

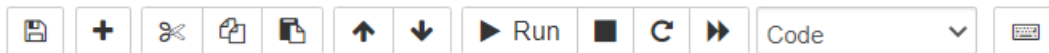
Icons for file operations (save, copy, paste, etc.) and a dropdown menu set to 'Code'.

```
In [1]: 1 import pandas as pd
        2 import numpy as np
        3 from matplotlib import pyplot as plt
        4 import seaborn as sb
```

```
In [10]: 1 data=pd.read_csv('Car_sales.csv')
        2 data
```

Out[10]:

le_type	Price_in_thousands	Engine_size	Horsepower	Wheelbase	Width	Length	Curb_weight	Fuel_capacity	Fuel_efficiency	Latest_Launch	Power_perf_factor
ssenger	21.50	1.8	140.0	101.2	67.3	172.4	2.639	13.2	28.0	2/2/2012	58.280150
ssenger	28.40	3.2	225.0	108.1	70.3	192.9	3.517	17.2	25.0	6/3/2011	91.370778
ssenger	NaN	3.2	225.0	106.9	70.6	192.0	3.470	17.2	26.0	1/4/2012	NaN
ssenger	42.00	3.5	210.0	114.6	71.4	196.6	3.850	18.0	22.0	3/10/2011	91.389779
ssenger	23.99	1.8	150.0	102.6	68.2	178.0	2.998	16.4	27.0	10/8/2011	62.777639
...
ssenger	24.40	1.9	160.0	100.5	67.6	176.6	3.042	15.8	25.0	9/21/2011	66.498812
ssenger	27.50	2.4	168.0	104.9	69.3	185.9	3.208	17.9	25.0	11/24/2012	70.654495
ssenger	28.80	2.4	168.0	104.9	69.3	186.2	3.259	17.9	25.0	6/25/2011	71.155978
ssenger	45.50	2.3	236.0	104.9	71.5	185.7	3.601	18.5	23.0	4/26/2011	101.623357
ssenger	36.00	2.9	201.0	109.9	72.1	189.8	3.600	21.1	24.0	11/14/2011	85.735655



In [11]: 1 data.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 157 entries, 0 to 156
Data columns (total 16 columns):
#   Column                      Non-Null Count  Dtype
---  ---
0   Manufacturer                157 non-null    object
1   Model                      157 non-null    object
2   Sales_in_thousands         157 non-null    float64
3   __year_resale_value        121 non-null    float64
4   Vehicle_type               157 non-null    object
5   Price_in_thousands         155 non-null    float64
6   Engine_size                 156 non-null    float64
7   Horsepower                  156 non-null    float64
8   Wheelbase                   156 non-null    float64
9   Width                       156 non-null    float64
10  Length                      156 non-null    float64
11  Curb_weight                 155 non-null    float64
12  Fuel_capacity               156 non-null    float64
13  Fuel_efficiency             154 non-null    float64
14  Latest_Launch              157 non-null    object
15  Power_perf_factor           155 non-null    float64
dtypes: float64(12), object(4)
memory usage: 19.8+ KB
```

In [12]: 1 data.head()

Out[12]:

	Manufacturer	Model	Sales_in_thousands	__year_resale_value	Vehicle_type	Price_in_thousands	Engine_size	Horsepower	Wheelbase	Width	Length	Curb_weight
0	Acura	Integra	16.919	16.360	Passenger	21.50	1.8	140.0	101.2	67.3	172.4	
1	Acura	TL	39.384	19.875	Passenger	28.40	3.2	225.0	108.1	70.3	192.9	

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car sales 2012

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In [12]:

1 data.head()

Out[12]:

	Manufacturer	Model	Sales_in_thousands	__year_resale_value	Vehicle_type	Price_in_thousands	Engine_size	Horsepower	Wheelbase	Width	Length	Curt
0	Acura	Integra	16.919	16.360	Passenger	21.50	1.8	140.0	101.2	67.3	172.4	
1	Acura	TL	39.384	19.875	Passenger	28.40	3.2	225.0	108.1	70.3	192.9	
2	Acura	CL	14.114	18.225	Passenger	NaN	3.2	225.0	106.9	70.6	192.0	
3	Acura	RL	8.588	29.725	Passenger	42.00	3.5	210.0	114.6	71.4	196.6	
4	Audi	A4	20.397	22.255	Passenger	23.99	1.8	150.0	102.6	68.2	178.0	

In [13]:

1 data.tail()

Out[13]:

Model	Sales_in_thousands	__year_resale_value	Vehicle_type	Price_in_thousands	Engine_size	Horsepower	Wheelbase	Width	Length	Curb_weight	Fuel_capa
V40	3.545	NaN	Passenger	24.4	1.9	160.0	100.5	67.6	176.6	3.042	
S70	15.245	NaN	Passenger	27.5	2.4	168.0	104.9	69.3	185.9	3.208	
V70	17.531	NaN	Passenger	28.8	2.4	168.0	104.9	69.3	186.2	3.259	
C70	3.493	NaN	Passenger	45.5	2.3	236.0	104.9	71.5	185.7	3.601	
S80	18.969	NaN	Passenger	36.0	2.9	201.0	109.9	72.1	189.8	3.600	

In [14]:

1 data.isnull().sum() # check null values

Out[14]:

Manufacturer 0
Model 0
Sales_in_thousands 0
__year_resale_value 36
Vehicle_type 0
Price_in_thousands 2
Engine_size 1
Horsepower 1
Wheelbase 1



```
In [17]: 1 a=data.drop(['__year_resale_value'],axis='columns',inplace=True)
```

