

Advanced programming

Wanwen liu Sun He

First we upload our datasets

```
import pandas as pd  
  
cars = pd.read_csv('auto-mpg.data',  
delim_whitespace=True,header=None)  
  
cars.describe()  
  
modelyear = cars.iloc[:,6].tolist()  
  
mpg = cars.iloc[:,0].tolist()
```

Second we modify our data for preparing next steps. We built a new dataset of modelyear and mpg.

```
mymodelyear=list(set(modelyear))  
  
cars1=cars.iloc[:,np.r_[0,6]]  
  
cars1.columns=["mpg","modelyear"]  
  
cars1
```

then we get a new dataset with modelyear and mpg

	mpg	modelyear
0	18.0	70
1	15.0	70
2	18.0	70
3	16.0	70
4	17.0	70
5	15.0	70
6	14.0	70
7	14.0	70
8	14.0	70
9	15.0	70
10	15.0	70
11	14.0	70
12	15.0	70
13	14.0	70
14	24.0	70
15	22.0	70
16	18.0	70
17	21.0	70

380	36.0	82
381	36.0	82
382	34.0	82
383	38.0	82
384	32.0	82
385	38.0	82
386	25.0	82
387	38.0	82
388	26.0	82
389	22.0	82
390	32.0	82
391	36.0	82
392	27.0	82
393	27.0	82
394	44.0	82
395	32.0	82
396	28.0	82
397	31.0	82

398 rows × 2 columns

We build a dictionary of the dataset 'cars1', and use "get_dummies" to classify the data.

```
dict=cars1.set_index('modelyear').T.to_dict('list')
```

dict

```
cars2=pd.get_dummies(cars1,prefix=['myear'],columns=['modelyear'])
```

cars2

	mpg	myear_70	myear_71	myear_72	myear_73	myear_74	myear_75	myear_76	myear_77	myear_78	myear_79	myear_80	myear_81	myear_82
0	18.0	1	0	0	0	0	0	0	0	0	0	0	0	0
1	15.0	1	0	0	0	0	0	0	0	0	0	0	0	0
2	18.0	1	0	0	0	0	0	0	0	0	0	0	0	0
3	16.0	1	0	0	0	0	0	0	0	0	0	0	0	0
4	17.0	1	0	0	0	0	0	0	0	0	0	0	0	0
5	15.0	1	0	0	0	0	0	0	0	0	0	0	0	0
6	14.0	1	0	0	0	0	0	0	0	0	0	0	0	0
7	14.0	1	0	0	0	0	0	0	0	0	0	0	0	0
8	14.0	1	0	0	0	0	0	0	0	0	0	0	0	0
9	15.0	1	0	0	0	0	0	0	0	0	0	0	0	0
10	15.0	1	0	0	0	0	0	0	0	0	0	0	0	0
11	14.0	1	0	0	0	0	0	0	0	0	0	0	0	0
12	15.0	1	0	0	0	0	0	0	0	0	0	0	0	0
13	14.0	1	0	0	0	0	0	0	0	0	0	0	0	0
14	24.0	1	0	0	0	0	0	0	0	0	0	0	0	0
15	22.0	1	0	0	0	0	0	0	0	0	0	0	0	0
381	36.0	0	0	0	0	0	0	0	0	0	0	0	0	1
382	34.0	0	0	0	0	0	0	0	0	0	0	0	0	1
383	38.0	0	0	0	0	0	0	0	0	0	0	0	0	1
384	32.0	0	0	0	0	0	0	0	0	0	0	0	0	1
385	38.0	0	0	0	0	0	0	0	0	0	0	0	0	1
386	25.0	0	0	0	0	0	0	0	0	0	0	0	0	1
387	38.0	0	0	0	0	0	0	0	0	0	0	0	0	1
388	26.0	0	0	0	0	0	0	0	0	0	0	0	0	1
389	22.0	0	0	0	0	0	0	0	0	0	0	0	0	1
390	32.0	0	0	0	0	0	0	0	0	0	0	0	0	1
391	36.0	0	0	0	0	0	0	0	0	0	0	0	0	1
392	27.0	0	0	0	0	0	0	0	0	0	0	0	0	1
393	27.0	0	0	0	0	0	0	0	0	0	0	0	0	1
394	44.0	0	0	0	0	0	0	0	0	0	0	0	0	1
395	32.0	0	0	0	0	0	0	0	0	0	0	0	0	1
396	28.0	0	0	0	0	0	0	0	0	0	0	0	0	1
397	31.0	0	0	0	0	0	0	0	0	0	0	0	0	1

