

```
In [1]: import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
%matplotlib inline
import warnings
warnings.filterwarnings('ignore')
```

```
In [2]: df=pd.read_csv("cars.csv")
```

```
In [3]: df.head()
```

Out[3]:

	symboling	normalized- losses	make	fuel- type	body- style	drive- wheels	engine- location	width	height	engine- type	eng s
0	3	?	alfa-romero	gas	convertible	rwd	front	64.1	48.8	dohc	
1	3	?	alfa-romero	gas	convertible	rwd	front	64.1	48.8	dohc	
2	1	?	alfa-romero	gas	hatchback	rwd	front	65.5	52.4	ohcv	
3	2	164	audi	gas	sedan	fwd	front	66.2	54.3	ohc	
4	2	164	audi	gas	sedan	4wd	front	66.4	54.3	ohc	

```
In [4]: df.shape
```

Out[4]: (205, 15)

```
In [5]: df.describe()
```

Out[5]:

	symboling	width	height	engine-size	city-mpg	highway-mpg	price
count	205.000000	205.000000	205.000000	205.000000	205.000000	205.000000	205.000000
mean	0.834146	65.907805	53.724878	126.907317	25.219512	30.751220	13227.478049
std	1.245307	2.145204	2.443522	41.642693	6.542142	6.886443	7902.651615
min	-2.000000	60.300000	47.800000	61.000000	13.000000	16.000000	5118.000000
25%	0.000000	64.100000	52.000000	97.000000	19.000000	25.000000	7788.000000
50%	1.000000	65.500000	54.100000	120.000000	24.000000	30.000000	10345.000000
75%	2.000000	66.900000	55.500000	141.000000	30.000000	34.000000	16500.000000
max	3.000000	72.300000	59.800000	326.000000	49.000000	54.000000	45400.000000