```
In [18]: 1 data
```

```
Out[18]:
```

	Manufacturer	Model	Sales_in_thousands	Vehicle_type	Price_in_thousands	Engine_size	Horsepower	Wheelbase	Width	Length	Curb_weight	Fuel_capacity
0	Acura	Integra	16.919	Passenger	21.50	1.8	140.0	101.2	67.3	172.4	2.639	
1	Acura	TL	39.384	Passenger	28.40	3.2	225.0	108.1	70.3	192.9	3.517	
2	Acura	CL	14.114	Passenger	NaN	3.2	225.0	106.9	70.6	192.0	3.470	
3	Acura	RL	8.588	Passenger	42.00	3.5	210.0	114.6	71.4	196.6	3.850	
4	Audi	A4	20.397	Passenger	23.99	1.8	150.0	102.6	68.2	178.0	2.998	
...
152	Volvo	V40	3.545	Passenger	24.40	1.9	160.0	100.5	67.6	176.6	3.042	
153	Volvo	S70	15.245	Passenger	27.50	2.4	168.0	104.9	69.3	185.9	3.208	
154	Volvo	V70	17.531	Passenger	28.80	2.4	168.0	104.9	69.3	186.2	3.259	
155	Volvo	C70	3.493	Passenger	45.50	2.3	236.0	104.9	71.5	185.7	3.601	
156	Volvo	S80	18.969	Passenger	36.00	2.9	201.0	109.9	72.1	189.8	3.600	

157 rows × 15 columns



```
In [19]: 1 data.isnull().sum()
```

```
Out[19]: Manufacturer    0
Model                  0
Sales_in_thousands    0
Vehicle_type           0
Price_in_thousands    2
Engine_size            1
Horsepower             1
Wheelbase              1
Width                  1
Length                 1
Curb_weight            1
Fuel_capacity           1
```



dtype: int64

In [20]: 1 data.dropna(axis=0,inplace=True) # is remove null values

In [21]: 1 data.isnull().sum()

Out[21]: Manufacturer 0
Model 0
Sales_in_thousands 0
Vehicle_type 0
Price_in_thousands 0
Engine_size 0
Horsepower 0
Wheelbase 0
Width 0
Length 0
Curb_weight 0
Fuel_capacity 0
Fuel_efficiency 0
Latest_Launch 0
Power_perf_factor 0
dtype: int64

In [22]: 1 data

Out[22]:

	Manufacturer	Model	Sales_in_thousands	Vehicle_type	Price_in_thousands	Engine_size	Horsepower	Wheelbase	Width	Length	Curb_weight	Fuel_capa
0	Acura	Integra	16.919	Passenger	21.50	1.8	140.0	101.2	67.3	172.4	2.639	
1	Acura	TL	39.384	Passenger	28.40	3.2	225.0	108.1	70.3	192.9	3.517	
3	Acura	RL	8.588	Passenger	42.00	3.5	210.0	114.6	71.4	196.6	3.850	
4	Audi	A4	20.397	Passenger	23.99	1.8	150.0	102.6	68.2	178.0	2.998	
5	Audi	A6	18.780	Passenger	33.95	2.8	200.0	108.7	76.1	192.0	3.561	
...
152	Volvo	V40	3.545	Passenger	24.40	1.9	160.0	100.5	67.6	176.6	3.042	
153	Volvo	S70	15.245	Passenger	27.50	2.4	168.0	104.9	69.3	185.9	3.208	

```
In [32]: 1 data.dtypes
```

```
Out[32]: Manufacturer      object
         Model              object
         Sales_in_thousands float64
         Vehicle_type       object
         Price_in_thousands float64
         Engine_size        float64
         Horsepower         float64
         Wheelbase          float64
         Width              float64
         Length             float64
         Curb_weight        float64
         Fuel_capacity       float64
         Fuel_efficiency     float64
         Latest_Launch      object
         Power_perf_factor   float64
         dtype: object
```

