

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

<IPython.core.display.HTML object>
```

Saving auto.csv to auto.csv

```
import pandas as pd
path = 'auto.csv'
df = pd.read_csv(path)
df.tail(5)
```

```
      3    ? alfa-romero    gas    std    two convertible    rwd    front
88.6  \
199 -1  95      volvo    gas    std    four      sedan    rwd    front
109.1
200 -1  95      volvo    gas    turbo    four      sedan    rwd    front
109.1
201 -1  95      volvo    gas    std    four      sedan    rwd    front
109.1
202 -1  95      volvo    diesel    turbo    four      sedan    rwd    front
109.1
203 -1  95      volvo    gas    turbo    four      sedan    rwd    front
109.1
```

```
      ...  130  mpfi  3.47  2.68  9.0  111  5000  21  27  13495
199  ...  141  mpfi  3.78  3.15  9.5  114  5400  23  28  16845
200  ...  141  mpfi  3.78  3.15  8.7  160  5300  19  25  19045
201  ...  173  mpfi  3.58  2.87  8.8  134  5500  18  23  21485
202  ...  145  idi   3.01  3.40  23.0  106  4800  26  27  22470
203  ...  141  mpfi  3.78  3.15  9.5  114  5400  19  25  22625
```

[5 rows x 26 columns]

```
headers = ["symboling", "normalized-losses", "make", "fuel-
type", "aspiration", "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"]
print("headers\n", headers)
```

```
headers
['symboling', 'normalized-losses', 'make', 'fuel-type', 'aspiration',
'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']
```

```
df.columns = headers
df.head(10)
```

	symboling	normalized-losses	make	fuel-type	aspiration	num-
of-doors \						
0	3	?	alfa-romero	gas	std	
two						
1	1	?	alfa-romero	gas	std	
two						
2	2	164	audi	gas	std	
four						
3	2	164	audi	gas	std	
four						
4	2	?	audi	gas	std	
two						
5	1	158	audi	gas	std	
four						
6	1	?	audi	gas	std	
four						
7	1	158	audi	gas	turbo	
four						
8	0	?	audi	gas	turbo	
two						
9	2	192	bmw	gas	std	
two						

	body-style	drive-wheels	engine-location	wheel-base	...	engine-
size \						
0	convertible	rwd	front	88.6	...	
130						
1	hatchback	rwd	front	94.5	...	
152						
2	sedan	fwd	front	99.8	...	
109						
3	sedan	4wd	front	99.4	...	
136						
4	sedan	fwd	front	99.8	...	
136						
5	sedan	fwd	front	105.8	...	
136						
6	wagon	fwd	front	105.8	...	
136						
7	sedan	fwd	front	105.8	...	
131						
8	hatchback	4wd	front	99.5	...	
131						
9	sedan	rwd	front	101.2	...	
108						

```
fuel-system bore stroke compression-ratio horsepower peak-rpm
```

city-mpg	\					
0	mpfi	3.47	2.68	9.0	111	5000
21						
1	mpfi	2.68	3.47	9.0	154	5000
19						
2	mpfi	3.19	3.40	10.0	102	5500
24						
3	mpfi	3.19	3.40	8.0	115	5500
18						
4	mpfi	3.19	3.40	8.5	110	5500
19						
5	mpfi	3.19	3.40	8.5	110	5500
19						
6	mpfi	3.19	3.40	8.5	110	5500
19						
7	mpfi	3.13	3.40	8.3	140	5500
17						
8	mpfi	3.13	3.40	7.0	160	5500
16						
9	mpfi	3.50	2.80	8.8	101	5800
23						

highway-mpg	price
0	27 16500
1	26 16500
2	30 13950
3	22 17450
4	25 15250
5	25 17710
6	25 18920
7	20 23875
8	22 ?
9	29 16430

[10 rows x 26 columns]

df.dtypes

symboling	int64
normalized-losses	object
make	object
fuel-type	object
aspiration	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             object
stroke          object
compression-ratio float64
horsepower       object
peak-rpm         object
city-mpg         int64
highway-mpg      int64
price           object
dtype: object

