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PC 66

Week 9 demo

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

fruit_datafruit = pd.Series(['apple', 'orange', 'apple', 'apple']*2)
fruit_datafruit.value_counts()

apple      6
orange     2
Name: count, dtype: int64

fruit_datafruit.unique()

array(['apple', 'orange'], dtype=object)

prod_ser = pd.Series(["a", "b", "c", "a", "b", "c", "a"])
prod_cat= prod_ser.astype("category")
prod_cat

0      a
1      b
2      c
3      a
4      b
5      c
6      a
dtype: category
Categories (3, object): ['a', 'b', 'c']

prod_typ = prod_cat.array
prod_typ.categories
prod_typ.codes

array([0, 1, 2, 0, 1, 2, 0], dtype=int8)

data = pd.Series(["low", "medium", "high", "low", "high"])
Income_cat = pd.Categorical(data)
print(Income_cat)
```

```

['low', 'medium', 'high', 'low', 'high']
Categories (3, object): ['high', 'low', 'medium']

data_cat = pd.Categorical(data, categories =
["low", "medium", "high"], ordered = True)
print(data_cat)

['low', 'medium', 'high', 'low', 'high']
Categories (3, object): ['low' < 'medium' < 'high']

num_series = pd.Series([10, 25, 50, 75, 45, 105, 13, 45])
bins = [0, 30, 60, 100]
labels = ["low", "medium", "high"]
num_cat = pd.cut(num_series, bins = bins, labels = labels)
print("Count in each bin")
num_cat.value_counts()

```

Count in each bin

```

low      3
medium   3
high     1
Name: count, dtype: int64

```

```

data = pd.read_csv("mtcars.csv", index_col=0)
mpg = data["mpg"]
bins = [0, 18, 25, 40]
labels = ["low", "medium", "high"]
mpg_cat = pd.cut(mpg, bins = bins, labels = labels)
data['mpgCat'] = mpg_cat
data

```

gear \ model	mpg	cyl	disp	hp	drat	wt	qsec	vs	am
Mazda RX4	21.0	6	160.0	110	3.90	2.620	16.46	0	1
Mazda RX4 Wag	21.0	6	160.0	110	3.90	2.875	17.02	0	1
Datsun 710	22.8	4	108.0	93	3.85	2.320	18.61	1	1
Hornet 4 Drive	21.4	6	258.0	110	3.08	3.215	19.44	1	0
Hornet Sportabout	18.7	8	360.0	175	3.15	3.440	17.02	0	0
Valiant	18.1	6	225.0	105	2.76	3.460	20.22	1	0
Duster 360	14.3	8	360.0	245	3.21	3.570	15.84	0	0
Merc 240D	24.4	4	146.7	62	3.69	3.190	20.00	1	0

4
Merc 230
22.8
4
140.8
95
3.92
3.150
22.90
1
0
4
Merc 280
19.2
6
167.6
123
3.92
3.440
18.30
1
0
4
Merc 280C
17.8
6
167.6
123
3.92
3.440
18.90
1
0
4
Merc 450SE
16.4
8
275.8
180
3.07
4.070
17.40
0
0
3
Merc 450SL
17.3
8
275.8
180
3.07
3.730
17.60
0
0
3
Merc 450SLC
15.2
8
275.8
180
3.07
3.780
18.00
0
0
3
Cadillac Fleetwood
10.4
8
472.0
205
2.93
5.250
17.98
0
0
3
Lincoln Continental
10.4
8
460.0
215
3.00
5.424
17.82
0
0
3
Chrysler Imperial
14.7
8
440.0
230
3.23
5.345
17.42
0
0
3
Fiat 128
32.4
4
78.7
66
4.08
2.200
19.47
1
1
4
Honda Civic
30.4
4
75.7
52
4.93
1.615
18.52
1
1
4
Toyota Corolla
33.9
4
71.1
65
4.22
1.835
19.90
1
1
4
Toyota Corona
21.5
4
120.1
97
3.70
2.465
20.01
1
0
3
Dodge Challenger
15.5
8
318.0
150
2.76
3.520
16.87
0
0
3
AMC Javelin
15.2
8
304.0
150
3.15
3.435
17.30
0
0
3
Camaro Z28
13.3
8
350.0
245
3.73
3.840
15.41
0
0
3
Pontiac Firebird
19.2
8
400.0
175
3.08
3.845
17.05
0
0
3
Fiat X1-9
27.3
4
79.0
66
4.08
1.935
18.90
1
1
4
Porsche 914-2
26.0
4
120.3
91
4.43
2.140
16.70
0
1
5
Lotus Europa
30.4
4
95.1
113
3.77
1.513
16.90
1
1
5
Ford Pantera L
15.8
8
351.0
264
4.22
3.170
14.50
0
1
5
Ferrari Dino
19.7
6
145.0
175
3.62
2.770
15.50
0
1
5
Maserati Bora
15.0
8
301.0
335
3.54
3.570
14.60
0
1
5
Volvo 142E
21.4
4
121.0
109
4.11
2.780
18.60
1
1
4

	carb	mpgCat
model		
Mazda RX4	4	medium
Mazda RX4 Wag	4	medium
Datsun 710	1	medium
Hornet 4 Drive	1	medium
Hornet Sportabout	2	medium
Valiant	1	medium
Duster 360	4	low
Merc 240D	2	medium
Merc 230	2	medium
Merc 280	4	medium
Merc 280C	4	low
Merc 450SE	3	low
Merc 450SL	3	low
Merc 450SLC	3	low
Cadillac Fleetwood	4	low
Lincoln Continental	4	low
Chrysler Imperial	4	low
Fiat 128	1	high
Honda Civic	2	high
Toyota Corolla	1	high
Toyota Corona	1	medium
Dodge Challenger	2	low
AMC Javelin	2	low
Camaro Z28	4	low
Pontiac Firebird	2	medium
Fiat X1-9	1	high
Porsche 914-2	2	high
Lotus Europa	2	high
Ford Pantera L	4	low
Ferrari Dino	6	medium
Maserati Bora	8	low
Volvo 142E	2	medium