```
In [37]: 1 data[data['Price_in_thousands']>=50]
```

Out[37]:

	Manufacturer	Model	Sales_in_thousands	Vehicle_type	Price_in_thousands	Engine_size	Horsepower	Wheelbase	Width	Length	Curb_weight	Fuel_cap
6	Audi	A8	1.380	Passenger	62.000	4.2	310.0	113.0	74.0	198.2	3.902	
39	Dodge	Viper	0.916	Passenger	69.725	8.0	450.0	96.2	75.7	176.7	3.375	
73	Lexus	LS400	6.375	Passenger	54.005	4.0	290.0	112.2	72.0	196.7	3.890	
74	Lexus	LX470	9.126	Car	60.105	4.7	230.0	112.2	76.4	192.5	5.401	
94	Mercedes-B	S-Class	16.774	Passenger	69.700	4.3	275.0	121.5	73.1	203.1	4.133	
95	Mercedes-B	SL-Class	3.311	Passenger	82.600	5.0	302.0	99.0	71.3	177.1	4.125	
99	Mercedes-B	CL500	0.954	Passenger	85.500	5.0	302.0	113.6	73.1	196.6	4.115	
125	Porsche	Carrera Coupe	1.280	Passenger	71.020	3.4	300.0	92.6	69.5	174.5	3.032	
126	Porsche	Carrera	1.033	Passenger	71.020	3.4	300.0	92.6	69.5	174.5	3.032	

dtype: object

In [37]: 1 data[data['Price_in_thousands']>=50]

Out[37]:

	Manufacturer	Model	Sales_in_thousands	Vehicle_type	Price_in_thousands	Engine_size	Horsepower	Wheelbase	Width	Length	Curb_weight	Fuel_cap
6	Audi	A8	1.380	Passenger	62.000	4.2	310.0	113.0	74.0	198.2	3.902	
39	Dodge	Viper	0.916	Passenger	69.725	8.0	450.0	96.2	75.7	176.7	3.375	
73	Lexus	LS400	6.375	Passenger	54.005	4.0	290.0	112.2	72.0	196.7	3.890	
74	Lexus	LX470	9.126	Car	60.105	4.7	230.0	112.2	76.4	192.5	5.401	
94	Mercedes-B	S-Class	16.774	Passenger	69.700	4.3	275.0	121.5	73.1	203.1	4.133	
95	Mercedes-B	SL-Class	3.311	Passenger	82.600	5.0	302.0	99.0	71.3	177.1	4.125	
99	Mercedes-B	CL500	0.954	Passenger	85.500	5.0	302.0	113.6	73.1	196.6	4.115	
125	Porsche	Carrera Coupe	1.280	Passenger	71.020	3.4	300.0	92.6	69.5	174.5	3.032	
126	Porsche	Carrera Cabrio	1.866	Passenger	74.970	3.4	300.0	92.6	69.5	174.5	3.075	
144	Toyota	Land Cruiser	9.835	Car	51.728	4.7	230.0	112.2	76.4	192.5	5.115	

In [41]: 1 data[data['Sales_in_thousands']>=120]

Out[41]:

	Manufacturer	Model	Sales_in_thousands	Vehicle_type	Price_in_thousands	Engine_size	Horsepower	Wheelbase	Width	Length	Curb_weight	Fuel_c
19	Chevrolet	Cavalier	145.519	Passenger	13.260	2.2	115.0	104.1	67.9	180.9	2.676	
20	Chevrolet	Malibu	135.126	Passenger	16.535	3.1	170.0	107.0	69.4	190.4	3.051	
40	Dodge	Ram Pickup	227.061	Car	19.460	5.2	230.0	138.7	79.3	224.2	4.470	
45	Dodge	Caravan	181.749	Car	19.565	2.4	150.0	113.3	76.8	186.3	3.533	
49	Ford	Taurus	245.815	Passenger	17.885	3.0	155.0	108.5	73.0	197.6	3.368	



In [41]: 1 data[data['Sales_in_thousands']>=120]

Out[41]:

	Manufacturer	Model	Sales_in_thousands	Vehicle_type	Price_in_thousands	Engine_size	Horsepower	Wheelbase	Width	Length	Curb_weight	Fuel_c
19	Chevrolet	Cavalier	145.519	Passenger	13.260	2.2	115.0	104.1	67.9	180.9	2.676	
20	Chevrolet	Malibu	135.126	Passenger	16.535	3.1	170.0	107.0	69.4	190.4	3.051	
40	Dodge	Ram Pickup	227.061	Car	19.460	5.2	230.0	138.7	79.3	224.2	4.470	
