

```
pip install seaborn
```

Requirement already satisfied: seaborn in c:\users\shagu\anaconda3\lib\site-packages (0.12.2)

Requirement already satisfied: numpy!=1.24.0,>=1.17 in c:\users\shagu\anaconda3\lib\site-packages (from seaborn) (1.24.3)

Requirement already satisfied: pandas>=0.25 in c:\users\shagu\anaconda3\lib\site-packages (from seaborn) (1.5.3)

Requirement already satisfied: matplotlib!=3.6.1,>=3.1 in c:\users\shagu\anaconda3\lib\site-packages (from seaborn) (3.7.1)

Requirement already satisfied: contourpy>=1.0.1 in c:\users\shagu\anaconda3\lib\site-packages (from matplotlib!=3.6.1,>=3.1->seaborn) (1.0.5)

Requirement already satisfied: cycler>=0.10 in c:\users\shagu\anaconda3\lib\site-packages (from matplotlib!=3.6.1,>=3.1->seaborn) (0.11.0)

Requirement already satisfied: fonttools>=4.22.0 in c:\users\shagu\anaconda3\lib\site-packages (from matplotlib!=3.6.1,>=3.1->seaborn) (4.25.0)

Requirement already satisfied: kiwisolver>=1.0.1 in c:\users\shagu\anaconda3\lib\site-packages (from matplotlib!=3.6.1,>=3.1->seaborn) (1.4.4)

Requirement already satisfied: packaging>=20.0 in c:\users\shagu\anaconda3\lib\site-packages (from matplotlib!=3.6.1,>=3.1->seaborn) (23.0)

Requirement already satisfied: pillow>=6.2.0 in c:\users\shagu\anaconda3\lib\site-packages (from matplotlib!=3.6.1,>=3.1->seaborn) (9.4.0)

Requirement already satisfied: pyparsing>=2.3.1 in c:\users\shagu\anaconda3\lib\site-packages (from matplotlib!=3.6.1,>=3.1->seaborn) (3.0.9)

Requirement already satisfied: python-dateutil>=2.7 in c:\users\shagu\anaconda3\lib\site-packages (from matplotlib!=3.6.1,>=3.1->seaborn) (2.8.2)

Requirement already satisfied: pytz>=2020.1 in c:\users\shagu\anaconda3\lib\site-packages (from pandas>=0.25->seaborn) (2022.7)

Requirement already satisfied: six>=1.5 in c:\users\shagu\anaconda3\lib\site-packages (from python-dateutil>=2.7->matplotlib!=3.6.1,>=3.1->seaborn) (1.16.0)

Note: you may need to restart the kernel to use updated packages.

```
import seaborn as sns
```

```
import pandas as pd
```

```
import matplotlib.pyplot as plt
```

```
df=pd.read_csv(r"C:\Users\shagu\Downloads\mtcars (1).csv")  
df
```

	model	mpg	cyl	disp	hp	drat	wt	qsec	vs
0	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
3	Hornet Sportabout	18.7	8	360.0	175	3.15	3.440	17.02	0
4	Valiant	18.1	6	225.0	105	2.76	3.460	20.22	1
5	Duster 360	14.3	8	360.0	245	3.21	3.570	15.84	0
6	Merc 240D	24.4	4	146.7	62	3.69	3.190	20.00	1
7	Merc 230	22.8	4	140.8	95	3.92	3.150	22.90	1
8	Merc 280	19.2	6	167.6	123	3.92	3.440	18.30	1
9	Merc 280C	17.8	6	167.6	123	3.92	3.440	18.90	1
10	Merc 450SE	16.4	8	275.8	180	3.07	4.070	17.40	0
11	Merc 450SL	17.3	8	275.8	180	3.07	3.730	17.60	0
12	Merc 450SLC	15.2	8	275.8	180	3.07	3.780	18.00	0
13	Cadillac Fleetwood	10.4	8	472.0	205	2.93	5.250	17.98	0
14	Lincoln Continental	10.4	8	460.0	215	3.00	5.424	17.82	0
15	Chrysler Imperial	14.7	8	440.0	230	3.23	5.345	17.42	0
16	Fiat 128	32.4	4	78.7	66	4.08	2.200	19.47	1
17	Honda Civic	30.4	4	75.7	52	4.93	1.615	18.52	1
18	Toyota Corolla	33.9	4	71.1	65	4.22	1.835	19.90	1
19	Toyota Corona	21.5	4	120.1	97	3.70	2.465	20.01	1
20	Dodge Challenger	15.5	8	318.0	150	2.76	3.520	16.87	0
21	AMC Javelin	15.2	8	304.0	150	3.15	3.435	17.30	0
22	Camaro Z28	13.3	8	350.0	245	3.73	3.840	15.41	0