```

```
df["price"] = df["price"].astype("float")
```

```

-----
-----
ValueError                                Traceback (most recent call
last)
<ipython-input-6-564c5f370205> in <module>
----> 1 df["price"] = df["price"].astype("float")

/usr/local/lib/python3.8/dist-packages/pandas/core/generic.py in
astype(self, dtype, copy, errors)
    5813         else:
    5814             # else, only a single dtype is given
-> 5815             new_data = self._mgr.astype(dtype=dtype,
copy=copy, errors=errors)
    5816             return
self._constructor(new_data).__finalize__(self, method="astype")
    5817

/usr/local/lib/python3.8/dist-packages/pandas/core/internals/managers.
py in astype(self, dtype, copy, errors)
    416
    417     def astype(self: T, dtype, copy: bool = False, errors: str
= "raise") -> T:
--> 418         return self.apply("astype", dtype=dtype, copy=copy,
errors=errors)
    419
    420     def convert(

/usr/local/lib/python3.8/dist-packages/pandas/core/internals/managers.
py in apply(self, f, align_keys, ignore_failures, **kwargs)
    325         applied = b.apply(f, **kwargs)
    326     else:
-> 327         applied = getattr(b, f)(**kwargs)

```

```

328             except (TypeError, NotImplementedError):
329                 if not ignore_failures:

/usr/local/lib/python3.8/dist-packages/pandas/core/internals/blocks.py
in astype(self, dtype, copy, errors)
    589         values = self.values
    590
--> 591         new_values = astype_array_safe(values, dtype,
copy=copy, errors=errors)
    592
    593         new_values = maybe_coerce_values(new_values)

/usr/local/lib/python3.8/dist-packages/pandas/core/dtypes/cast.py in
astype_array_safe(values, dtype, copy, errors)
    1307
    1308     try:
-> 1309         new_values = astype_array(values, dtype, copy=copy)
    1310     except (ValueError, TypeError):
    1311         # e.g. astype_nansafe can fail on object-dtype of
strings

/usr/local/lib/python3.8/dist-packages/pandas/core/dtypes/cast.py in
astype_array(values, dtype, copy)
    1255
    1256     else:
-> 1257         values = astype_nansafe(values, dtype, copy=copy)
    1258
    1259     # in pandas we don't store numpy str dtypes, so convert to
object

/usr/local/lib/python3.8/dist-packages/pandas/core/dtypes/cast.py in
astype_nansafe(arr, dtype, copy, skipna)
    1199     if copy or is_object_dtype(arr.dtype) or
is_object_dtype(dtype):
    1200         # Explicit copy, or required since NumPy can't view
from / to object.
-> 1201         return arr.astype(dtype, copy=True)
    1202
    1203     return arr.astype(dtype, copy=copy)

```

ValueError: could not convert string to float: '?'

df.describe()

	symboling	wheel-base	length	width	height \
count	204.000000	204.000000	204.000000	204.000000	204.000000
mean	0.823529	98.806373	174.075000	65.916667	53.749020
std	1.239035	5.994144	12.362123	2.146716	2.424901
min	-2.000000	86.600000	141.100000	60.300000	47.800000
25%	0.000000	94.500000	166.300000	64.075000	52.000000

50%	1.000000	97.000000	173.200000	65.500000	54.100000
75%	2.000000	102.400000	183.200000	66.900000	55.500000
max	3.000000	120.900000	208.100000	72.300000	59.800000

	curb-weight	engine-size	compression-ratio	city-mpg
highway-mpg				
count	204.000000	204.000000	204.000000	204.000000
204.000000				
mean	2555.602941	126.892157	10.148137	25.240196
30.769608				
std	521.960820	41.744569	3.981000	6.551513
6.898337				
min	1488.000000	61.000000	7.000000	13.000000
16.000000				
25%	2145.000000	97.000000	8.575000	19.000000
25.000000				
50%	2414.000000	119.500000	9.000000	24.000000
30.000000				
75%	2939.250000	142.000000	9.400000	30.000000
34.500000				
max	4066.000000	326.000000	23.000000	49.000000
54.000000				