In [6]: df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 205 entries, 0 to 204
Data columns (total 15 columns):
#   Column                Non-Null Count  Dtype
---  -
0   symboling              205 non-null    int64
1   normalized-losses      205 non-null    object
2   make                   205 non-null    object
3   fuel-type              205 non-null    object
4   body-style             205 non-null    object
5   drive-wheels           205 non-null    object
6   engine-location        205 non-null    object
7   width                  205 non-null    float64
8   height                 205 non-null    float64
9   engine-type            205 non-null    object
10  engine-size            205 non-null    int64
11  horsepower             205 non-null    object
12  city-mpg               205 non-null    int64
13  highway-mpg            205 non-null    int64
14  price                  205 non-null    int64
dtypes: float64(2), int64(5), object(8)
memory usage: 24.1+ KB
```

In [7]: num_col=df.select_dtypes(include=['int','float']).columns

In [8]: num_col

Out[8]: Index(['symboling', 'width', 'height', 'engine-size', 'city-mpg',
'highway-mpg', 'price'],
dtype='object')

In [9]: len(num_col)

Out[9]: 7

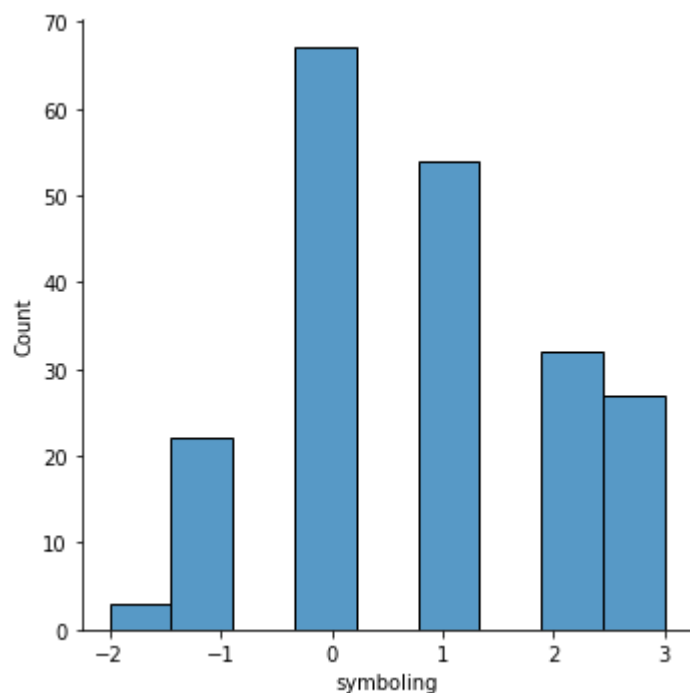
In [10]: df.describe()

Out[10]:

	symboling	width	height	engine-size	city-mpg	highway-mpg	price
count	205.000000	205.000000	205.000000	205.000000	205.000000	205.000000	205.000000
mean	0.834146	65.907805	53.724878	126.907317	25.219512	30.751220	13227.478049
std	1.245307	2.145204	2.443522	41.642693	6.542142	6.886443	7902.651615
min	-2.000000	60.300000	47.800000	61.000000	13.000000	16.000000	5118.000000
25%	0.000000	64.100000	52.000000	97.000000	19.000000	25.000000	7788.000000
50%	1.000000	65.500000	54.100000	120.000000	24.000000	30.000000	10345.000000
75%	2.000000	66.900000	55.500000	141.000000	30.000000	34.000000	16500.000000
max	3.000000	72.300000	59.800000	326.000000	49.000000	54.000000	45400.000000

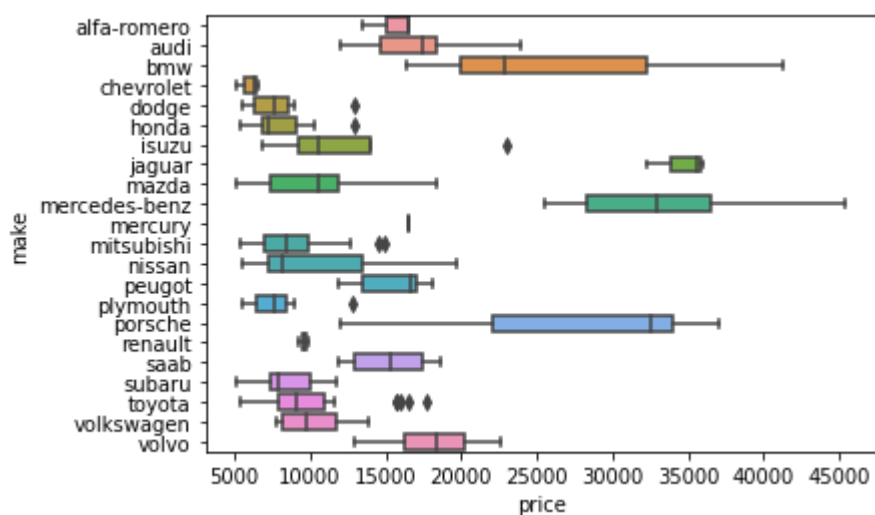
```
In [11]: sns.displot(df['symboling'])
```

```
Out[11]: <seaborn.axisgrid.FacetGrid at 0x19bf7a4b610>
```



```
In [12]: sns.boxplot(x='price',y='make',data=df)
```

```
Out[12]: <AxesSubplot:xlabel='price', ylabel='make'>
```



```
In [13]: df.isnull().sum()
```

```
Out[13]: symboling          0
normalized-losses         0
make                      0
fuel-type                 0
body-style                0
drive-wheels              0
engine-location           0
width                    0
height                   0
engine-type               0
engine-size               0
horsepower                0
city-mpg                  0
highway-mpg               0
price                     0
dtype: int64
```

```
In [14]: df.make.unique()
```

```
Out[14]: array(['alfa-romero', 'audi', 'bmw', 'chevrolet', 'dodge', 'honda',
               'isuzu', 'jaguar', 'mazda', 'mercedes-benz', 'mercury',
               'mitsubishi', 'nissan', 'peugot', 'plymouth', 'porsche', 'renault',
               'saab', 'subaru', 'toyota', 'volkswagen', 'volvo'], dtype=object)
```