```
tab3 = data.groupby('mpgCat')['wt'].mean()
print("Table displaying categories of MPG and their average weight")
print(tab3)
```

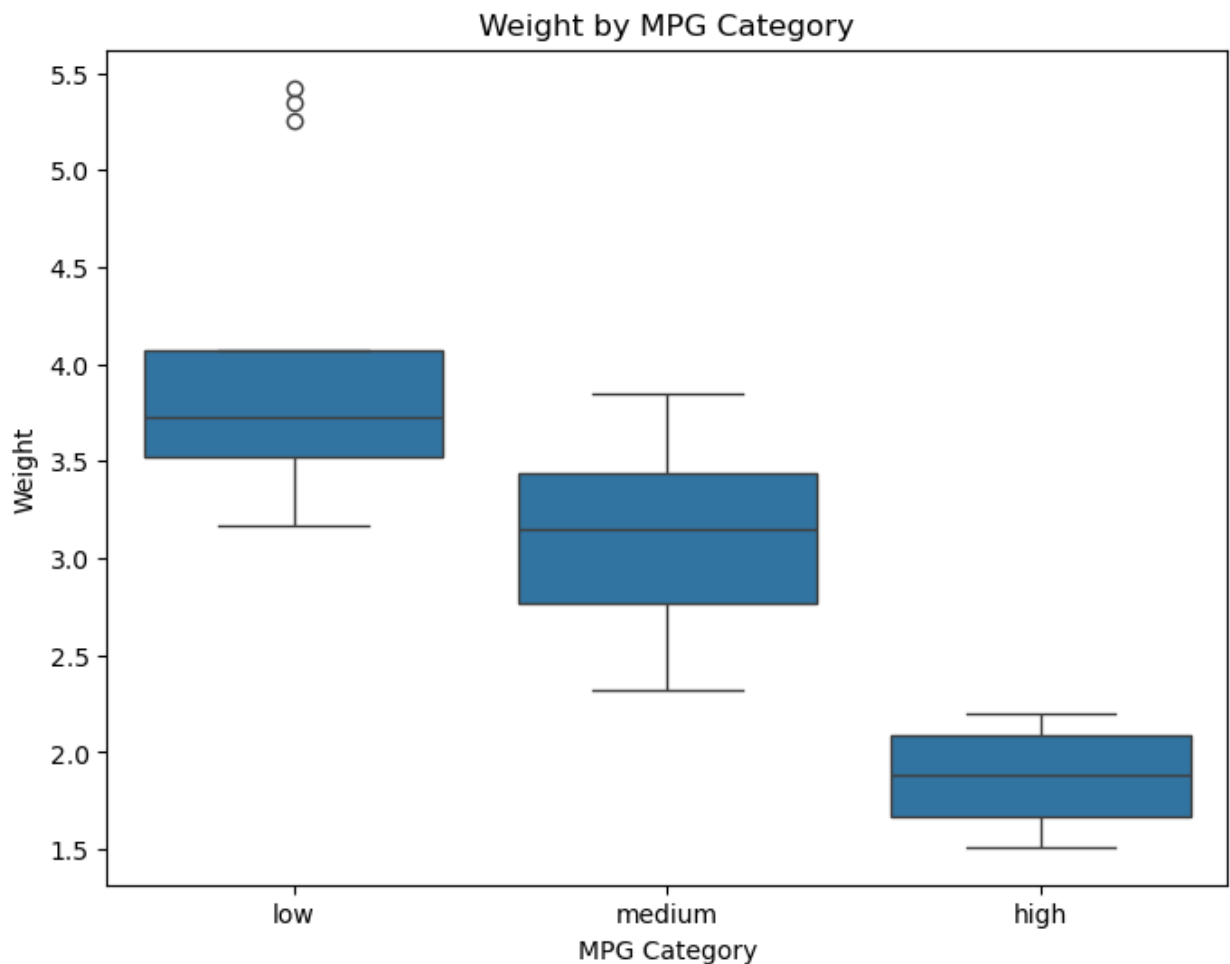
Table displaying categories of MPG and their average weight