df.isnull().sum()

symboling	0
normalized-losses	0
make	0
fuel-type	0
aspiration	0
num-of-doors	0
body-style	0
drive-wheels	0
engine-location	0
wheel-base	0
length	0
width	0
height	0
curb-weight	0
engine-type	0
num-of-cylinders	0
engine-size	0
fuel-system	0
bore	0
stroke	0
compression-ratio	0
horsepower	0
peak-rpm	0
city-mpg	0
highway-mpg	0

```
price          0
dtype: int64
```

```
df['normalized-losses'].value_counts()
```

```
?      40
161    11
91      8
150     7
134     6
128     6
104     6
85      5
94      5
65      5
102     5
74      5
168     5
103     5
95      5
106     4
93      4
118     4
148     4
122     4
83      3
125     3
154     3
115     3
137     3
101     3
119     2
87      2
89      2
192     2
197     2
158     2
81      2
188     2
194     2
153     2
129     2
108     2
110     2
164     2
145     2
113     2
256     1
107     1
90      1
231     1
```

```

142      1
121      1
78       1
98       1
186      1
77       1

```

Name: normalized-losses, dtype: int64

```

import numpy as np
df.replace("?", np.nan, inplace = True)
df.head()

```

	symboling	normalized-losses	make	fuel-type	aspiration	num-
of-doors \						
0	3	NaN	alfa-romero	gas	std	
two						
1	1	NaN	alfa-romero	gas	std	
two						
2	2	164	audi	gas	std	
four						
3	2	164	audi	gas	std	
four						
4	2	NaN	audi	gas	std	
two						

	body-style	drive-wheels	engine-location	wheel-base	...	engine-
size \						
0	convertible	rwd	front	88.6	...	
130						
1	hatchback	rwd	front	94.5	...	
152						
2	sedan	fwd	front	99.8	...	
109						
3	sedan	4wd	front	99.4	...	
136						
4	sedan	fwd	front	99.8	...	
136						

	fuel-system	bore	stroke	compression-ratio	horsepower	peak-rpm
city-mpg \						
0	mpfi	3.47	2.68	9.0	111	5000
21						
1	mpfi	2.68	3.47	9.0	154	5000
19						
2	mpfi	3.19	3.40	10.0	102	5500
24						
3	mpfi	3.19	3.40	8.0	115	5500
18						
4	mpfi	3.19	3.40	8.5	110	5500
19						

	highway-mpg	price
0	27	16500
1	26	16500
2	30	13950
3	22	17450
4	25	15250

[5 rows x 26 columns]

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

```

symboling          0
normalized-losses  40
make              0
fuel-type         0
aspiration        0
num-of-doors      2
body-style        0
drive-wheels      0
engine-location   0
wheel-base       0
length           0
width            0
height           0
curb-weight       0
engine-type       0
num-of-cylinders  0
engine-size       0
fuel-system       0
bore             4
stroke           4
compression-ratio 0
horsepower        2
peak-rpm          2
city-mpg          0
highway-mpg       0
price            4
dtype: int64

```

```
df.dropna(subset=["normalized-losses"], axis=0)
```

	symboling	normalized-losses	make	fuel-type	aspiration	num-of-
doors \						
2	2	164	audi	gas	std	
four						
3	2	164	audi	gas	std	
four						
5	1	158	audi	gas	std	
four						
7	1	158	audi	gas	turbo	

four						
9	2	192	bmw	gas	std	
two						
..	
...						
199	-1	95	volvo	gas	std	
four						
200	-1	95	volvo	gas	turbo	
four						
201	-1	95	volvo	gas	std	
four						
202	-1	95	volvo	diesel	turbo	
four						
203	-1	95	volvo	gas	turbo	
four						

size \	body-style	drive-wheels	engine-location	wheel-base	...	engine-
2	sedan	fwd	front	99.8	...	
109						
3	sedan	4wd	front	99.4	...	
136						
5	sedan	fwd	front	105.8	...	
136						
7	sedan	fwd	front	105.8	...	
131						
9	sedan	rwd	front	101.2	...	
108						
..	
...						
199	sedan	rwd	front	109.1	...	
141						
200	sedan	rwd	front	109.1	...	
141						
201	sedan	rwd	front	109.1	...	
173						
202	sedan	rwd	front	109.1	...	
145						
203	sedan	rwd	front	109.1	...	
141						