45	Dodge	Caravan	181.749	Car	19.565	2.4	150.0	113.3	76.8	186.3	3.533	
49	Ford	Taurus	245.815	Passenger	17.885	3.0	155.0	108.5	73.0	197.6	3.368	
50	Ford	Focus	175.670	Passenger	12.315	2.0	107.0	103.0	66.9	174.8	2.564	
52	Ford	Explorer	276.747	Car	31.930	4.0	210.0	111.6	70.2	190.7	3.876	
53	Ford	Windstar	155.787	Car	21.410	3.0	150.0	120.7	76.6	200.9	3.761	
54	Ford	Expedition	125.338	Car	36.135	4.6	240.0	119.0	78.7	204.6	4.808	
55	Ford	Ranger	220.650	Car	12.050	2.5	119.0	117.5	69.4	200.7	3.086	
56	Ford	F-Series	540.561	Car	26.935	4.6	220.0	138.5	79.1	224.5	4.241	
57	Honda	Civic	199.685	Passenger	12.885	1.6	106.0	103.2	67.1	175.1	2.339	
58	Honda	Accord	230.902	Passenger	15.350	2.3	135.0	106.9	70.3	188.8	2.932	
69	Jeep	Grand Cherokee	157.040	Car	26.895	4.0	195.0	105.9	72.3	181.5	3.880	
119	Pontiac	Grand Am	131.097	Passenger	19.720	3.4	175.0	107.0	70.4	186.3	3.091	
136	Toyota	Corolla	142.535	Passenger	13.108	1.8	120.0	97.0	66.7	174.0	2.420	
137	Toyota	Camry	247.994	Passenger	17.518	2.2	133.0	105.2	70.1	188.5	2.998	



137	Toyota	Camry	247.994	Passenger	17.518	2.2	133.0	105.2	70.1	188.5	2.998
-----	--------	-------	---------	-----------	--------	-----	-------	-------	------	-------	-------

In [42]: 1 data[data.Manufacturer=='Honda'] # only Honda company

Out[42]:

	Manufacturer	Model	Sales_in_thousands	Vehicle_type	Price_in_thousands	Engine_size	Horsepower	Wheelbase	Width	Length	Curb_weight	Fuel_cap
57	Honda	Civic	199.685	Passenger	12.885	1.6	106.0	103.2	67.1	175.1	2.339	
58	Honda	Accord	230.902	Passenger	15.350	2.3	135.0	106.9	70.3	188.8	2.932	
59	Honda	CR-V	73.203	Car	20.550	2.0	146.0	103.2	68.9	177.6	3.219	
60	Honda	Passport	12.855	Car	26.600	3.2	205.0	106.4	70.4	178.2	3.857	
61	Honda	Odyssey	76.029	Car	26.000	3.5	210.0	118.1	75.6	201.2	4.288	

In [43]: 1 data[data.Manufacturer=='BMW']

Out[43]:

	Manufacturer	Model	Sales_in_thousands	Vehicle_type	Price_in_thousands	Engine_size	Horsepower	Wheelbase	Width	Length	Curb_weight	Fuel_capacity
7	BMW	323i	19.747	Passenger	26.99	2.5	170.0	107.3	68.4	176.0	3.179	16.
8	BMW	328i	9.231	Passenger	33.40	2.8	193.0	107.3	68.5	176.0	3.197	16.
9	BMW	528i	17.527	Passenger	38.90	2.8	193.0	111.4	70.9	188.0	3.472	18.

In [47]: 1 from matplotlib import style
2 from matplotlib import figure

In [51]: 1 HONDA=data[data.Manufacturer=='Honda']

In [388]: 1 plt.bar(HONDA['Price_in_thousands'],HONDA['Sales_in_thousands'],color='r')
2 plt.title('Honda cars sales 2012',fontsize=20)
3 plt.ylabel('sales in thousands',fontsize=16)
4 plt.xlabel('prices in thousands',fontsize=16)
5 plt.gcf().set_size_inches(8,8)
6 plt.xticks(rotation=90)

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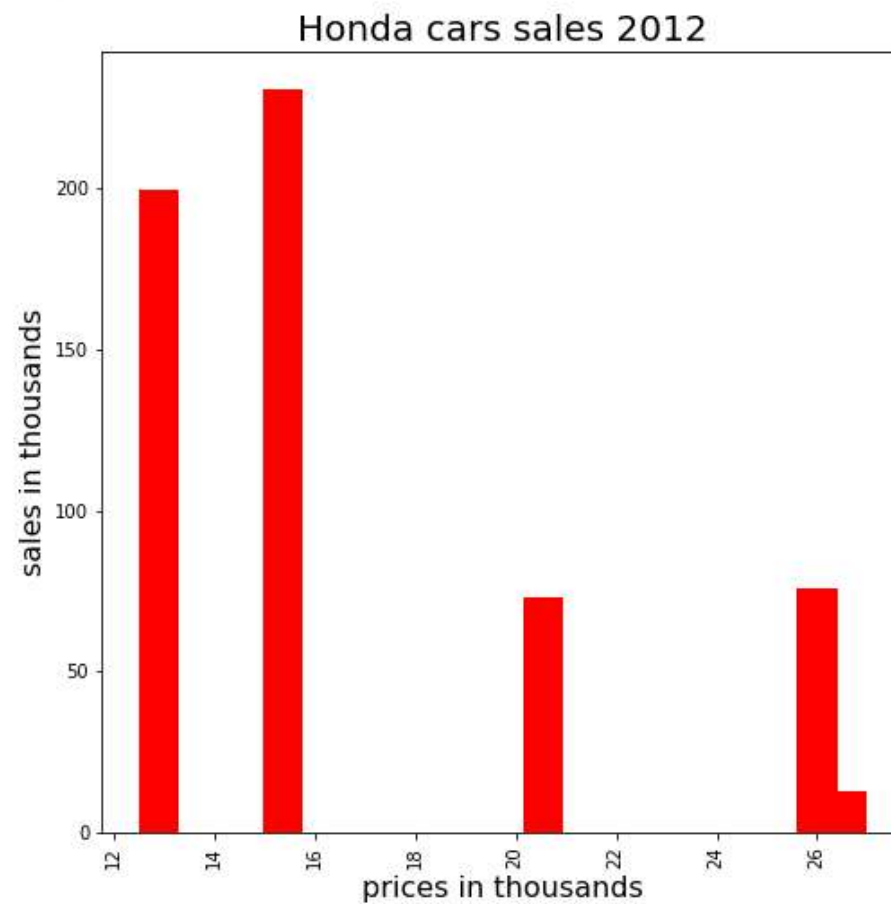
Python 3



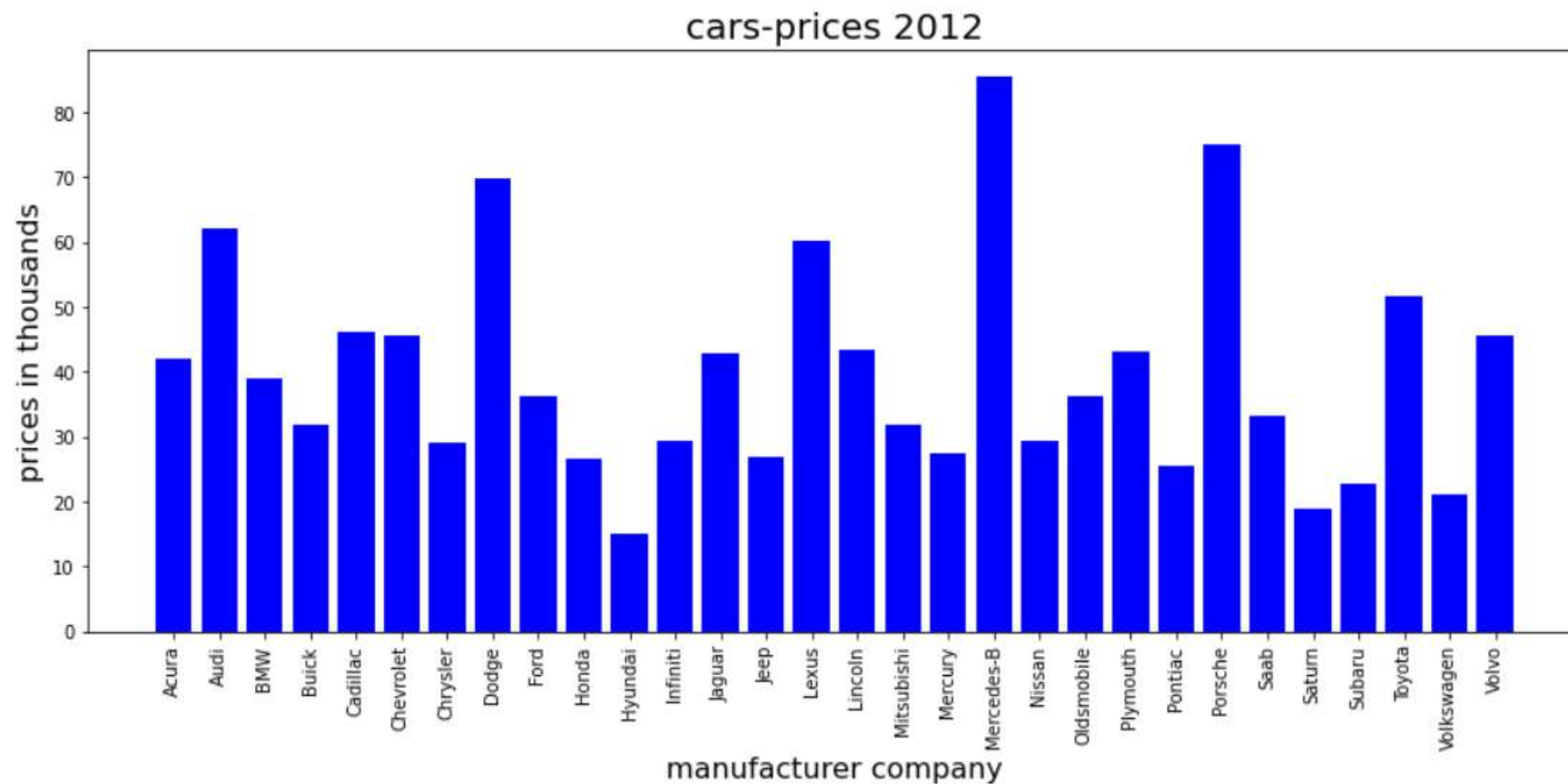
```
In [47]: 1 from matplotlib import style
        2 from matplotlib import figure
```