398 rows x 14 columns

Building an index

```
from sklearn.preprocessing import OneHotEncoder
```

```
ohc=OneHotEncoder()
```

```
ohe=ohc.fit_transform(cars1.modelyear.values.reshape(-1,1)).toarray()
```

```
a=cars2.columns
```

```
from sklearn.preprocessing import OneHotEncoder
ohc=OneHotEncoder()
ohe=ohc.fit_transform(cars1.modelyear.values.reshape(-1,1)).toarray()
a=cars2.columns
```

/srv/conda/lib/python3.6/site-packages/sklearn/preprocessing/_encoders.py:371: FutureWarning: The handling of integer data will change in version 0.22. Currently, the categories are determined based on the range [0, max(values)], while in the future they will be determined based on the unique values.

If you want the future behaviour and silence this warning, you can specify "categories='auto'".

In case you used a LabelEncoder before this OneHotEncoder to convert the categories to integers, then you can now use the OneHotEncoder directly.

warnings.warn(msg, FutureWarning)

a

```
Index(['mpg', 'myear_70', 'myear_71', 'myear_72', 'myear_73', 'myear_74',
      'myear_75', 'myear_76', 'myear_77', 'myear_78', 'myear_79', 'myear_80',
      'myear_81', 'myear_82'],
      dtype='object')
```



Calculating the sum of the numbers and the whole number of data ,modify them

```
dfOneHot=pd.DataFrame(ohe,columns=a[1:])
```

```
dfOneHot
```

```
dfh1=pd.concat([cars1,dfOneHot],axis=1)
```

```
dfh1
```

```
colnum=dfh1.apply(lambda x: x.sum())
```

```
colnum=dfh1.apply(lambda x: x.sum())
```

colnum

```
mpg          9358.8
modelyear    30252.0
myear_70     29.0
myear_71     28.0
myear_72     28.0
myear_73     40.0
myear_74     27.0
myear_75     30.0
myear_76     34.0
myear_77     28.0
myear_78     36.0
myear_79     29.0
myear_80     29.0
myear_81     29.0
myear_82     31.0
dtype: float64
```

Using a while to multiply every column in the dataset with mpg. So that ,we can replace number of “year” to the mpg of them.

 $i=2$

```
while i<15:
```

```
dfh1.iloc[:,i]=dfh1.iloc[:,i]*mpg
```

$$i=i+1;$$

dfh1

[illegible]

34.0	82	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	34.0
38.0	82	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	38.0
32.0	82	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	32.0
38.0	82	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	38.0
25.0	82	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	25.0
38.0	82	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	38.0
26.0	82	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	26.0
22.0	82	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	22.0
32.0	82	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	32.0
36.0	82	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	36.0
27.0	82	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	27.0
27.0	82	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	27.0
44.0	82	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	44.0
32.0	82	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	32.0
28.0	82	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	28.0
31.0	82	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	31.0

ows × 15 columns

Calculate the sum of each column and average

```
colsum=dfh1.apply(lambda x: x.sum())
```

```
average=colsum/colnum
```

```
average=average[2:]
```

Using a for loop to replace every column to mpg.

```
for i in mymodelyear:
```

```
dict[i]=average[i-70]
```

```
dict
```

```
cars1['y']=cars1.modelyear.map(dict)
```

```
cars1
```

	mpg	modelyear	y				
0	18.0	70	17.689655	380	36.0	82	31.709677
1	15.0	70	17.689655	381	36.0	82	31.709677
2	18.0	70	17.689655	382	34.0	82	31.709677
3	16.0	70	17.689655	383	38.0	82	31.709677
4	17.0	70	17.689655	384	32.0	82	31.709677
5	15.0	70	17.689655	385	38.0	82	31.709677
6	14.0	70	17.689655	386	25.0	82	31.709677
7	14.0	70	17.689655	387	38.0	82	31.709677
8	14.0	70	17.689655	388	26.0	82	31.709677
9	15.0	70	17.689655	389	22.0	82	31.709677
10	15.0	70	17.689655	390	32.0	82	31.709677
11	14.0	70	17.689655	391	36.0	82	31.709677
12	15.0	70	17.689655	392	27.0	82	31.709677
13	14.0	70	17.689655	393	27.0	82	31.709677
14	24.0	70	17.689655	394	44.0	82	31.709677
15	22.0	70	17.689655	395	32.0	82	31.709677
16	18.0	70	17.689655	396	28.0	82	31.709677
17	21.0	70	17.689655	397	31.0	82	31.709677
18	27.0	70	17.689655	398 rows x 3 columns			