\	fuel-system	bore	stroke	compression-ratio	horsepower	peak-rpm
2	mpfi	3.19	3.40	10.0	102	5500
3	mpfi	3.19	3.40	8.0	115	5500
5	mpfi	3.19	3.40	8.5	110	5500
7	mpfi	3.13	3.40	8.3	140	5500

9	mpfi	3.50	2.80	8.8	101	5800
..
199	mpfi	3.78	3.15	9.5	114	5400
200	mpfi	3.78	3.15	8.7	160	5300
201	mpfi	3.58	2.87	8.8	134	5500
202	idi	3.01	3.40	23.0	106	4800
203	mpfi	3.78	3.15	9.5	114	5400

	city-mpg	highway-mpg	price
2	24	30	13950
3	18	22	17450
5	19	25	17710
7	17	20	23875
9	23	29	16430
..
199	23	28	16845
200	19	25	19045
201	18	23	21485
202	26	27	22470
203	19	25	22625

[164 rows x 26 columns]

```
avg_norm_loss = df["normalized-losses"].astype("float").mean(axis=0)
print("Average of normalized-losses:", avg_norm_loss)
```

Average of normalized-losses: 122.0

```
df["normalized-losses"].replace(np.nan, avg_norm_loss, inplace=True)
```

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

symboling	0
normalized-losses	0
make	0
fuel-type	0
aspiration	0
num-of-doors	2
body-style	0
drive-wheels	0
engine-location	0
wheel-base	0

```

length          0
width           0
height          0
curb-weight     0
engine-type     0
num-of-cylinders 0
engine-size     0
fuel-system     0
bore            4
stroke          4
compression-ratio 0
horsepower      2
peak-rpm        2
city-mpg        0
highway-mpg     0
price           4
dtype: int64

```

```
df['length'].head()
```

```

0    168.8
1    171.2
2    176.6
3    176.6
4    177.3

```

```
Name: length, dtype: float64
```

```

df['length'] = df['length']/df['length'].max()
df['width'] = df['width']/df['width'].max()

```

```
df['length'].head()
```

```

0    0.811148
1    0.822681
2    0.848630
3    0.848630
4    0.851994

```

```
Name: length, dtype: float64
```

```
df.head()
```

	symboling	normalized-losses	make	...	city-mpg	highway-mpg
price						
0	3	122.0	alfa-romero	...	21	27
16500						
1	1	122.0	alfa-romero	...	19	26
16500						
2	2	164	audi	...	24	30
13950						
3	2	164	audi	...	18	22
17450						
4	2	122.0	audi	...	19	25

```
15250
```

```
[5 rows x 26 columns]
```

```
pd.get_dummies(df["fuel-type"])
```

	diesel	gas
0	0	1
1	0	1
2	0	1
3	0	1
4	0	1
...
199	0	1
200	0	1
201	0	1
202	1	0
203	0	1

```
[204 rows x 2 columns]
```