```
In [15]: df.make.value_counts()
```

```
Out[15]: toyota          32
nissan                 18
mazda                 17
mitsubishi            13
honda                 13
volkswagen            12
subaru                12
peugot                11
volvo                 11
dodge                 9
mercedes-benz         8
bmw                   8
audi                  7
plymouth              7
saab                  6
porsche               5
isuzu                 4
jaguar                3
chevrolet             3
alfa-romero           3
renault               2
mercury               1
Name: make, dtype: int64
```

```
In [16]: df[df['make']=="toyota"]
```

```
Out[16]:
```

	symboling	normalized-losses	make	fuel-type	body-style	drive-wheels	engine-location	width	height	engine-type	ei
150	1	87	toyota	gas	hatchback	fwd	front	63.6	54.5	ohc	
151	1	87	toyota	gas	hatchback	fwd	front	63.6	54.5	ohc	
152	1	74	toyota	gas	hatchback	fwd	front	63.6	54.5	ohc	
153	0	77	toyota	gas	wagon	fwd	front	63.6	59.1	ohc	
154	0	81	toyota	gas	wagon	4wd	front	63.6	59.1	ohc	
155	0	91	toyota	gas	wagon	4wd	front	63.6	59.1	ohc	
156	0	91	toyota	gas	sedan	fwd	front	64.4	53.0	ohc	
157	0	91	toyota	gas	hatchback	fwd	front	64.4	52.8	ohc	
158	0	91	toyota	diesel	sedan	fwd	front	64.4	53.0	ohc	
159	0	91	toyota	diesel	hatchback	fwd	front	64.4	52.8	ohc	
160	0	91	toyota	gas	sedan	fwd	front	64.4	53.0	ohc	
161	0	91	toyota	gas	hatchback	fwd	front	64.4	52.8	ohc	
162	0	91	toyota	gas	sedan	fwd	front	64.4	52.8	ohc	
163	1	168	toyota	gas	sedan	rwd	front	64.0	52.6	ohc	
164	1	168	toyota	gas	hatchback	rwd	front	64.0	52.6	ohc	
165	1	168	toyota	gas	sedan	rwd	front	64.0	52.6	dohc	
166	1	168	toyota	gas	hatchback	rwd	front	64.0	52.6	dohc	
167	2	134	toyota	gas	hardtop	rwd	front	65.6	52.0	ohc	
168	2	134	toyota	gas	hardtop	rwd	front	65.6	52.0	ohc	
169	2	134	toyota	gas	hatchback	rwd	front	65.6	52.0	ohc	
170	2	134	toyota	gas	hardtop	rwd	front	65.6	52.0	ohc	
171	2	134	toyota	gas	hatchback	rwd	front	65.6	52.0	ohc	
172	2	134	toyota	gas	convertible	rwd	front	65.6	53.0	ohc	
173	-1	65	toyota	gas	sedan	fwd	front	66.5	54.9	ohc	
174	-1	65	toyota	diesel	sedan	fwd	front	66.5	54.9	ohc	
175	-1	65	toyota	gas	hatchback	fwd	front	66.5	53.9	ohc	
176	-1	65	toyota	gas	sedan	fwd	front	66.5	54.9	ohc	
177	-1	65	toyota	gas	hatchback	fwd	front	66.5	53.9	ohc	
178	3	197	toyota	gas	hatchback	rwd	front	67.7	52.0	dohc	
179	3	197	toyota	gas	hatchback	rwd	front	67.7	52.0	dohc	
180	-1	90	toyota	gas	sedan	rwd	front	66.5	54.1	dohc	
181	-1	?	toyota	gas	wagon	rwd	front	66.5	54.1	dohc	

