                                                                         4
              19995
19996
                19997
                19998
```

Data Exploration && Data Cleaning

```
data=chunksize
  data.duplicated()
               False
               False
              False
             False
  19995
  19996
19997
              False
False
 19998 False
19999 False
Length: 20000, dtype: bool
  data.isnul1().sum()
  review_id
  user_id
business_id
                       0
  stars
useful
                        0
  funny
   cool
  text
  date
 dtype: int64
 data['stars'].value_counts()
data['useful'].value_counts()
data['funny'].value_counts()
data['cool'].value_counts()
 #通过分析后发现stars数据最为正常,决定首先分析stars评论星级为5的文本
stars=data[data['stars'].isin([5])]
 text=stars.text
   import nltk
   from nltk.tokenize import sent_tokenize, word_tokenize
   word=
for i in range(2000):
    word=word+' ' +text.iloc[i]
   word=n1tk.word_tokenize(word)
   from nltk.corpus import stopwords
   word=[word for word in word if word not in stopwords.words('english')]
   from nltk.corpus import stopwords
from nltk import word_tokenize,pos_tag
from nltk.stem import WordNetLemmatizer
  #对文本进行清洗
  import re
from nltk.corpus import stopwords
from nltk import word_tokenize, pos_tag
from nltk.stem import WordNetLemmatizer
  def tokenize(sentence):
      ,,, 去除多余空白、分词、词性标注
      sentence = re.sub(r'\s+', '', sentence)
token_words = word_tokenize(sentence)
token_words = pos_tag(token_words)
return token_words
  wordnet_lematizer = WordNetLemmatizer()
 sr = stopwords.words('english')
def delete_stopwords(token_words):
      ,,, 去停用词
       cleaned_words = [word for word in token_words if word not in sr]
return cleaned_words
 def is_number(s):
       ,,, 判断字符串是否为数字
      try:
    float(s)
    return True
except ValueError:
    pass
```

```
import unicodedata
          unicodedata.numeric(s)
         return True
     except (TypeError, ValueError):
         pass
    return False
 \text{characters} = [\ '\ ',\ 'i\ ',\ '\text{the'}\ ',\ '\text{we'}\ ,\ ',\ '\ ',\ '\text{DBSCAN'},\ ':\ ',\ '?',\ '(',\ ')',\ '[',\ ']',\ '\&',\ '!',\ '*',\ '\&',\ ''\#',\ '\&',\ '-',\ '...',\ '^{'},\ '(',\ ')']   \text{def} \ \text{delete\_characters}(\text{token\_words}): 
    ,,, 去除特殊字符、数字
     words_list = [word for word in token_words if word not in characters and not is_number(word)]
     return words_list
def to_lower(token_words):
    ,,, 统一为小写
    words_lists = [x.lower() for x in token_words]
return words_lists
def pre_process(text):
    ,,, 文本预处理
    token_words = tokenize(text)
    token_words = stem(token_words)
token_words = delete_stopwords(token_words)
    token_words = delete_characters(token_words)
token_words = to_lower(token_words)
 return token_words
```

word = delete_stopwords(word) word = delete_characters(word) word = to_lower(word)

Analyze

tops[7][0]=time

tops[7][1]=0.00 tops[8][0]=best

[1]=0.006699532296801921

tops[8][1]=0.005642316222521 tops[9][0]=back tops[9][1]=0.005309063329541145

```
from nltk.probability import FreqDist fdist = FreqDist(word) tops=fdist.most_common(50) print(tops)

[('great', 1057), ("'s", 929), ('place', 906), ("n't", 860), ('food', 848), ('good', 735), ('service', 656), ('time', 593), ('best', 491), ('back', 462), ('one', 430), ('get', 429), ('love', 423), ('like', 420), ('amazing', 409), ('really', 407), ('go', 392), ('would', 382), ('ve', 378), ('friendly', 375), ('also', 373), ('always', 355), ('nice', 342), ('delicious', 338), ('staff', 326), ('definitely', 317), ('well', 310), ('recommend', 290), ('us', 289), ('got', 285), ('made', 273), ('restaurant', 271), ('try', 266), ('awesome', 242), ("m', 232), ('even', 232), ('even', 232), ('new', 229), ('could', 229), ('everything', 224), ('much', 223), ('come', 222), ('everything', 224), ('much', 223), ('come', 221), ('went', 212), ('fresh', 209), ('little', 204)]
```