```
mpgCat
low      4.011077
medium   3.043846
high     1.873000
Name: wt, dtype: float64
```

/tmp/ipykernel_110065/1719140303.py:1: FutureWarning: The default of observed=False is deprecated and will be changed to True in a future version of pandas. Pass observed=False to retain current behavior or

```
observed=True to adopt the future default and silence this warning.  
tab3 = data.groupby('mpgCat')['wt'].mean()
```

```
df1 = data[['mpgCat', 'cyl', 'wt', 'mpg']]  
plt.figure(figsize=(8, 6))  
sns.boxplot(x='mpgCat', y='wt', data=df1)  
plt.title('Weight by MPG Category')  
plt.xlabel('MPG Category')  
plt.ylabel('Weight')  
plt.show()
```



```
str1= "  Data visualization is very interesting. I love Python  
Programming.  "  
print('String length',len(str1))  
str1=str1.strip()  
print('String 1:',str1)  
str2=str1.split('.')  
print('String 2:',str2)
```

```
str3=str1.lower()  
print('String 3:', str3)
```

String length 71

String 1: Data visualization is very interesting. I love Python Programming.

String 2: ['Data visualization is very interesting', ' I love Python Programming', '']

String 3: data visualization is very interesting. i love python programming.

```
"python" in str3  
print(str1.index('Python'))  
str2=str1.replace('interesting','amazing')  
print(str2)  
str2.count('.')
```

47

Data visualization is very amazing. I love Python Programming.

2