

```
from google.colab import files
files.upload()
```

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Saving clean_df.csv to clean_df.csv

{'clean_df.csv': b',svmboling,normalized-losses,make,num-of-doors,body-style,drive-wheel

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
df=pd.read_csv('clean_df.csv')
df.head()
```

↗

	Unnamed: 0	symboling	normalized-losses	make	num-of-doors	body-style	drive-wheels	engine-location	wheel-base
0	0	3	122	alfa-romero	two	convertible	rwd	front	88.6
1	1	3	122	alfa-romero	two	convertible	rwd	front	88.6
2	2	1	122	alfa-romero	two	hatchback	rwd	front	94.5
3	3	2	164	audi	four	sedan	fwd	front	99.8
4	4	2	164	audi	four	sedan	4wd	front	99.4

```
df.describe()
```

	Unnamed: 0	symboling	normalized-losses	wheel-base	length	width	l
count	201	201	201	201	201	201	201

df.dtypes

```
Unnamed: 0      int64
symboling      int64
normalized-losses  int64
make           object
num-of-doors    object
body-style      object
drive-wheels    object
engine-location  object
wheel-base     float64
length         float64
width          float64
height         float64
curb-weight     int64
engine-type     object
num-of-cylinders object
engine-size     int64
fuel-system     object
bore           float64
stroke         float64
compression-ratio float64
horsepower     float64
peak-rpm       float64
city-mpg       int64
highway-mpg    int64
Price          float64
city-1/100km   float64
highway-mpg-1  float64
horsepower-binned object
diesel         int64
gas            int64
aspiration-std  int64
aspiration-turbo int64
dtype: object
```

df.shape

```
(201, 32)
```

df.ndim

```
2
```

```
b=df.describe(include=["object"])
b
```

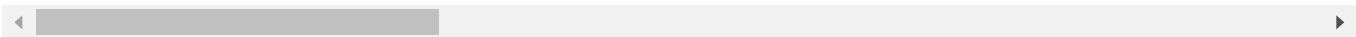
	make	num- of- doors	body- style	drive- wheels	engine- location	engine- type	num-of- cylinders	fuel- system	horsepower- binned
count	201	199	201	201	201	201	201	201	200
unique	22	2	5	3	2	6	7	8	3
top	toyota	four	sedan	fwd	front	ohc	four	mpfi	Low

b.shape

(4, 9)

df.describe(include=[np.number])

	Unnamed: 0	symboling	normalized- losses	wheel- base	length	width	height
count	201.000000	201.000000	201.00000	201.000000	201.000000	201.000000	201.000000
mean	100.000000	0.840796	122.00000	98.797015	0.837102	0.915126	0.899108
std	58.167861	1.254802	31.99625	6.066366	0.059213	0.029187	0.040933
min	0.000000	-2.000000	65.00000	86.600000	0.678039	0.837500	0.799331
25%	50.000000	0.000000	101.00000	94.500000	0.801538	0.890278	0.869565
50%	100.000000	1.000000	122.00000	97.000000	0.832292	0.909722	0.904682
75%	150.000000	2.000000	137.00000	102.400000	0.881788	0.925000	0.928094
max	200.000000	3.000000	256.00000	120.900000	1.000000	1.000000	1.000000



a=df.describe(include="all")
a

	Unnamed: 0	symboling	normalized- losses	make	num- of- doors	body- style	drive- wheels	engine- location
count	201.000000	201.000000	201.00000	201	199	201	201	201
unique	NaN	NaN	NaN	22	2	5	3	2
top	NaN	NaN	NaN	toyota	four	sedan	fwd	front
freq	NaN	NaN	NaN	32	113	94	118	198

a.shape

(11, 32)

0 000000 0 000000 05 00000 0000 0000 0000 0000 0000

df.axes