```
In [51]: 1 HONDA=data[data.Manufacturer=='Honda']
```

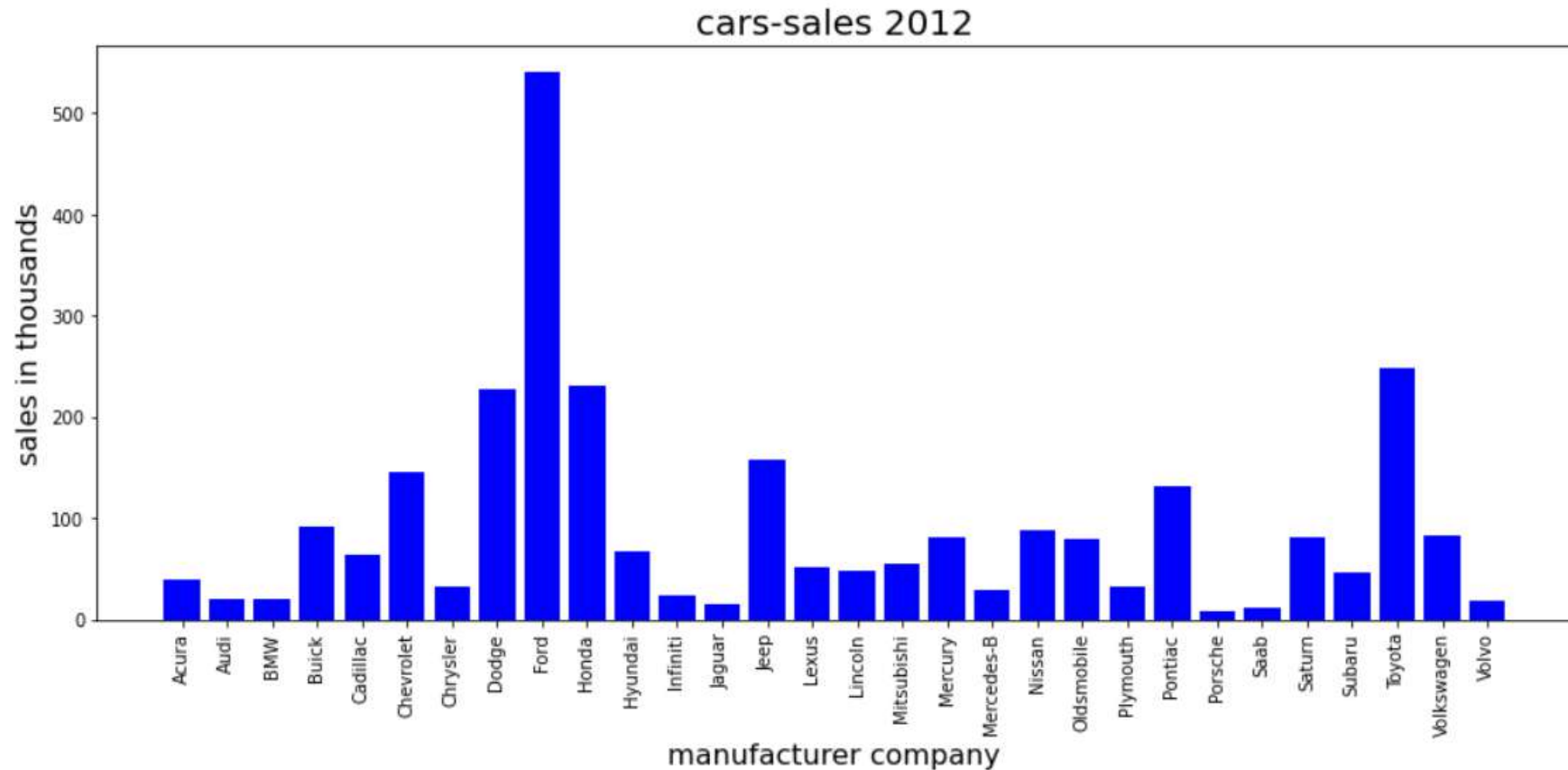
```
In [388]: 1 plt.bar(HONDA['Price_in_thousands'],HONDA['Sales_in_thousands'],color='r')
        2 plt.title('Honda cars sales 2012',fontsize=20)
        3 plt.ylabel('sales in thousands',fontsize=16)
        4 plt.xlabel('prices in thousands',fontsize=16)
        5 plt.gcf().set_size_inches(8,8)
        6 plt.xticks(rotation=90)
        7 plt.show()
```