```
#计算高频词出现的频率
def ret (x):
return x
for a in map(ret, tops):
     j=0
for x in a:
           ... a.
print("tops[%d][%d]="%(k, j), end=""),
if j!=0:
                 print(int(x)/87021," ")
           print(x, " ")
j=j+1
     print
     k+=1
 tops[0][0]=great
tops[0][1]=0.012146493375162317
tops[1][0]='s
 tops[1][0]= s
tops[1][1]=0.010675584054423645
tops[2][0]=place
tops[2][1]=0.010411280035853414
 tops[3][0]=n't
tops[3][1]=0.009882671998712955
 tops[4][0]=food
tops[4][1]=0.009744774249893704
 tops[5][0]=good
tops[5][1]=0.008446237115179094
 tops[6][0]=service
tops[6][1]=0.007538410268785696
```

Use Citysearch Corpus to begin the Supervised Learning

```
: try_text=pd.read_table('test.txt') #导入Citysearch Corpus语科库进行监督学习 ## the corpus Citysearch Corpus is a restearant's comment set from Citysearch New-York Net

    Always a fun place ... the food is deeelish!
     1 The staff is n't the friendliest or most comp...
 2 Great for groups , great for a date , great f...
 3 Another great place to take out-of-towners ! 4 :)
  3323 Was there Friday night .
  3324 Best Pastrami I ever had and great portion wi...
  3325 And I 've been to many NYC delis .
  3326 My wife had the fried shrimp which are huge a...
 3327 Price no more than a Jersey deli but way bett...
 3328 rows × 1 columns
 word='
 for i in range (3327):
 word=word+try_text.iloc[i].values
 str = ','.join(str(i) for i in word)
 word=nltk.word_tokenize(str)
 tops_word=[]
 def ret (x):
return x
k=0
 for a in map(ret, tops):
     j=0
for x in a:
      tops_word.append(x)
j=j+1
     print
 words=[] #为stars文本中的50个高频词汇
 while i<100:
    words.append(tops_word[i])
i=i+2;
  tops_fre=[] #频率
  while i<100:
      tops_fre.append(tops_word[i]/87021)
i=i+2;
 len (word)
54528
 fre= nltk. FreqDist(word)
 fre
FreqDist(('.': 2993, 'the': 2259, ',': 1943, 'and': 1629, 'a': 1151, 'to': 995, 'I': 959, 'is': 923, 'was': 731, 'of': 695, ...})
Calculate IDF
#if#IDF
import math
IDF=[]
for i in range(SO):
IDF.append(math.log(1/fre.freq(words[i])))
```

```
#计算文本的feature= IDF*Tops_fre 最后降序排列
     feature=[]
     for i in range (50):
     feature.append(IDF[i]*tops_fre[i]*100)
  #单词与频率组成字典
     nvs = zip(words, feature)
FinDict = dict( (name, value) for name, value in nvs)
      #将字典根据value进行排序
     FinDict=sorted(FinDict.items(), key=lambda item:item[1], reverse=True)
    FinDict
('great', 6.647469254792439),
('great', 5.495329311239328),
('place', 5.355997878710176),
("n't', 5.2098284391548875),
('good', 4.710063622025854),
('good', 4.477069494445418),
('service', 4.250616646485828),
('time', 4.248924727496157),
('best', 3.462003916284749),
('love', 3.4404584346250386),
('amazing', 3.392278156657481),
('back', 3.324577245152318),
('really', 3.08812262075256),
('get', 3.008412262075256),
('gre', 2.9878564926922766),
('friendly', 2.9355522149683204)
(get , 3.003412202012200),

('one', 2.9878564926922766),

('friendly', 2.9355522149683204),

('also', 2.9271000080712226),

('like', 2.839185274271657),

("ve", 2.6726748950405437),

('would', 2.6577444872788174),

('definitely', 2.6576227440752116),

('always', 2.656562509863567),

('go', 2.615677189973681),

('nice', 2.600148789156507),

('delicious', 2.5440056363668493),

('awesome', 2.5347443193166543),

('staff', 2.4785044013597117),

('well', 2.337856262722727),

('got', 2.2778862333323615),

('made', 2.256536961619974),

('recommend', 2.2434859156813545),

('us', 2.1505885504405833),

('first', 2.130732166717449).
 Fin_word=[] #根据walue排序后的字典的特征单词的list def ret (x):
 return x
    for a in map(ret, FinDict):
                       j=0
                          for x in a:
                                           Fin_word.append(x)
                       print
  Fin_words=[]
    while i<100:
                       Fin_words.append(Fin_word[i])
     del str
        #将feature word绘制成词云
  #形feature word是刺版语云
from wordcloud import WordCloud
import matplotlib.pyplot as plt
wordcloud = WordCloud(
background_color="white",
width=1500,
height=960,
  neignt=900,
margin=10 ).generate(str(Fin_words))
plt.imshow(wordcloud)
plt.axis("off")
plt.show()
     wordcloud.to_file('Fin_words.png')
                              friendly messone amazing really vectors also vectors of the control of the contro
```

best case staff ress good like would us