```
[RangeIndex(start=0, stop=201, step=1),
Index(['Unnamed: 0', 'symboling', 'normalized-losses', 'make', 'num-of-doors',
      'body-style', 'drive-wheels', 'engine-location', 'wheel-base', 'length',
      'width', 'height', 'curb-weight', 'engine-type', 'num-of-cylinders',
      'engine-size', 'fuel-system', 'bore', 'stroke', 'compression-ratio',
      'horsepower', 'peak-rpm', 'city-mpg', 'highway-mpg', 'Price',
      'city-1/100km', 'highway-mpg-1', 'horsepower-binned', 'diesel', 'gas',
      'aspiration-std', 'aspiration-turbo'],
      dtype='object')]
```

df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 201 entries, 0 to 200
Data columns (total 32 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Unnamed: 0            201 non-null   int64
1   symboling              201 non-null   int64
2   normalized-losses     201 non-null   int64
3   make                  201 non-null   object
4   num-of-doors          199 non-null   object
5   body-style            201 non-null   object
6   drive-wheels          201 non-null   object
7   engine-location       201 non-null   object
8   wheel-base           201 non-null   float64
9   length                201 non-null   float64
10  width                 201 non-null   float64
11  height                201 non-null   float64
12  curb-weight           201 non-null   int64
13  engine-type           201 non-null   object
14  num-of-cylinders      201 non-null   object
15  engine-size           201 non-null   int64
16  fuel-system           201 non-null   object
```

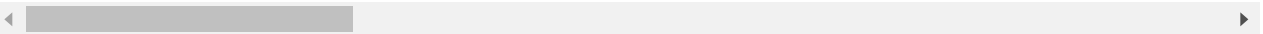
```
17 bore 197 non-null float64
18 stroke 201 non-null float64
19 compression-ratio 201 non-null float64
20 horsepower 201 non-null float64
21 peak-rpm 201 non-null float64
22 city-mpg 201 non-null int64
23 highway-mpg 201 non-null int64
24 Price 201 non-null float64
25 city-1/100km 201 non-null float64
26 highway-mpg-1 201 non-null float64
27 horsepower-binned 200 non-null object
28 diesel 201 non-null int64
29 gas 201 non-null int64
30 aspiration-std 201 non-null int64
31 aspiration-turbo 201 non-null int64
dtypes: float64(12), int64(11), object(9)
memory usage: 50.4+ KB
```

```
df.columns
```

```
Index(['Unnamed: 0', 'symboling', 'normalized-losses', 'make', 'num-of-doors',
      'body-style', 'drive-wheels', 'engine-location', 'wheel-base', 'length',
      'width', 'height', 'curb-weight', 'engine-type', 'num-of-cylinders',
      'engine-size', 'fuel-system', 'bore', 'stroke', 'compression-ratio',
      'horsepower', 'peak-rpm', 'city-mpg', 'highway-mpg', 'Price',
      'city-1/100km', 'highway-mpg-1', 'horsepower-binned', 'diesel', 'gas',
      'aspiration-std', 'aspiration-turbo'],
      dtype='object')
```

```
df.tail()
```

	Unnamed: 0	symboling	normalized-losses	make	num-of-doors	body-style	drive-wheels	engine-location	wheel-base
196	196	-1	95	volvo	four	sedan	rwd	front	109
197	197	-1	95	volvo	four	sedan	rwd	front	109
198	198	-1	95	volvo	four	sedan	rwd	front	109
199	199	-1	95	volvo	four	sedan	rwd	front	109
200	200	-1	95	volvo	four	sedan	rwd	front	109



```
df['drive-wheels'].value_counts()
```

```
fwd    118
rwd     75
```

```
4wd      8
Name: drive-wheels, dtype: int64
```

```
df['make'].value_counts()
```

```
toyota      32
nissan      18
mazda       17
honda       13
mitsubishi  13
subaru      12
volkswagen  12
peugot      11
volvo       11
dodge        9
mercedes-benz 8
bmw          8
plymouth     7
saab         6
audi         6
porsche      4
jaguar       3
chevrolet    3
alfa-romero  3
isuzu        2
renault      2
mercury      1
Name: make, dtype: int64
```

```
drive_wheels_counts = df['drive-wheels'].value_counts().to_frame()
drive_wheels_counts.rename(columns={'drive-wheels': 'value_counts'}, inplace=True)
drive_wheels_counts
```