```
In [389]: 1 plt.bar(data['Manufacturer'],data['Price_in_thousands'],color='b')
2 plt.title('cars-prices 2012',fontsize=20)
3 plt.ylabel('prices in thousands',fontsize=16)
4 plt.xlabel('manufacturer company',fontsize=16)
5 plt.gcf().set_size_inches(15,6)
6 plt.xticks(rotation=90)
7 plt.show()
```



```
In [381]: 1 plt.bar(data['Manufacturer'],data['Sales_in_thousands'],color='b')
2 plt.title('cars-sales 2012',fontsize=20)
3 plt.ylabel('sales in thousands',fontsize=16)
4 plt.xlabel('manufacturer company',fontsize=16)
5 plt.gcf().set_size_inches(15,6)
6 plt.xticks(rotation=90)
7 plt.show()
```





```
In [114]: 1 HONDA['Sales_in_thousands']
          2
```

```
Out[114]: 57    199.685
          58    230.902
          59     73.203
          60     12.855
          61     76.029
          Name: Sales_in_thousands, dtype: float64
```

```
In [115]: 1 HONDA
```

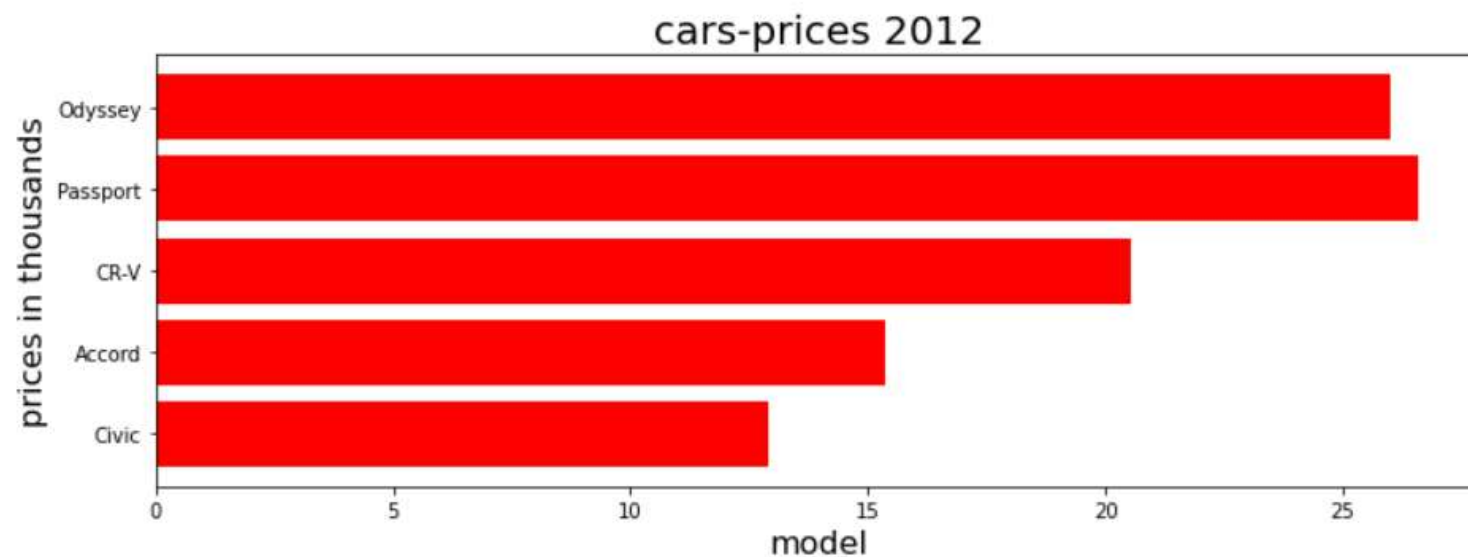
```
Out[115]:
```

	Manufacturer	Model	Sales_in_thousands	Vehicle_type	Price_in_thousands	Engine_size	Horsepower	Wheelbase	Width	Length	Curb_weight	Fuel_cap
57	Honda	Civic	199.685	Passenger	12.885	1.6	106.0	103.2	67.1	175.1	2.339	
58	Honda	Accord	230.902	Passenger	15.350	2.3	135.0	106.9	70.3	188.8	2.932	
59	Honda	CR-V	73.203	Car	20.550	2.0	146.0	103.2	68.9	177.6	3.219	
60	Honda	Passport	12.855	Car	26.600	3.2	205.0	106.4	70.4	178.2	3.857	
61	Honda	Odyssey	76.029	Car	26.000	3.5	210.0	118.1	75.6	201.2	4.288	

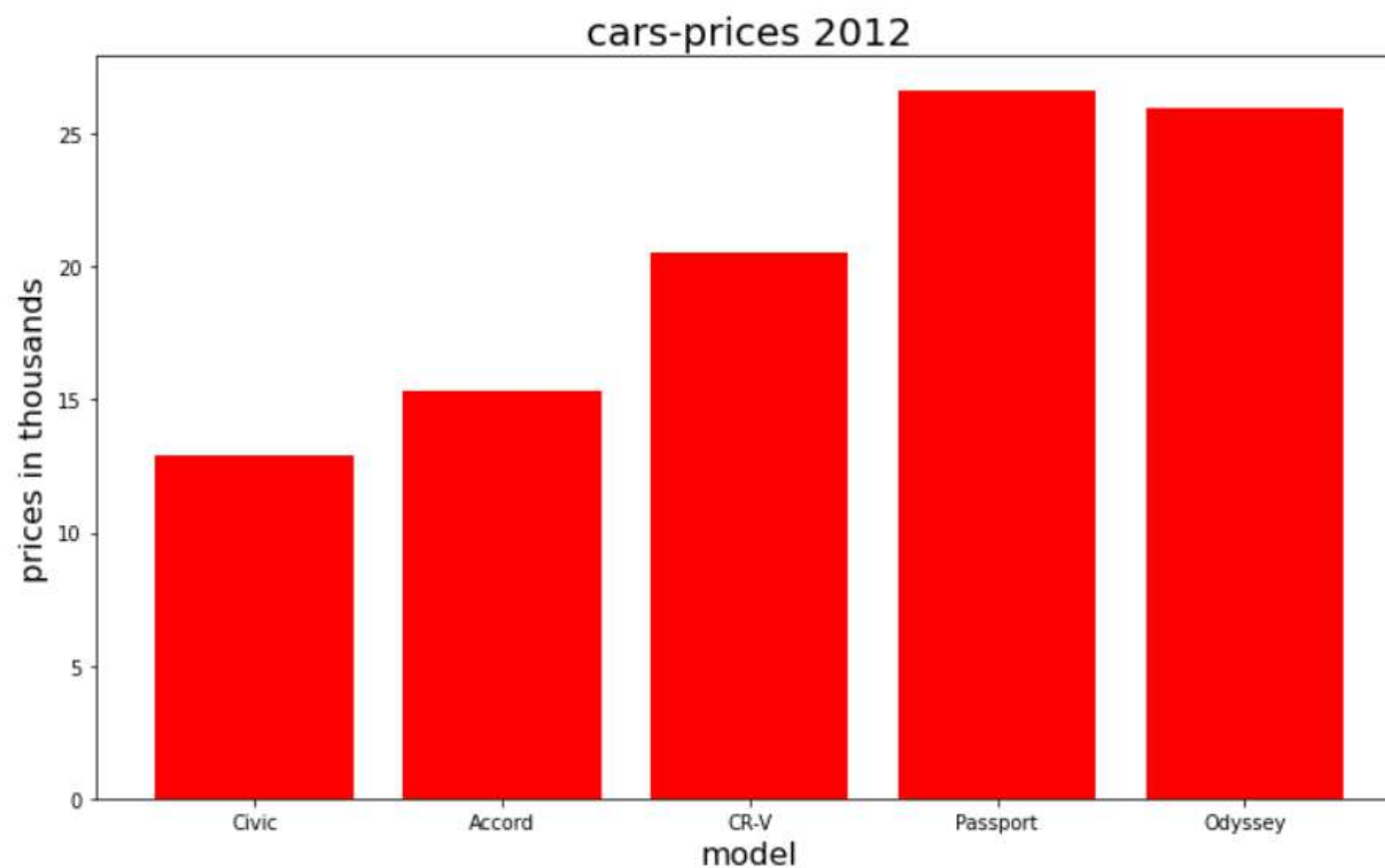
```
In [391]: 1 plt.barh(HONDA['Model'],HONDA['Price_in_thousands'],color='r') # using plt.barh
          2 plt.title('cars-prices 2012',fontsize=20)
          3 plt.ylabel('prices in thousands',fontsize=16)
          4 plt.xlabel('model',fontsize=16)
          5 plt.gcf().set_size_inches(12,4)
          6
          7 plt.show()
```