24	Pontiac Firebird	19.2	8	400.0	175	3.08	3.845	17.05	0
0									
25	Fiat X1-9	27.3	4	79.0	66	4.08	1.935	18.90	1
1									
26	Porsche 914-2	26.0	4	120.3	91	4.43	2.140	16.70	0
1									
27	Lotus Europa	30.4	4	95.1	113	3.77	1.513	16.90	1
1									
28	Ford Pantera L	15.8	8	351.0	264	4.22	3.170	14.50	0
1									
29	Ferrari Dino	19.7	6	145.0	175	3.62	2.770	15.50	0
1									
30	Maserati Bora	15.0	8	301.0	335	3.54	3.570	14.60	0
1									
31	Volvo 142E	21.4	4	121.0	109	4.11	2.780	18.60	1
1									
	gear	carb							
0	4	4							
1	4	4							
2	4	1							
3	3	1							
4	3	2							
5	3	1							
6	3	4							
7	4	2							
8	4	2							
9	4	4							
10	4	4							
11	3	3							
12	3	3							
13	3	3							
14	3	4							
15	3	4							
16	3	4							
17	4	1							
18	4	2							
19	4	1							
20	3	1							
21	3	2							
22	3	2							
23	3	4							
24	3	2							
25	4	1							
26	5	2							
27	5	2							
28	5	4							
29	5	6							

```
30      5      8
31      4      2
```

```
df.head()
```

		model	mpg	cyl	disp	hp	drat	wt	qsec	vs
am	gear \									
0		Mazda RX4	21.0	6	160.0	110	3.90	2.620	16.46	0
1	4									
1		Mazda RX4 Wag	21.0	6	160.0	110	3.90	2.875	17.02	0
1	4									
2		Datsun 710	22.8	4	108.0	93	3.85	2.320	18.61	1
1	4									
3		Hornet 4 Drive	21.4	6	258.0	110	3.08	3.215	19.44	1
0	3									
4		Hornet Sportabout	18.7	8	360.0	175	3.15	3.440	17.02	0
0	3									

	carb
0	4
1	4
2	1
3	1
4	2

```
df.tail()
```

		model	mpg	cyl	disp	hp	drat	wt	qsec	vs	am
gear \											
27		Lotus Europa	30.4	4	95.1	113	3.77	1.513	16.9	1	1
5											
28		Ford Pantera L	15.8	8	351.0	264	4.22	3.170	14.5	0	1
5											
29		Ferrari Dino	19.7	6	145.0	175	3.62	2.770	15.5	0	1
5											
30		Maserati Bora	15.0	8	301.0	335	3.54	3.570	14.6	0	1
5											
31		Volvo 142E	21.4	4	121.0	109	4.11	2.780	18.6	1	1
4											

	carb
27	2
28	4
29	6
30	8
31	2

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 32 entries, 0 to 31
```

```
Data columns (total 12 columns):
#   Column  Non-Null Count  Dtype
---  -
0   model    32 non-null      object
1   mpg       32 non-null      float64
2   cyl       32 non-null      int64
3   disp      32 non-null      float64
4   hp        32 non-null      int64
5   drat      32 non-null      float64
6   wt        32 non-null      float64
7   qsec      32 non-null      float64
8   vs        32 non-null      int64
9   am        32 non-null      int64
10  gear      32 non-null      int64
11  carb      32 non-null      int64
dtypes: float64(5), int64(6), object(1)
memory usage: 3.1+ KB
```