	value_counts
fwd	118
rwd	75
4wd	8

```
#give name to index
drive_wheels_counts.index.name="wheels"
drive_wheels_counts
```

df

	Unnamed: 0	symboling	normalized- losses	make	num- of- doors	body- style	value_counts	en loc
0	0	3	122	alfa-romero	two	convertible	rwd	
1	1	3	122	alfa-romero	two	convertible	rwd	
2	2	1	122	alfa-romero	two	hatchback	rwd	
3	3	2	164	audi	four	sedan	fwd	
4	4	2	164	audi	four	sedan	4wd	
...	
196	196	-1	95	volvo	four	sedan	rwd	
197	197	-1	95	volvo	four	sedan	rwd	
198	198	-1	95	volvo	four	sedan	rwd	
199	199	-1	95	volvo	four	sedan	rwd	
200	200	-1	95	volvo	four	sedan	rwd	

201 rows × 32 columns



```
df.rename(columns={'value_counts': 'drive-wheels'}, inplace=True)
df
```

	Unnamed: 0	symboling	normalized- losses	make	num- of- doors	body- style	drive- wheels	engine- location	wheel bas
0	0	3	122	alfa-romero	two	convertible	rwd	front	88.
1	1	3	122	alfa-romero	two	convertible	rwd	front	88.
2	2	1	122	alfa-romero	two	hatchback	rwd	front	94.

```
df['drive-wheels'].unique()
```

```
array(['rwd', 'fwd', '4wd'], dtype=object)
```

```
...      ...      ...      ...      ...      ...      ...      ...      ...      ...
```

```
df_group_one=df[['drive-wheels','body-style','Price']]
```

```
df_group_one
```

	drive-wheels	body-style	Price
0	rwd	convertible	13495.0
1	rwd	convertible	16500.0
2	rwd	hatchback	16500.0
3	fwd	sedan	13950.0
4	4wd	sedan	17450.0
...
196	rwd	sedan	16845.0
197	rwd	sedan	19045.0
198	rwd	sedan	21485.0
199	rwd	sedan	22470.0
200	rwd	sedan	22625.0

201 rows × 3 columns

▼ Groupby

```
df_group_one=df[['drive-wheels','body-style','Price']]
```

```
df_group_one=df_group_one.groupby(['drive-wheels'],as_index= False).mean()
```

```
df_group_one
```


	drive-wheels	Price
0	4wd	10241.000000
1	fwd	9244.779661
2	rwd	19757.613333

```
df_gptest=df[['drive-wheels','body-style','Price']]
grouped_test1=df_gptest.groupby(['drive-wheels','body-style'],as_index= False).mean()
grouped_test1
```

	drive-wheels	body-style	Price
0	4wd	hatchback	7603.000000
1	4wd	sedan	12647.333333
2	4wd	wagon	9095.750000
3	fwd	convertible	11595.000000
4	fwd	hardtop	8249.000000
5	fwd	hatchback	8396.387755
6	fwd	sedan	9811.800000
7	fwd	wagon	9997.333333
8	rwd	convertible	23949.600000
9	rwd	hardtop	24202.714286
10	rwd	hatchback	14337.777778
11	rwd	sedan	21711.833333
12	rwd	wagon	16994.222222

```
# Use the "groupby" function to find the average "price" of each car based on "body-style" ?
df_group_body_price = df[['Price', 'body-style']]
df_group_3 =df_group_body_price.groupby(['body-style'],as_index= False).mean()
df_group_3
```