```
In [391]: 1 plt.barh(HONDA['Model'],HONDA['Price_in_thousands'],color='r') # using plt.barh
2 plt.title('cars-prices 2012',fontsize=20)
3 plt.ylabel('prices in thousands',fontsize=16)
4 plt.xlabel('model',fontsize=16)
5 plt.gcf().set_size_inches(12,4)
6
7 plt.show()
```




```
In [392]: 1 plt.bar(HONDA['Model'],HONDA['Price_in_thousands'],color='r') # using plt.barh
2 plt.title('cars-prices 2012',fontsize=20)
3 plt.ylabel('prices in thousands',fontsize=16)
4 plt.xlabel('model',fontsize=16)
5 plt.gcf().set_size_inches(12,7)
6
7 plt.show()
```





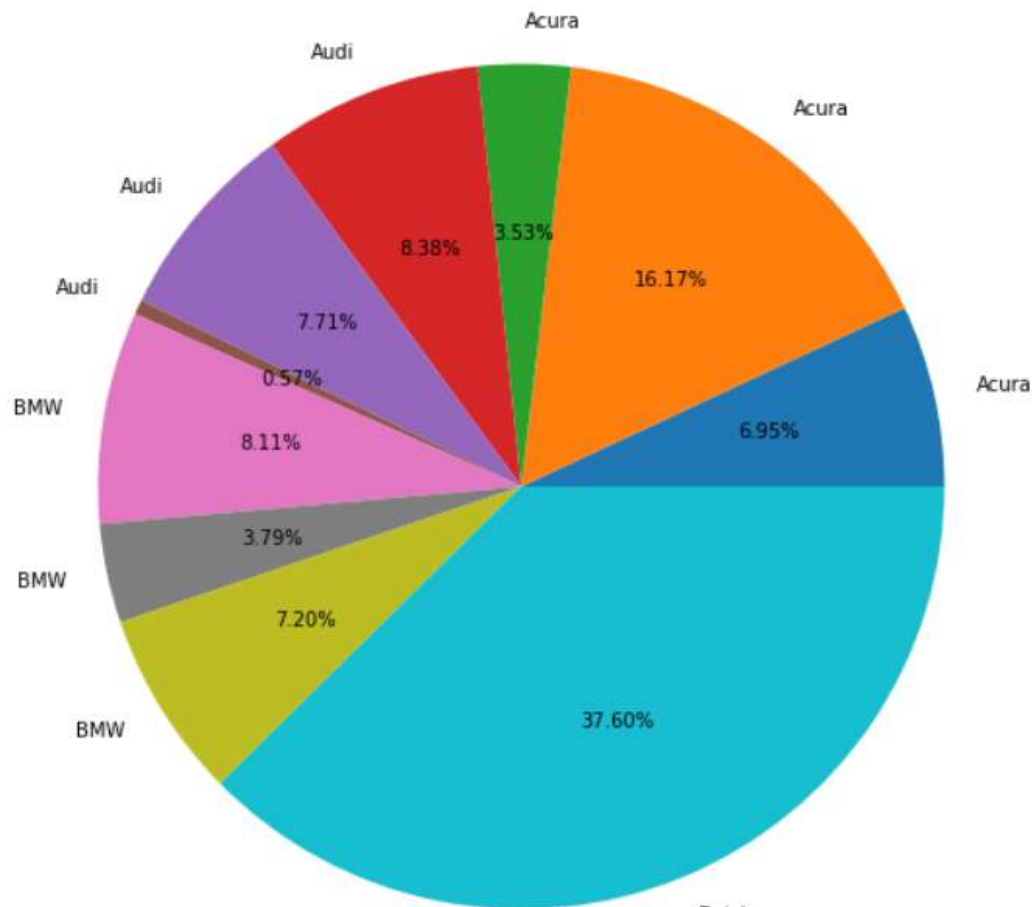
```
In [271]: 1 data['Manufacturer'].value_counts().head(5)
```