```
In [17]: df[(df['make']=="toyota") & (df['price']>15000)]
```

```
Out[17]:
```

	symboling	normalized-losses	make	fuel-type	body-style	drive-wheels	engine-location	width	height	engine-type	en
172	2	134	toyota	gas	convertible	rwd	front	65.6	53.0	ohc	
178	3	197	toyota	gas	hatchback	rwd	front	67.7	52.0	dohc	
179	3	197	toyota	gas	hatchback	rwd	front	67.7	52.0	dohc	
180	-1	90	toyota	gas	sedan	rwd	front	66.5	54.1	dohc	
181	-1	?	toyota	gas	wagon	rwd	front	66.5	54.1	dohc	

```
In [25]: df.drop(index=[172,178,179,180,181],inplace=True)
```

```
In [27]: df.isnull().sum()
```

```
Out[27]: symboling      0
normalized-losses    0
make                 0
fuel-type            0
body-style           0
drive-wheels         0
engine-location       0
width                0
height               0
engine-type           0
engine-size           0
horsepower            0
city-mpg              0
highway-mpg           0
price                0
dtype: int64
```

```
In [ ]: maean=df['normalized-losses'].mean()
mean
```

```
In [19]: df['make'].isnull().sum()
```

```
Out[19]: 0
```

In [20]: df

Out[20]:

	symboling	normalized-losses	make	fuel-type	body-style	drive-wheels	engine-location	width	height	engine-type
0	3	?	alfa-romero	gas	convertible	rwd	front	64.1	48.8	dohc
1	3	?	alfa-romero	gas	convertible	rwd	front	64.1	48.8	dohc
2	1	?	alfa-romero	gas	hatchback	rwd	front	65.5	52.4	ohcv
3	2	164	audi	gas	sedan	fwd	front	66.2	54.3	ohc
4	2	164	audi	gas	sedan	4wd	front	66.4	54.3	ohc
...
200	-1	95	volvo	gas	sedan	rwd	front	68.9	55.5	ohc
201	-1	95	volvo	gas	sedan	rwd	front	68.8	55.5	ohc
202	-1	95	volvo	gas	sedan	rwd	front	68.9	55.5	ohcv
203	-1	95	volvo	diesel	sedan	rwd	front	68.9	55.5	ohc
204	-1	95	volvo	gas	sedan	rwd	front	68.9	55.5	ohc

205 rows × 15 columns



In [22]: df['normalized-losses']=df['normalized-losses'].replace('?',np.nan)

In [25]: df.dropna(inplace=True)

In [26]: df['normalized-losses'].isnull().sum()

Out[26]: 0

In [27]: cat_col=df.select_dtypes('object').columns

In [28]: cat_col

Out[28]: Index(['normalized-losses', 'make', 'fuel-type', 'body-style', 'drive-wheels', 'engine-location', 'engine-type', 'horsepower'], dtype='object')

In [29]: df['fuel-type'].unique()

Out[29]: array(['gas', 'diesel'], dtype=object)

```
In [30]: df['horsepower'].unique()
```

```
Out[30]: array(['102', '115', '110', '140', '101', '121', '48', '70', '68', '88',  
              '145', '58', '76', '60', '86', '100', '176', '135', '84', '120',  
              '123', '155', '116', '69', '55', '97', '152', '160', '200', '95',  
              '142', '143', '73', '82', '94', '111', '62', '56', '112', '92',  
              '161', '156', '52', '85', '90', '114', '162', '134', '106'],  
             dtype=object)
```