```
df.describe()
```

	mpg	cyl	disp	hp	drat
count	32.000000	32.000000	32.000000	32.000000	32.000000
mean	20.090625	6.187500	230.721875	146.687500	3.596563
std	6.026948	1.785922	123.938694	68.562868	0.534679
min	10.400000	4.000000	71.100000	52.000000	2.760000
25%	15.425000	4.000000	120.825000	96.500000	3.080000
50%	19.200000	6.000000	196.300000	123.000000	3.695000
75%	22.800000	8.000000	326.000000	180.000000	3.920000
max	33.900000	8.000000	472.000000	335.000000	4.930000

	qsec	vs	am	gear	carb
count	32.000000	32.000000	32.000000	32.000000	32.0000
mean	17.848750	0.437500	0.406250	3.687500	2.8125
std	1.786943	0.504016	0.498991	0.737804	1.6152
min	14.500000	0.000000	0.000000	3.000000	1.0000
25%	16.892500	0.000000	0.000000	3.000000	2.0000
50%	17.710000	0.000000	0.000000	4.000000	2.0000
75%	18.900000	1.000000	1.000000	4.000000	4.0000
max	22.900000	1.000000	1.000000	5.000000	8.0000

```
type(df)
```

```
pandas.core.frame.DataFrame
```

```
df.shape
```

```
(32, 12)
```

```
df.isnull().sum()
```

```
model      0
```

```
mpg        0
```

```
cyl        0
```

```
disp       0
```

```
hp         0
```

```
drat       0
```

```
wt         0
```

```
qsec       0
```

```
vs         0
```

```
am         0
```

```
gear       0
```

```
carb       0
```

```
dtype: int64
```

```
df.loc[:, "model"].head(10)
```

```
0      Mazda RX4
```

```
1    Mazda RX4 Wag
```

```
2    Datsun 710
```

```
3    Hornet 4 Drive
```

```
4  Hornet Sportabout
```

```
5      Valiant
```

```
6    Duster 360
```

```
7    Merc 240D
```

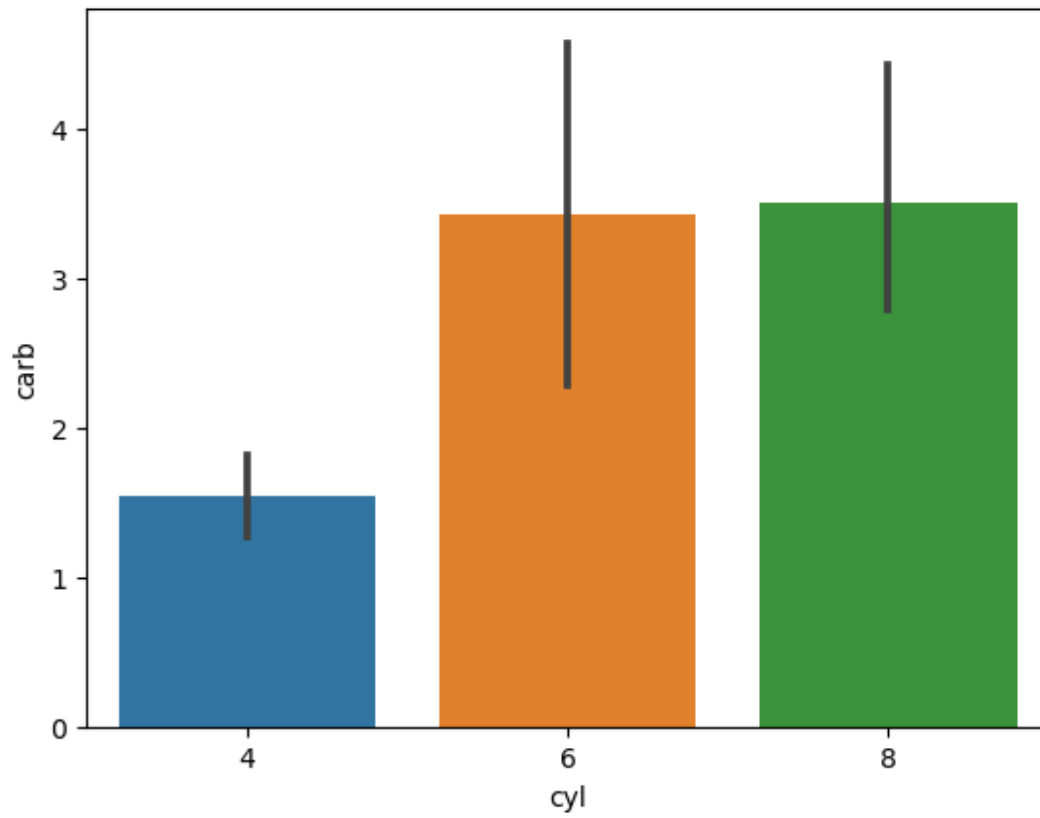
```
8      Merc 230
```