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

```
fwd    120
rwd     75
4wd      9
Name: drive-wheels, dtype: int64
```

```
df.corr()
```

	symboling	wheel-base	...	city-mpg	highway-mpg
symboling	1.000000	-0.525095	...	-0.030557	0.039598
wheel-base	-0.525095	1.000000	...	-0.479633	-0.552897
length	-0.356792	0.877612	...	-0.673251	-0.706635
width	-0.227799	0.795115	...	-0.647177	-0.681169
height	-0.533078	0.582603	...	-0.055659	-0.113995
curb-weight	-0.229281	0.781763	...	-0.758238	-0.798088
engine-size	-0.107229	0.573989	...	-0.654101	-0.677775
compression-ratio	-0.177413	0.249199	...	0.324186	0.264677
city-mpg	-0.030557	-0.479633	...	1.000000	0.971311
highway-mpg	0.039598	-0.552897	...	0.971311	1.000000

```
[10 rows x 10 columns]
```

```
df["price"] = df["price"].astype(float)
```

```
df.dtypes
```

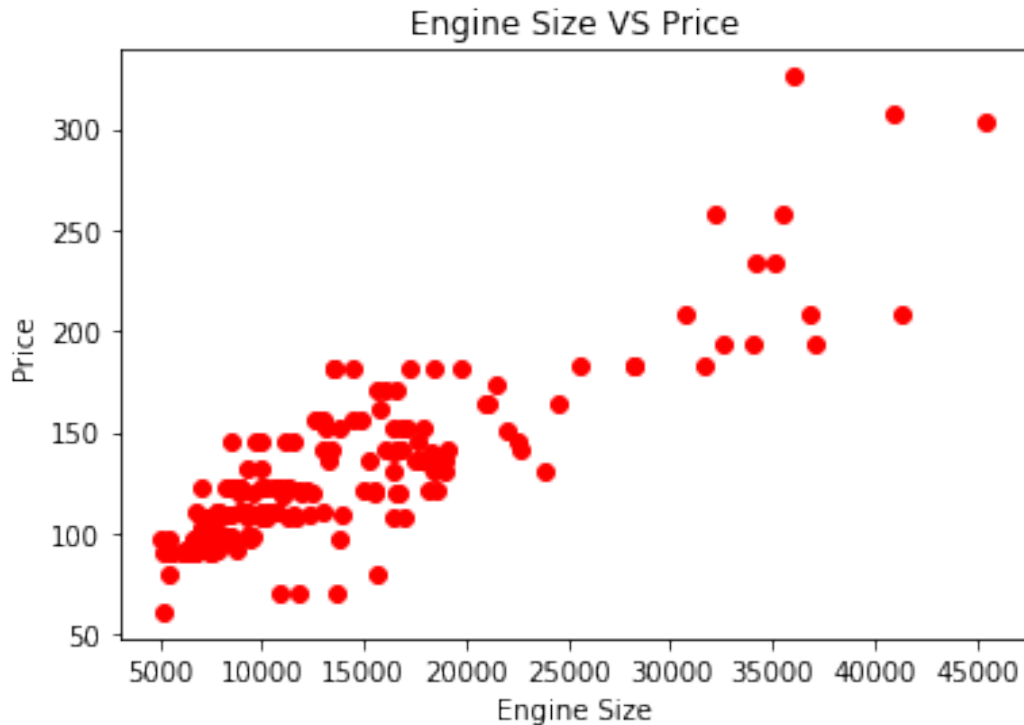
symboling	int64
normalized-losses	object
make	object
fuel-type	object

```
aspiration          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               object
stroke             object
compression-ratio   float64
horsepower          object
peak-rpm           object
city-mpg            int64
highway-mpg         int64
price              float64
dtype: object
```

```
import matplotlib.pyplot as plt
y = df["engine-size"]
x = df["price"]

plt.scatter(x,y, color = "Red")
plt.xlabel("Engine Size")
plt.ylabel("Price")
plt.title("Engine Size VS Price")

Text(0.5, 1.0, 'Engine Size VS Price')
```



```
import seaborn as sns
```

```
sns.set(style = "darkgrid")
```

```
sns.regplot(x,y,data=df, fit_reg = False, marker = "*", color = "blue")
```

```
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43:  
FutureWarning: Pass the following variables as keyword args: x, y.  
From version 0.12, the only valid positional argument will be `data`,  
and passing other arguments without an explicit keyword will result in  
an error or misinterpretation.
```

```
FutureWarning
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7ff728e76f90>
```