```
In [31]: df['horsepower'].isnull().sum()
```

```
Out[31]: 0
```

```
In [34]: df['horsepower']=df['horsepower'].replace("?",np.nan)
```

```
In [35]: df['horsepower'].isnull().sum()
```

```
Out[35]: 0
```

```
In [36]: pip install sklearn
```

Requirement already satisfied: sklearn in c:\users\mainawati\appdata\local\programs\python\python310\lib\site-packages (0.0)

Requirement already satisfied: scikit-learn in c:\users\mainawati\appdata\local\programs\python\python310\lib\site-packages (from sklearn) (1.1.1)

Requirement already satisfied: threadpoolctl>=2.0.0 in c:\users\mainawati\appdata\local\programs\python\python310\lib\site-packages (from scikit-learn->sklearn) (3.1.0)

Requirement already satisfied: joblib>=1.0.0 in c:\users\mainawati\appdata\local\programs\python\python310\lib\site-packages (from scikit-learn->sklearn) (1.1.0)

Requirement already satisfied: numpy>=1.17.3 in c:\users\mainawati\appdata\local\programs\python\python310\lib\site-packages (from scikit-learn->sklearn) (1.23.1)

Requirement already satisfied: scipy>=1.3.2 in c:\users\mainawati\appdata\local\programs\python\python310\lib\site-packages (from scikit-learn->sklearn) (1.8.1)

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

WARNING: You are using pip version 22.0.4; however, version 22.2.2 is available.

You should consider upgrading via the 'C:\Users\mainawati\AppData\Local\Programs\Python\Python310\python.exe -m pip install --upgrade pip' command.

```
In [37]: df.columns
```

```
Out[37]: Index(['symboling', 'normalized-losses', 'make', 'fuel-type', 'body-style',  
              'drive-wheels', 'engine-location', 'width', 'height', 'engine-type',  
              'engine-size', 'horsepower', 'city-mpg', 'highway-mpg', 'price'],  
             dtype='object')
```



```
In [38]: from sklearn.preprocessing import LabelEncoder
```

```
In [39]: s=LabelEncoder()
```

```
In [42]: df['fuel-type']=s.fit_transform(df['fuel-type'])
```

```
In [43]: df['fuel-type']
```

```
Out[43]: 3      1
         4      1
         6      1
         8      1
        10      1
         ..
       200      1
       201      1
       202      1
       203      0
       204      1
Name: fuel-type, Length: 164, dtype: int32
```

```
In [44]: df['fuel-type'].unique()
```

```
Out[44]: array([1, 0])
```

```
In [45]: df.head()
```

```
Out[45]:
```

	symboling	normalized- losses	make	fuel- type	body- style	drive- wheels	engine- location	width	height	engine- type	engine- size
3	2	164	audi	1	sedan	fwd	front	66.2	54.3	ohc	109
4	2	164	audi	1	sedan	4wd	front	66.4	54.3	ohc	136
6	1	158	audi	1	sedan	fwd	front	71.4	55.7	ohc	136
8	1	158	audi	1	sedan	fwd	front	71.4	55.9	ohc	131
10	2	192	bmw	1	sedan	rwd	front	64.8	54.3	ohc	108

```
In [46]: cat_col
```

```
Out[46]: Index(['normalized-losses', 'make', 'fuel-type', 'body-style', 'drive-wheels',
               'engine-location', 'engine-type', 'horsepower'],
              dtype='object')
```

```
In [47]: num_col
```

```
Out[47]: Index(['symboling', 'width', 'height', 'engine-size', 'city-mpg',
               'highway-mpg', 'price'],
              dtype='object')
```

```
In [54]: #s=LabelEncoder()
for col in cat_col:
    s=LabelEncoder()
    df[col]=s.fit_transform(df[col])
```

```
In [55]: df.head()
```

```
Out[55]:
```

	symboling	normalized- losses	make	fuel- type	body- style	drive- wheels	engine- location	width	height	engine- type	engine- size
3	2	27	0	1	3	1	0	66.2	54.3	2	109
4	2	27	0	1	3	0	0	66.4	54.3	2	136
6	1	25	0	1	3	1	0	71.4	55.7	2	136
8	1	25	0	1	3	1	0	71.4	55.9	2	131
10	2	31	1	1	3	2	0	64.8	54.3	2	108

```
In [56]: num_col
```

```
Out[56]: Index(['symboling', 'width', 'height', 'engine-size', 'city-mpg',
                'highway-mpg', 'price'],
                dtype='object')
```