```
9      Merc 280
```

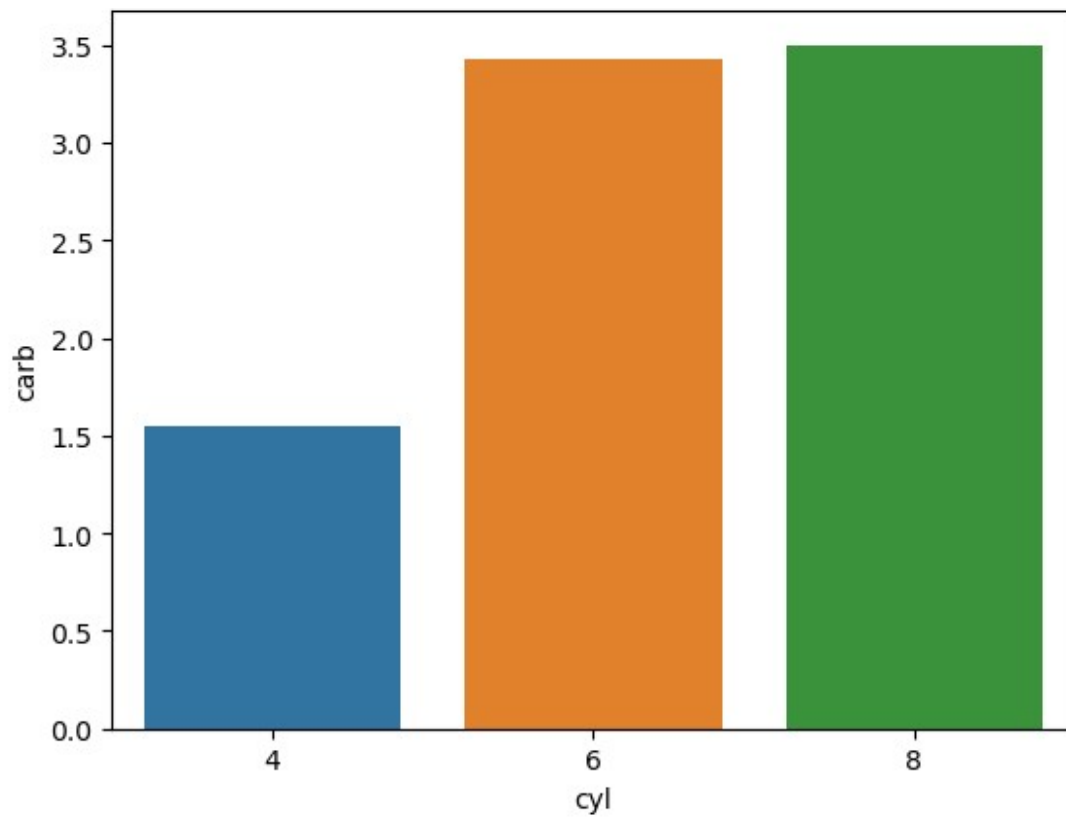
```
Name: model, dtype: object
```