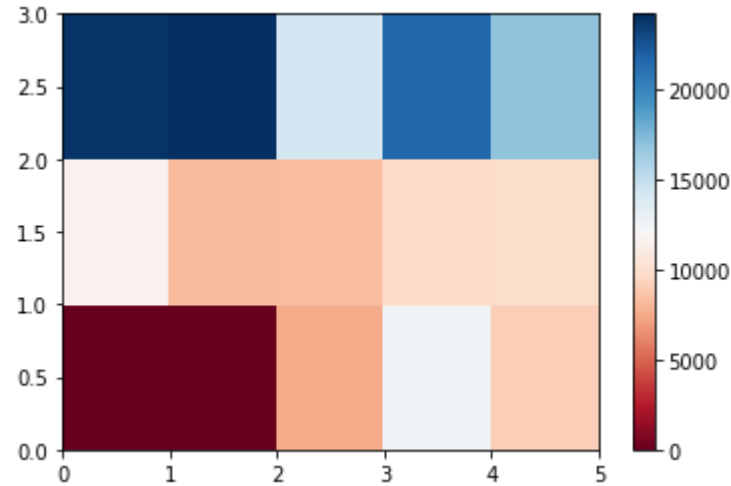
```
body-style      Price
grouped_pivot=grouped_test1.pivot(index='drive-wheels',columns='body-style')
grouped_pivot
```

	Price				
body-style	convertible	hardtop	hatchback	sedan	wagon
drive-wheels					
4wd	NaN	NaN	7603.000000	12647.333333	9095.750000
fwd	11595.0	8249.000000	8396.387755	9811.800000	9997.333333
rwd	23949.6	24202.714286	14337.777778	21711.833333	16994.222222

```
grouped_pivot=grouped_pivot.fillna(0) #fill missing values with 0
grouped_pivot
```

	Price				
body-style	convertible	hardtop	hatchback	sedan	wagon
drive-wheels					
4wd	0.0	0.000000	7603.000000	12647.333333	9095.750000
fwd	11595.0	8249.000000	8396.387755	9811.800000	9997.333333
rwd	23949.6	24202.714286	14337.777778	21711.833333	16994.222222

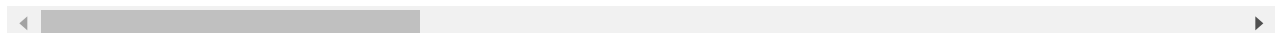
```
plt.pcolor(grouped_pivot,cmap='RdBu')
plt.colorbar()
plt.show()
```



Numeric variables Non_Graphical : Correlation lies between (-1 to 1)

```
#
df.corr()
```

	Unnamed: 0	symboling	normalized-losses	wheel-base	length	width	height
Unnamed: 0	1.000000	-0.162764	-0.241092	0.125517	0.161848	0.043976	0.252015
symboling	-0.162764	1.000000	0.466264	-0.535987	-0.365404	-0.242423	-0.550160
normalized-losses	-0.241092	0.466264	1.000000	-0.056661	0.019424	0.086802	-0.373737
wheel-base	0.125517	-0.535987	-0.056661	1.000000	0.876024	0.814507	0.590742
length	0.161848	-0.365404	0.019424	0.876024	1.000000	0.857170	0.492063
width	0.043976	-0.242423	0.086802	0.814507	0.857170	1.000000	0.306002
height	0.252015	-0.550160	-0.373737	0.590742	0.492063	0.306002	1.000000
curb-weight	0.064820	-0.233118	0.099404	0.782097	0.880665	0.866201	0.866201
engine-size	-0.047764	-0.110581	0.112360	0.572027	0.685025	0.729436	0.685025
bore	0.246289	-0.144324	-0.030035	0.494884	0.610051	0.544924	0.610051
stroke	-0.162490	-0.008153	0.055045	0.158018	0.123952	0.188822	-0.008153
compression-ratio	0.144301	-0.182196	-0.114713	0.250313	0.159733	0.189867	0.250313
horsepower	-0.022474	0.075819	0.217299	0.371147	0.579821	0.615077	-0.022474
peak-rpm	-0.195662	0.279740	0.239543	-0.360305	-0.285970	-0.245800	-0.195662
city-mpg	0.027956	-0.035527	-0.225016	-0.470606	-0.665192	-0.633531	-0.035527
highway-mpg	0.020344	0.036233	-0.181877	-0.543304	-0.698142	-0.680635	-0.181877
Price	-0.118214	-0.082391	0.133999	0.584642	0.690628	0.751265	0.133999
city-1/100km	-0.099157	0.066171	0.238567	0.476153	0.657373	0.673363	0.238567
highway-mpg-1	-0.078346	-0.029807	0.181189	0.577576	0.707108	0.736728	0.181189
diesel	0.121454	-0.196735	-0.101546	0.307237	0.211187	0.244356	0.211187
gas	-0.121454	0.196735	0.101546	-0.307237	-0.211187	-0.244356	-0.211187
aspiration-std	-0.082739	0.054615	0.006911	-0.256889	-0.230085	-0.305732	-0.006911
aspiration-turbo	0.082739	-0.054615	-0.006911	0.256889	0.230085	0.305732	0.006911



```
df[["wheel-base", 'Price']].corr()
```