```
Out[271]: Ford          11  
Dodge             10  
Chevrolet          9  
Mercedes-B         9  
Toyota             9  
Name: Manufacturer, dtype: int64
```



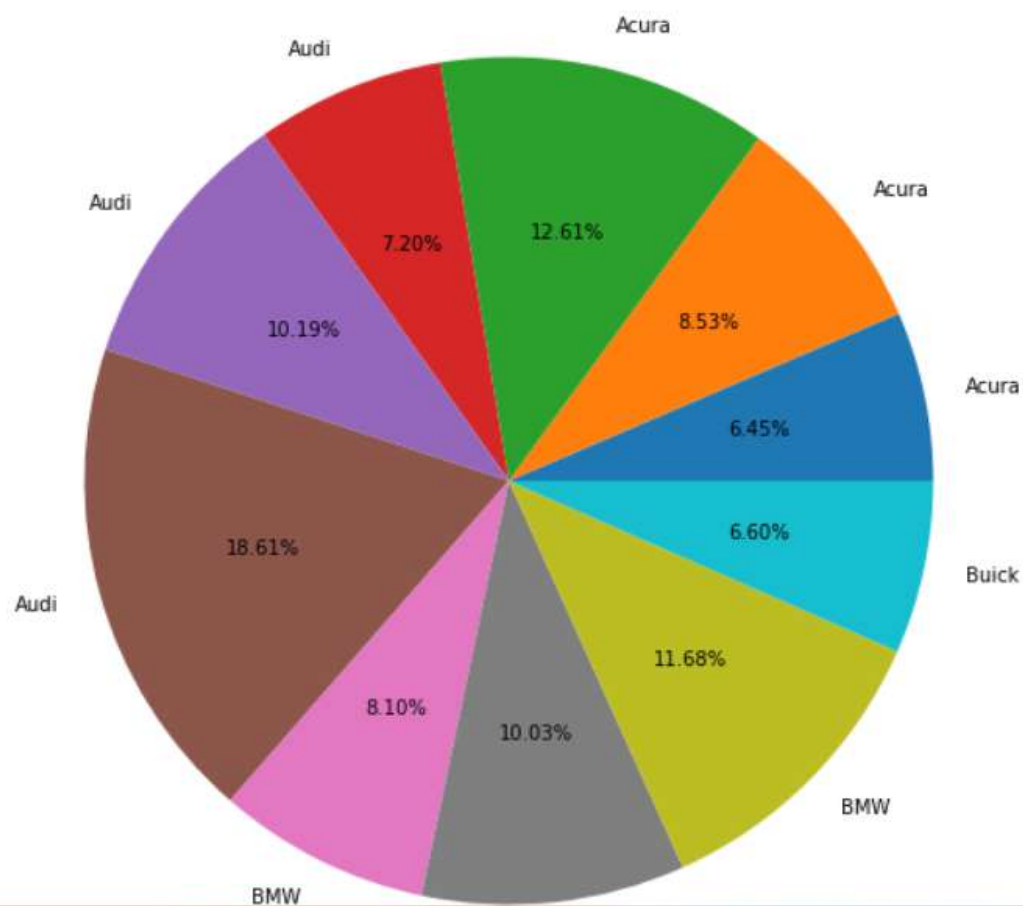
In [396]:

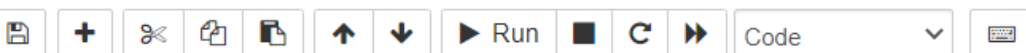
```
1 x=data['Manufacturer'].head(10)
2 y=data['Sales_in_thousands'].head(10)
3 plt.pie(y,labels=x,autopct='%2.2f%%')
4 plt.gcf().set_size_inches(10,10)
5 plt.show()
```





```
In [395]: 1 x=data['Price_in_thousands'].head(10)
2 y=data['Manufacturer'].head(10)
3 plt.pie(x,labels=y,autopct='%2.2f%%')
4 plt.gcf().set_size_inches(10,10)
5 plt.show()
```





In [334]: 1 data

Out[334]:

	Manufacturer	Model	Sales_in_thousands	Vehicle_type	Price_in_thousands	Engine_size	Horsepower	Wheelbase	Width	Length	Curb_weight	Fuel_capacity
0	Acura	Integra	16.919	Passenger	21.50	1.8	140.0	101.2	67.3	172.4	2.639	
1	Acura	TL	39.384	Passenger	28.40	3.2	225.0	108.1	70.3	192.9	3.517	
3	Acura	RL	8.588	Passenger	42.00	3.5	210.0	114.6	71.4	196.6	3.850	
4	Audi	A4	20.397	Passenger	23.99	1.8	150.0	102.6	68.2	178.0	2.998	
5	Audi	A6	18.780	Passenger	33.95	2.8	200.0	108.7	76.1	192.0	3.561	
...
152	Volvo	V40	3.545	Passenger	24.40	1.9	160.0	100.5	67.6	176.6	3.042	
153	Volvo	S70	15.245	Passenger	27.50	2.4	168.0	104.9	69.3	185.9	3.208	
154	Volvo	V70	17.531	Passenger	28.80	2.4	168.0	104.9	69.3	186.2	3.259	
155	Volvo	C70	3.493	Passenger	45.50	2.3	236.0	104.9	71.5	185.7	3.601	
156	Volvo	S80	18.969	Passenger	36.00	2.9	201.0	109.9	72.1	189.8	3.600	

152 rows × 15 columns



152 rows × 15 columns

```
In [402]: 1 x=data['Manufacturer'].head(80)
2 y=data['Sales_in_thousands'].head(80)
3 plt.xlabel('Manufacturer-company',fontsize=16)
4 plt.ylabel('Sales_in_thousands',fontsize=16)
5 plt.scatter(x,y,color='k')
6 plt.gcf().set_size_inches(15,4)
7 plt.show()
```