```
sns.barplot(data = df, x='cyl', y='carb')
```

```
<Axes: xlabel='cyl', ylabel='carb'>
```

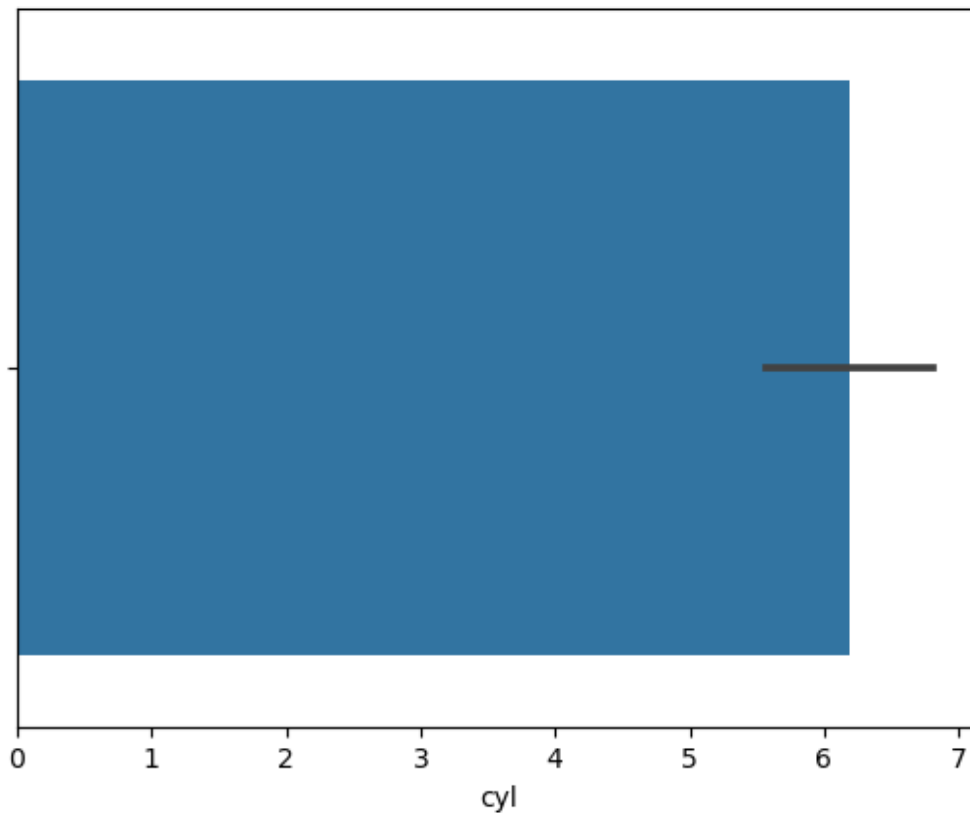


```
sns.barplot(data = df,x='cyl',y='carb', errorbar=None)  
<Axes: xlabel='cyl', ylabel='carb'>
```



```
sns.barplot(data = df,x='cyl')  
<Axes: xlabel='cyl'>
```





```
import pandas as pd
df = pd.read_csv(r"C:\Users\shagu\Downloads\supermarket sales Sheet1
(1).csv ")
df
```

	Invoice ID	Branch	City	Customer type	Gender	\
0	750-67-8428	A	Yangon	Member	Female	
1	226-31-3081	C	Naypyitaw	Normal	Female	
2	631-41-3108	A	Yangon	Normal	Male	
3	123-19-1176	A	Yangon	Member	Male	
4	373-73-7910	A	Yangon	Normal	Male	
...	...	...	...	...	...	
995	233-67-5758	C	Naypyitaw	Normal	Male	
996	303-96-2227	B	Mandalay	Normal	Female	
997	727-02-1313	A	Yangon	Member	Male	
998	347-56-2442	A	Yangon	Normal	Male	
999	849-09-3807	A	Yangon	Member	Female	

	Product line	Unit price	Quantity	Tax 5%	Total
0	Health and beauty	74.69	7	26.1415	548.9715
1	Electronic accessories	15.28	5	3.8200	80.2200
2	Home and lifestyle	46.33	7	16.2155	340.5255

3	Health and beauty	58.22	8	23.2880	489.0480
4	Sports and travel	86.31	7	30.2085	634.3785
..	...	...	...	...	...
995	Health and beauty	40.35	1	2.0175	42.3675
996	Home and lifestyle	97.38	10	48.6900	1022.4900
997	Food and beverages	31.84	1	1.5920	33.4320
998	Home and lifestyle	65.82	1	3.2910	69.1110
999	Fashion accessories	88.34	7	30.9190	649.2990

	Date	Time	Payment	cogs	gross margin percentage \
0	1/5/2019	13:08	Ewallet	522.83	4.761905
1	3/8/2019	10:29	Cash	76.40	4.761905
2	3/3/2019	13:23	Credit card	324.31	4.761905
3	1/27/2019	20:33	Ewallet	465.76	4.761905
4	2/8/2019	10:37	Ewallet	604.17	4.761905
..	...	...	...	...	...
995	1/29/2019	13:46	Ewallet	40.35	4.761905
996	3/2/2019	17:16	Ewallet	973.80	4.761905
997	2/9/2019	13:22	Cash	31.84	4.761905
998	2/22/2019	15:33	Cash	65.82	4.761905
999	2/18/2019	13:28	Cash	618.38	4.761905

	gross income	Rating
0	26.1415	9.1
1	3.8200	9.6
2	16.2155	7.4
3	23.2880	8.4
4	30.2085	5.3
..	...	...
995	2.0175	6.2
996	48.6900	4.4
997	1.5920	7.7
998	3.2910	4.1
999	30.9190	6.6

[1000 rows x 17 columns]