```
In [58]: for col in num_col:
print(col)
plt.figure(figsize=(10,10))
sns.distplot(df[col])
plt.show()
```

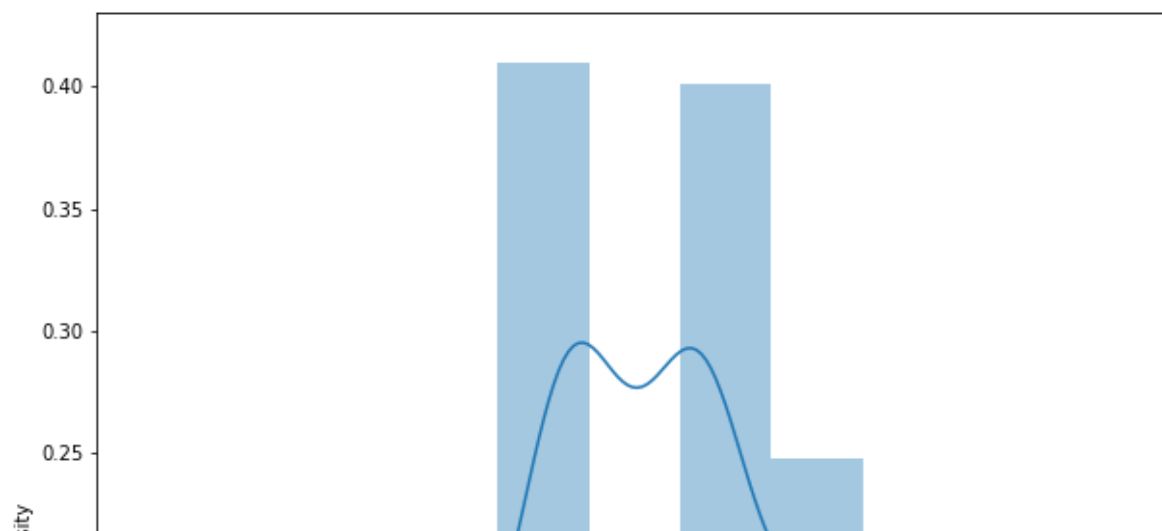
symboling

```
In [63]: for col in num_col:
          print(col)
          print('skewness',df[col].skew())
          print('kurtosis',df[col].kurt())
          plt.figure(figsize=(10,10))
          sns.distplot(df[col])
          plt.show()
```

symboling

skewness 0.10049657227421199

kurtosis -0.5947812272761666