	wheel-base	Price
wheel-base	1.000000	0.584642
Price	0.584642	1.000000

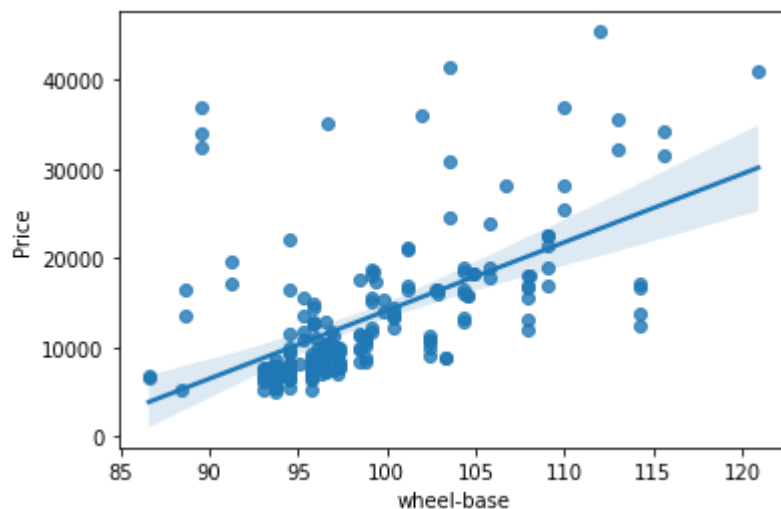
```
df[["engine-size", "Price", "horsepower"]].corr()
```

	engine-size	Price	horsepower
engine-size	1.000000	0.872335	0.822676
Price	0.872335	1.000000	0.809575
horsepower	0.822676	0.809575	1.000000

```
# for graphical we can use scatterplot
```

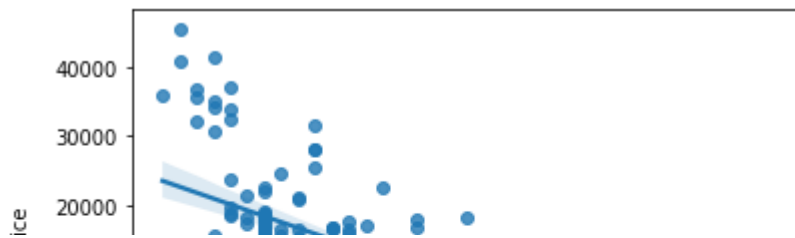
```
sns.regplot(x="wheel-base", y="Price", data=df)
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7fed0f452b50>
```



```
sns.regplot(x="city-mpg", y="Price", data=df)
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7fed0f3e4250>
```



```
sns.boxplot(x="drive-wheels",y="Price",data=df)
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7fed0ee4d090>
```