```
In [64]: corr=df.corr()
```

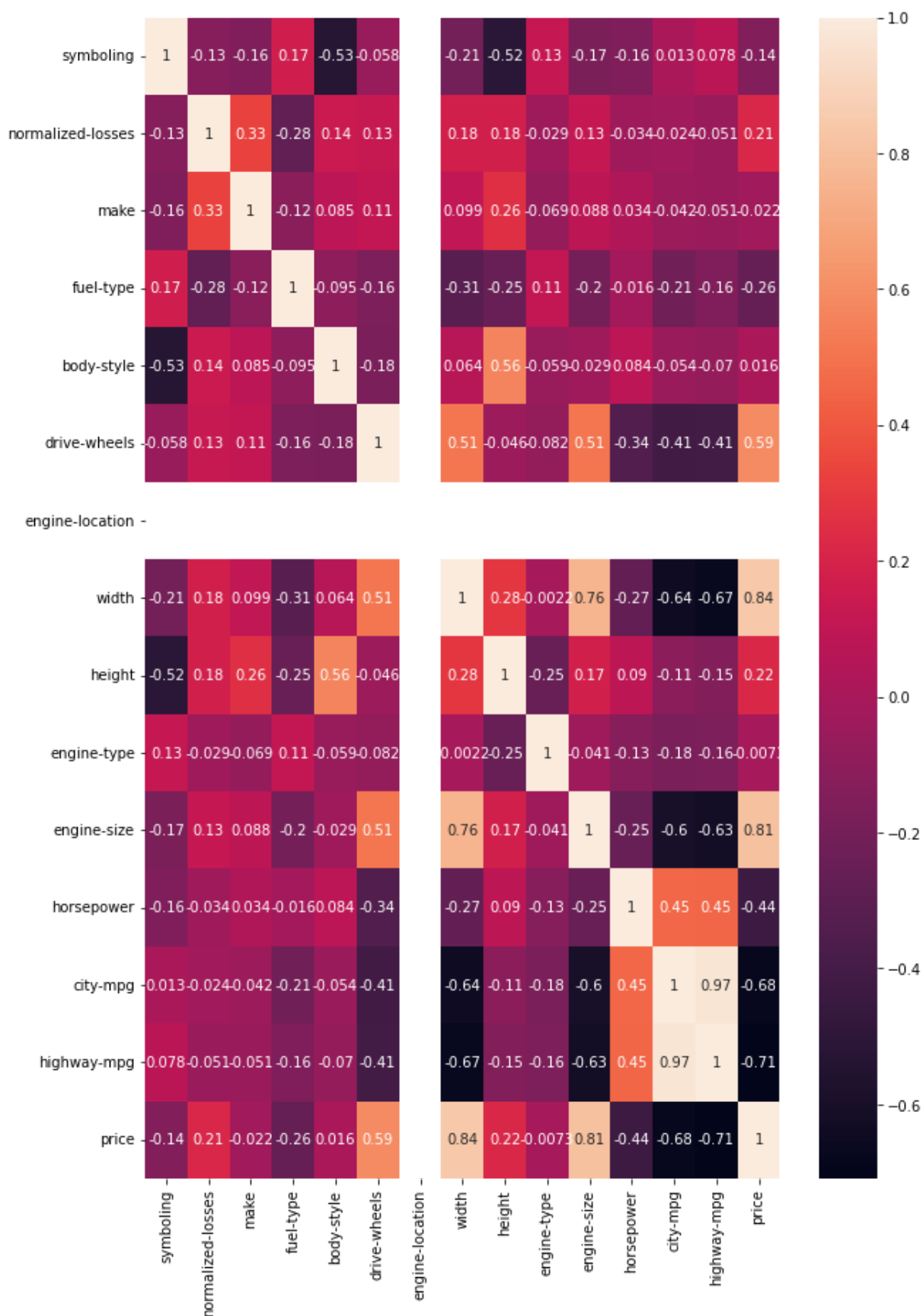
In [65]: corr

Out[65]:

	symboling	normalized-losses	make	fuel-type	body-style	drive-wheels	engine-location	width
symboling	1.000000	-0.132364	-0.155640	0.171179	-0.529376	-0.058496	NaN	-0.208303
normalized-losses	-0.132364	1.000000	0.328547	-0.277971	0.142251	0.128753	NaN	0.177806
make	-0.155640	0.328547	1.000000	-0.116861	0.084697	0.105792	NaN	0.098805
fuel-type	0.171179	-0.277971	-0.116861	1.000000	-0.095006	-0.163889	NaN	-0.312443
body-style	-0.529376	0.142251	0.084697	-0.095006	1.000000	-0.178639	NaN	0.063630
drive-wheels	-0.058496	0.128753	0.105792	-0.163889	-0.178639	1.000000	NaN	0.512294
engine-location	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
width	-0.208303	0.177806	0.098805	-0.312443	0.063630	0.512294	NaN	1.000000
height	-0.516420	0.178331	0.256464	-0.250596	0.562439	-0.046305	NaN	0.283885
engine-type	0.125153	-0.029012	-0.068594	0.111097	-0.059247	-0.081515	NaN	-0.002173
engine-size	-0.169342	0.125227	0.087860	-0.197530	-0.028671	0.506697	NaN	0.756317
horsepower	-0.155215	-0.034188	0.034086	-0.016486	0.084412	-0.342580	NaN	-0.274705
city-mpg	0.013028	-0.023534	-0.042058	-0.212318	-0.053769	-0.410807	NaN	-0.644124
highway-mpg	0.078220	-0.051395	-0.050584	-0.157270	-0.069524	-0.408630	NaN	-0.674959
price	-0.144078	0.214196	-0.022170	-0.259010	0.016015	0.591703	NaN	0.841883



```
In [68]: plt.figure(figsize=(10,15))
sns.heatmap(corr,annot=True)
plt.show()
```



In []: