fgmbh2c29

March 30, 2023

Data cleaning on Automobile 1985 dataset and perform descriptive analytics

```
[3]: import numpy as np
     import pandas as pd
    data = pd.read_csv('auto_clean.csv')
    data.head(2)
[7]:
        symboling normalized-losses
                                             make aspiration num-of-doors
     0
                3
                                 122
                                      alfa-romero
                                                          std
                                                                       t.WO
                3
     1
                                 122
                                      alfa-romero
                                                          std
                                                                       two
         body-style drive-wheels engine-location wheel-base
                                                                 length ...
     0 convertible
                             rwd
                                           front
                                                         88.6 0.811148
     1 convertible
                                           front
                                                         88.6 0.811148 ...
                             rwd
        compression-ratio horsepower peak-rpm city-mpg highway-mpg price
     0
                      9.0
                                111.0
                                         5000.0
                                                       21
                                                                   27 13495
     1
                      9.0
                                111.0
                                         5000.0
                                                       21
                                                                   27 16500
       city-L/100km horsepower-binned
     0
          11.190476
                                Medium
     1
          11.190476
                                Medium
                                             0
                                                   1
     [2 rows x 29 columns]
[8]: data.columns
[8]: Index(['symboling', 'normalized-losses', 'make', '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',
            'city-L/100km', 'horsepower-binned', 'diesel', 'gas'],
           dtype='object')
[9]: data.describe()
```

```
[9]:
               symboling
                          normalized-losses
                                               wheel-base
                                                                 length
                                                                               width
      count
              201.000000
                                   201.00000
                                               201.000000
                                                            201.000000
                                                                         201.000000
                0.840796
                                   122.00000
                                                98.797015
                                                              0.837102
                                                                           0.915126
      mean
      std
                1.254802
                                    31.99625
                                                 6.066366
                                                              0.059213
                                                                           0.029187
                                                86.600000
      min
               -2.000000
                                    65.00000
                                                              0.678039
                                                                           0.837500
      25%
                                   101.00000
                0.000000
                                                94.500000
                                                              0.801538
                                                                           0.890278
      50%
                1.000000
                                   122.00000
                                                97.000000
                                                               0.832292
                                                                           0.909722
      75%
                2.000000
                                   137.00000
                                               102.400000
                                                              0.881788
                                                                           0.925000
                3.000000
                                   256.00000
                                               120.900000
                                                               1.000000
                                                                           1.000000
      max
                           curb-weight
                                                                                 \
                  height
                                         engine-size
                                                             bore
                                                                        stroke
              201.000000
                            201.000000
                                          201.000000
      count
                                                       201.000000
                                                                    197.000000
               53.766667
                           2555.666667
                                                         3.330692
                                                                      3.256904
      mean
                                          126.875622
      std
                2.447822
                            517.296727
                                           41.546834
                                                         0.268072
                                                                      0.319256
      min
               47.800000
                           1488.000000
                                           61.000000
                                                         2.540000
                                                                      2.070000
      25%
               52.000000
                           2169.000000
                                           98.000000
                                                         3.150000
                                                                      3.110000
      50%
               54.100000
                           2414.000000
                                          120.000000
                                                         3.310000
                                                                      3.290000
      75%
               55.500000
                           2926.000000
                                          141.000000
                                                         3.580000
                                                                      3.410000
               59.800000
                           4066.000000
                                          326.000000
                                                         3.940000
                                                                      4.170000
      max
              compression-ratio
                                  horsepower
                                                               city-mpg
                                                                          highway-mpg
                                                  peak-rpm
                     201.000000
                                  201.000000
                                                201.000000
                                                             201.000000
                                                                           201.000000
      count
      mean
                      10.164279
                                  103.405534
                                               5117.665368
                                                               25.179104
                                                                            30.686567
      std
                       4.004965
                                   37.365700
                                                478.113805
                                                                6.423220
                                                                             6.815150
      min
                       7.000000
                                   48.000000
                                               4150.000000
                                                               13.000000
                                                                             16.000000
      25%
                       8.600000
                                   70.000000
                                               4800.000000
                                                               19.000000
                                                                            25.000000
      50%
                       9.000000
                                   95.000000
                                               5125.369458
                                                               24.000000
                                                                            30.000000
      75%
                       9.400000
                                  116.000000
                                               5500.000000
                                                               30.000000
                                                                            34.000000
                      23.000000
                                  262.000000
                                                               49.000000
                                                                            54.000000
                                               6600.000000
      max
                             city-L/100km
                     price
                                                diesel
                                                                 gas
                201.000000
                               201.000000
                                            201.000000
                                                         201.000000
      count
              13207.129353
                                 9.944145
                                              0.099502
                                                           0.900498
      mean
      std
               7947.066342
                                 2.534599
                                              0.300083
                                                           0.300083
                                 4.795918
      min
               5118.000000
                                              0.000000
                                                           0.000000
      25%
               7775.000000
                                 7.833333
                                              0.000000
                                                           1.000000
      50%
              10295.000000
                                 9.791667
                                              0.000000
                                                           1.000000
      75%
              16500.000000
                                12.368421
                                              0.000000
                                                           1.000000
                                18.076923
      max
              45400.000000
                                              1.000000
                                                           1.000000
[10]:
     data.info()
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 201 entries, 0 to 200
Data columns (total 29 columns):

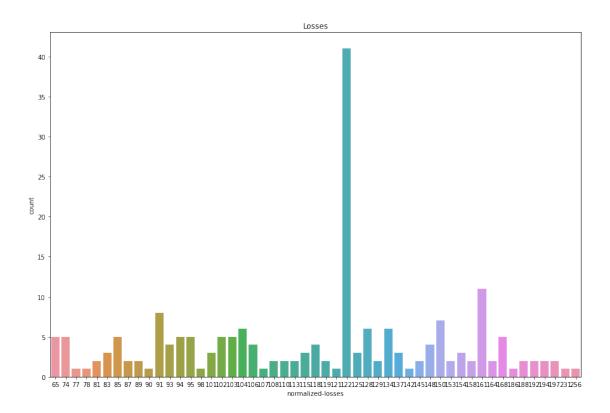
Column Non-Null Count Dtype

```
0
          symboling
                              201 non-null
                                              int64
          normalized-losses
                                              int64
      1
                              201 non-null
      2
          make
                              201 non-null
                                              object
      3
          aspiration
                              201 non-null
                                              object
      4
          num-of-doors
                              201 non-null
                                              object
      5
          body-style
                              201 non-null
                                              object
      6
          drive-wheels
                              201 non-null
                                              object
      7
          engine-location
                              201 non-null
                                              object
          wheel-base
                              201 non-null
                                              float64
      9
                                              float64
          length
                              201 non-null
      10 width
                                              float64
                              201 non-null
         height
                              201 non-null
                                              float64
      11
      12
          curb-weight
                              201 non-null
                                              int64
          engine-type
      13
                              201 non-null
                                              object
      14
          num-of-cylinders
                              201 non-null
                                              object
          engine-size
                              201 non-null
                                              int64
      16
          fuel-system
                              201 non-null
                                              object
      17
         bore
                              201 non-null
                                              float64
      18
          stroke
                              197 non-null
                                              float64
      19
          compression-ratio
                              201 non-null
                                              float64
      20
          horsepower
                              201 non-null
                                              float64
      21
                              201 non-null
                                              float64
          peak-rpm
          city-mpg
                              201 non-null
                                              int64
      23
         highway-mpg
                              201 non-null
                                              int64
      24 price
                              201 non-null
                                              int64
          city-L/100km
      25
                              201 non-null
                                              float64
      26
         horsepower-binned 200 non-null
                                              object
      27
                                              int64
          diesel
                              201 non-null
                              201 non-null
                                              int64
      28
     dtypes: float64(10), int64(9), object(10)
     memory usage: 45.7+ KB
[11]: # Loss
      import matplotlib.pyplot as plt
      import seaborn as sns
      plt.figure(figsize=(15,10))
      sns.countplot(data['normalized-losses'])
```

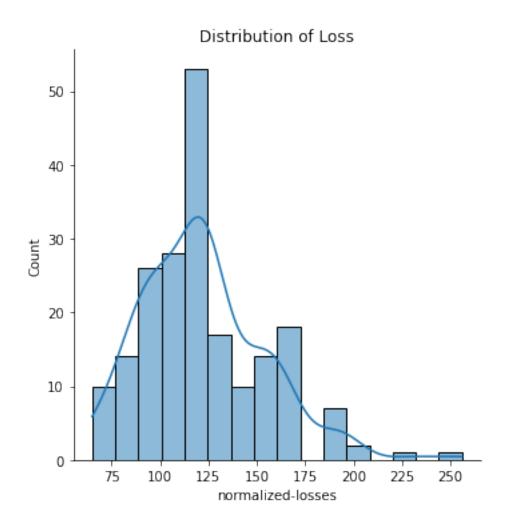
warnings.warn(

plt.title('Losses')

plt.show()



[13]: data['normalized-losses'].describe() [13]: count 201.00000 mean 122.00000 std 31.99625 min 65.00000 25% 101.00000 50% 122.00000 75% 137.00000 256.00000 max Name: normalized-losses, dtype: float64 [14]: sns.displot(data['normalized-losses'],kde=True) plt.title('Distribution of Loss') plt.show()

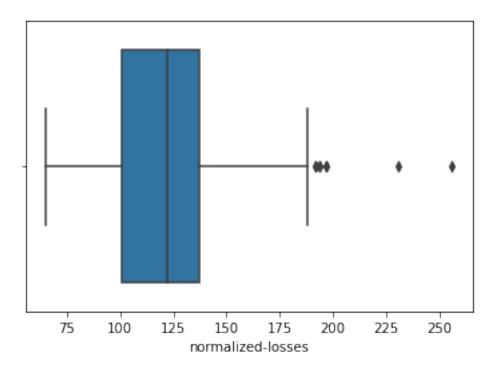


[15]: sns.boxplot(data['normalized-losses'])

C:\Users\87548\anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variable as a keyword arg: x. 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.

warnings.warn(

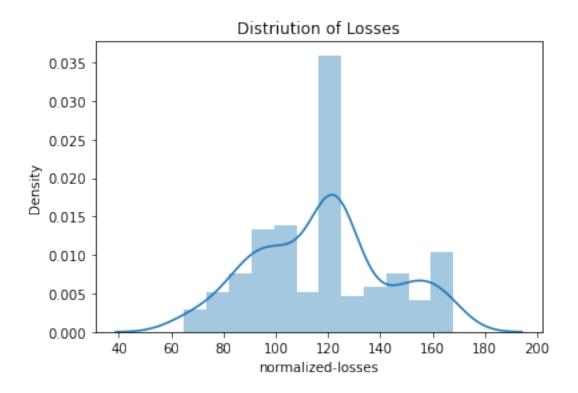
[15]: <AxesSubplot:xlabel='normalized-losses'>



```
[18]: sns.distplot(data['normalized-losses'],kde=True)
   plt.title('Distriution of Losses')
   plt.show()
```

C:\Users\87548\anaconda3\lib\site-packages\seaborn\distributions.py:2557:
FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)



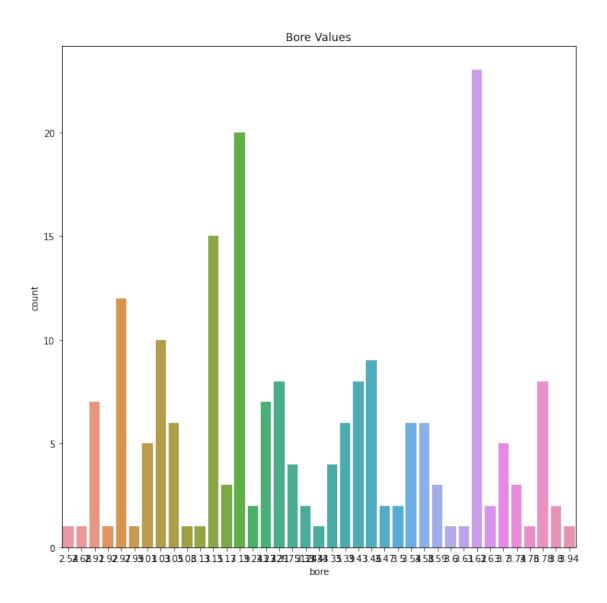
[19]: data.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 201 entries, 0 to 200
Data columns (total 29 columns):

#	Column	Non-Null Count	Dtype
0	symboling	201 non-null	int64
1	normalized-losses	201 non-null	float64
2	make	201 non-null	object
3	aspiration	201 non-null	object
4	num-of-doors	201 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
                       201 non-null
                                       float64
 18 stroke
                       197 non-null
                                       float64
    compression-ratio 201 non-null
                                       float64
 19
 20 horsepower
                       201 non-null
                                       float64
 21
    peak-rpm
                       201 non-null
                                       float64
22
    city-mpg
                       201 non-null
                                       int64
                                       int64
23 highway-mpg
                       201 non-null
 24 price
                       201 non-null
                                       int64
25 city-L/100km
                       201 non-null
                                       float64
 26 horsepower-binned 200 non-null
                                       object
 27 diesel
                       201 non-null
                                       int64
                       201 non-null
 28 gas
                                       int64
dtypes: float64(11), int64(8), object(10)
memory usage: 45.7+ KB
```

```
[20]: # Bore
      plt.figure(figsize=(10,10))
      sns.countplot(data['bore'])
      plt.title('Bore Values')
      plt.show()
```

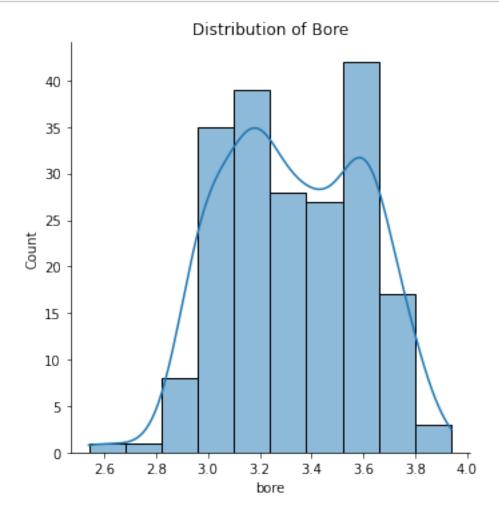


[21]: data['bore'].describe() [21]: count 201.000000 mean 3.330692

mean 3.330692 std 0.268072 min 2.540000 25% 3.150000 50% 3.310000 75% 3.580000 max 3.940000

Name: bore, dtype: float64

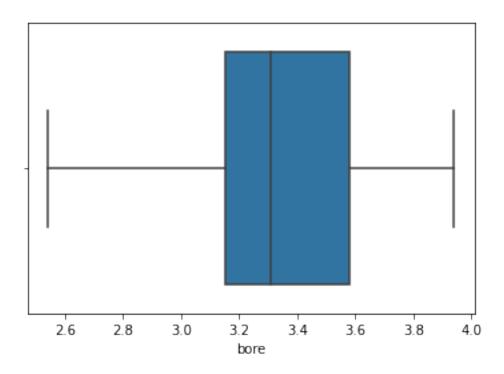
```
[22]: sns.displot(data['bore'],kde=True)
   plt.title('Distribution of Bore')
   plt.show()
```



```
[23]: # Handling outliers
sns.boxplot(data['bore'])
```

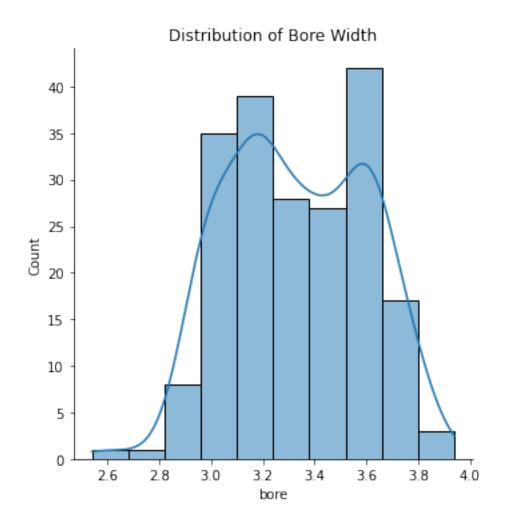
warnings.warn(

[23]: <AxesSubplot:xlabel='bore'>

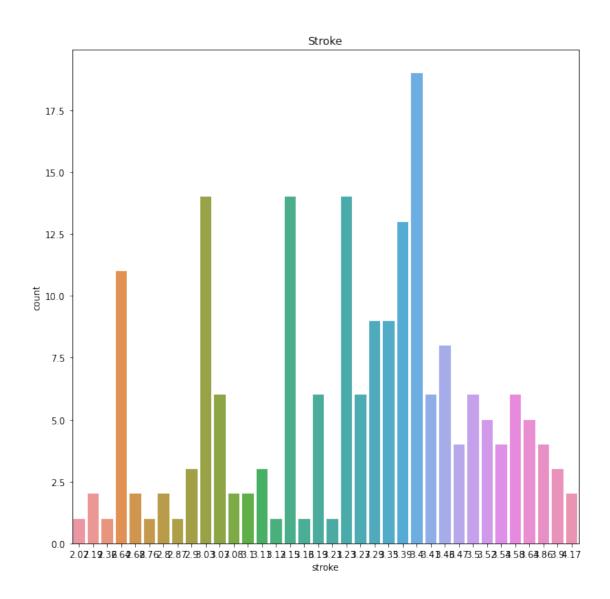


```
[24]: data['bore']=data['bore'].apply(lambda x: data['bore'].mean() if (x==0) else x)

[25]: sns.displot(data['bore'],kde=True)
    plt.title('Distribution of Bore Width')
    plt.show()
```



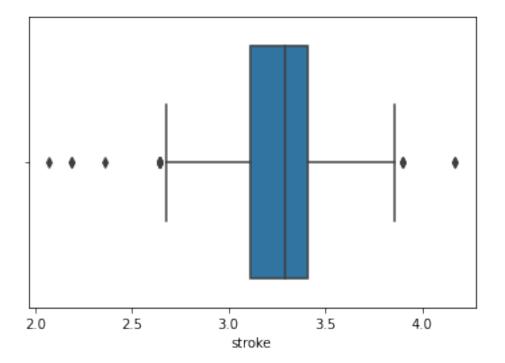
```
[26]: # Stroke
   plt.figure(figsize=(10,10))
   sns.countplot(data['stroke'])
   plt.title('Stroke')
   plt.show()
```



```
[27]: data['stroke'].describe()
[27]: count
               197.000000
                 3.256904
      mean
      std
                 0.319256
      min
                 2.070000
      25%
                 3.110000
      50%
                 3.290000
      75%
                 3.410000
                 4.170000
      max
      Name: stroke, dtype: float64
[28]: # Handling outliers
      sns.boxplot(data['stroke'])
```

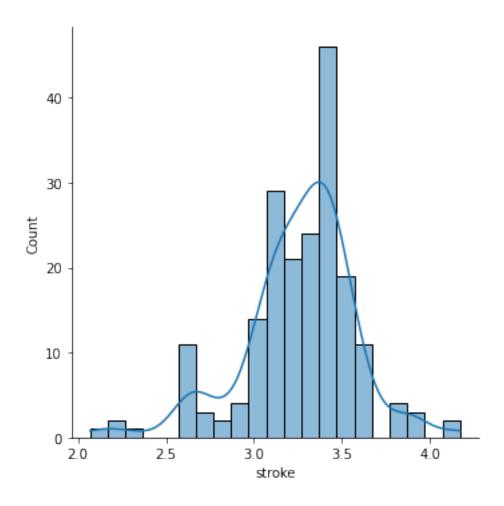
warnings.warn(

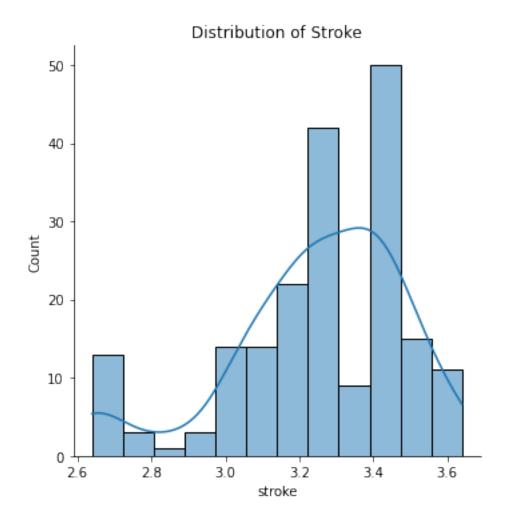
[28]: <AxesSubplot:xlabel='stroke'>



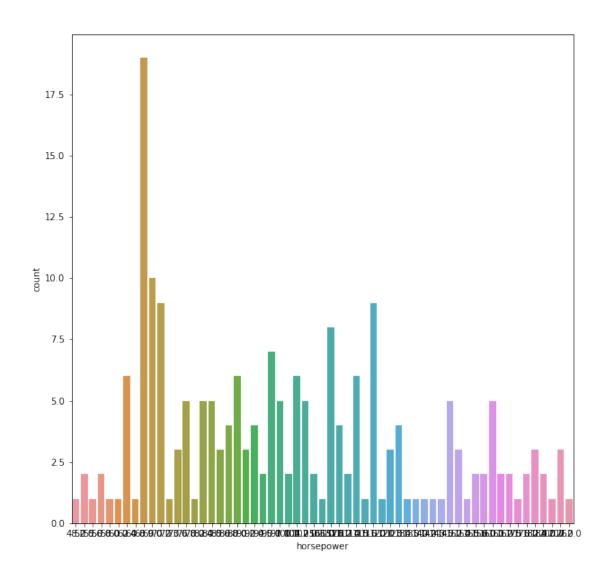
[29]: sns.displot(data['stroke'],kde=True)

[29]: <seaborn.axisgrid.FacetGrid at 0x2500f407a60>





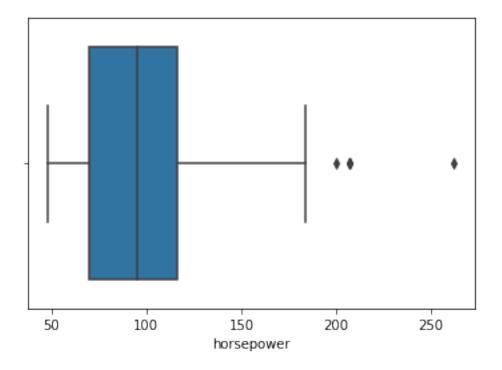
```
[32]: # hp
plt.figure(figsize=(10,10))
sns.countplot(data['horsepower'])
plt.show()
```



```
[33]: data['horsepower'].describe()
[33]: count
               201.000000
      mean
               103.405534
      std
                37.365700
                48.000000
      min
      25%
                70.000000
      50%
                95.000000
      75%
               116.000000
               262.000000
      max
      Name: horsepower, dtype: float64
[34]: # Handling Outliers
      sns.boxplot(data['horsepower'])
```

warnings.warn(

[34]: <AxesSubplot:xlabel='horsepower'>

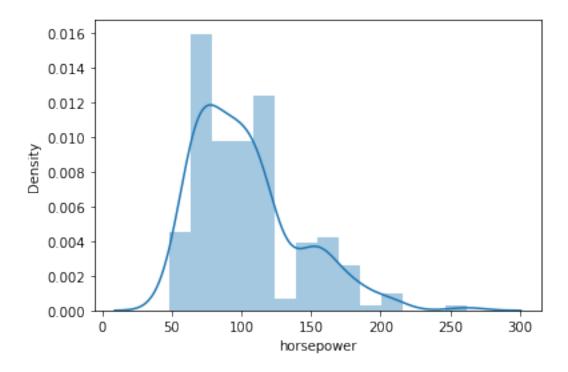


[35]: sns.distplot(data['horsepower'],kde=True)

C:\Users\87548\anaconda3\lib\site-packages\seaborn\distributions.py:2557:
FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)

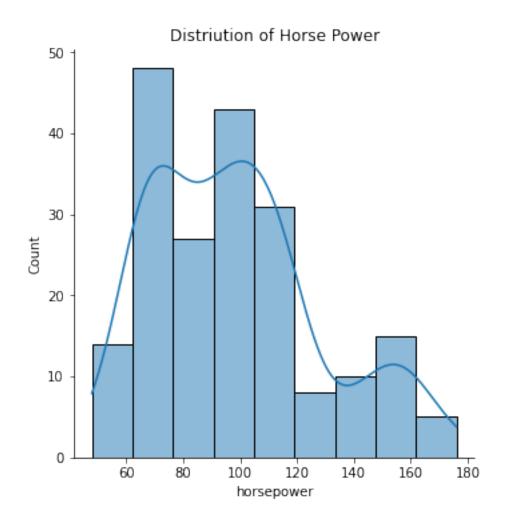
[35]: <AxesSubplot:xlabel='horsepower', ylabel='Density'>



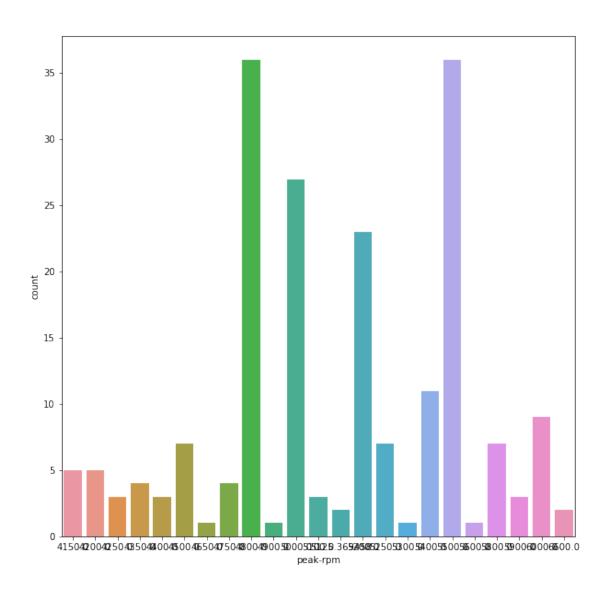
```
[36]: data['horsepower']=data['horsepower'].apply(lambda x : data['horsepower'].

→mean() if (x>180) else x)

[37]: sns.displot(data['horsepower'],kde=True)
plt.title('Distriution of Horse Power')
plt.show()
```

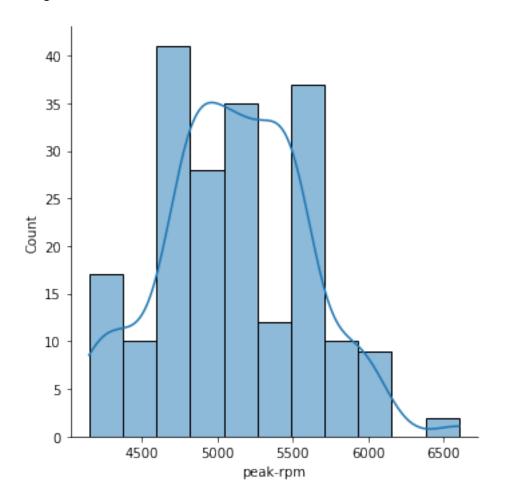


```
[38]: # Peak RPM
plt.figure(figsize=(10,10))
sns.countplot(data['peak-rpm'])
plt.show()
```



```
[39]: data['peak-rpm'].describe()
[39]: count
                 201.000000
                5117.665368
      mean
                 478.113805
      \operatorname{std}
                4150.000000
      min
      25%
                4800.000000
      50%
                5125.369458
      75%
                5500.000000
                6600.000000
      max
      Name: peak-rpm, dtype: float64
[40]: # Handling Outliers
      sns.displot(data['peak-rpm'],kde=True)
```

[40]: <seaborn.axisgrid.FacetGrid at 0x2500ee112b0>

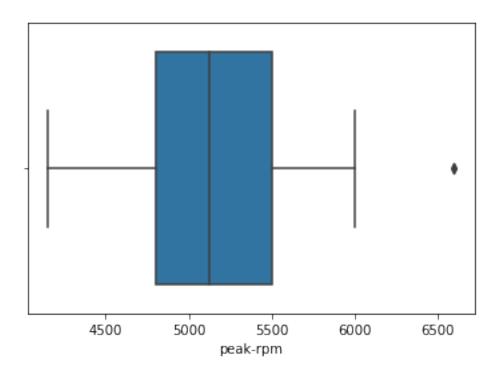


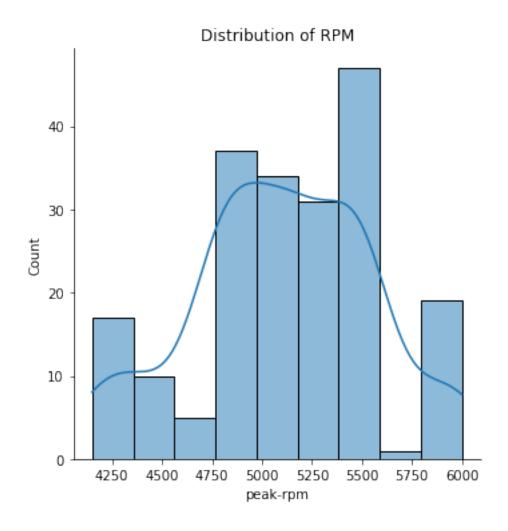
[41]: sns.boxplot(data['peak-rpm'])

C:\Users\87548\anaconda3\lib\site-packages\seaborn_decorators.py:36:
FutureWarning: Pass the following variable as a keyword arg: x. 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.

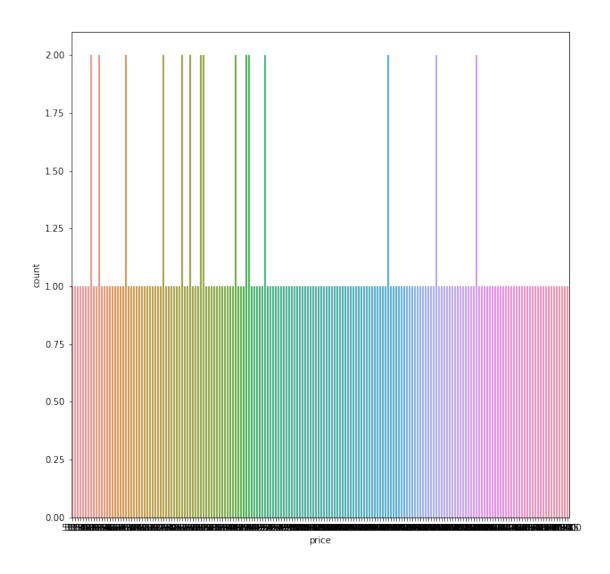
warnings.warn(

[41]: <AxesSubplot:xlabel='peak-rpm'>





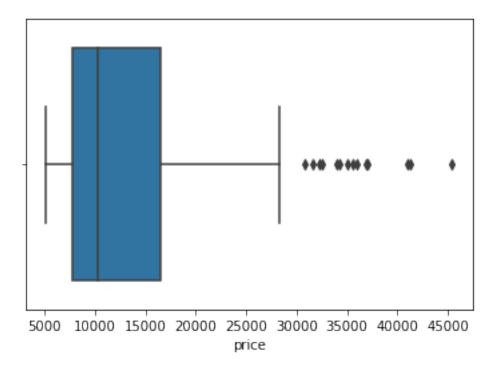
```
[49]: # Price
plt.figure(figsize=(10,10))
sns.countplot(data['price'])
plt.show()
```



```
[45]: data['price'].describe()
[45]: count
                 201.000000
      mean
               13207.129353
      std
                7947.066342
      min
                5118.000000
      25%
                7775.000000
      50%
               10295.000000
      75%
               16500.000000
      max
               45400.000000
      Name: price, dtype: float64
[50]: # Handling Outliers
      sns.boxplot(data['price'])
```

warnings.warn(

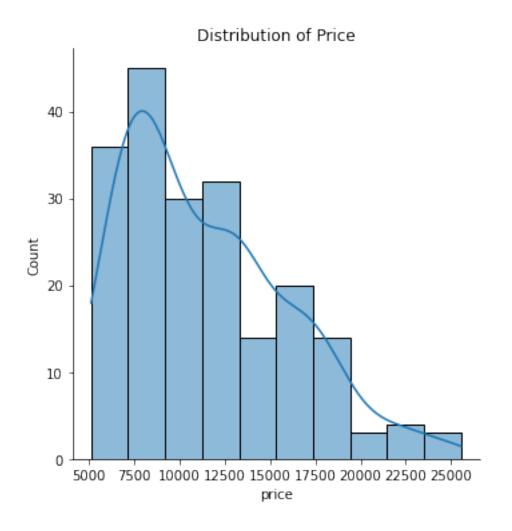
[50]: <AxesSubplot:xlabel='price'>



```
[51]: data['price']=data['price'].apply(lambda x:data['price'].mean() if (x>28000)

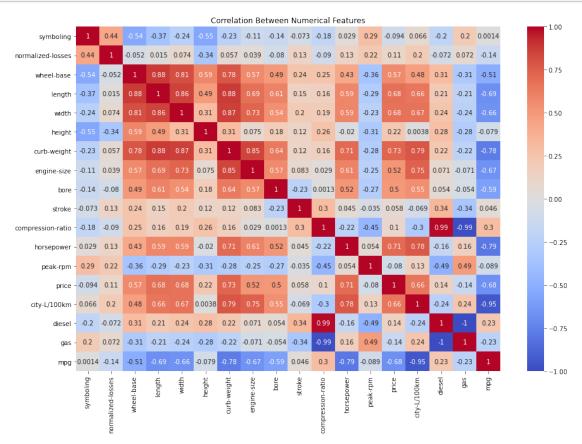
delse x)

[52]: sns.displot(data['price'],kde=True)
plt.title('Distribution of Price')
plt.show()
```



```
[58]: data.select_dtypes(include='number').head(2)
[58]:
        symboling normalized-losses
                                      wheel-base
                                                     length
                                                                width height \
                                                   0.811148
                                                                         48.8
                3
                                122.0
                                             88.6
                                                             0.890278
      0
                3
      1
                                122.0
                                             88.6
                                                   0.811148
                                                             0.890278
                                                                         48.8
        curb-weight engine-size bore
                                         stroke
                                                compression-ratio horsepower \
      0
                2548
                                   3.47
                                           2.68
                                                               9.0
                                                                         111.0
                              130
                2548
                                           2.68
                                                               9.0
      1
                              130 3.47
                                                                         111.0
                    price city-L/100km diesel
        peak-rpm
                                                        mpg
           5000.0
                  13495.0
                               11.190476
      0
                                                       24.0
           5000.0 16500.0
                               11.190476
                                               0
                                                       24.0
[59]: plt.figure(figsize=(15,10))
      sns.heatmap(data.select_dtypes(include='number').
       ⇔corr(),annot=True,cmap='coolwarm')
```

plt.title('Correlation Between Numerical Features') plt.show()



Combing height, weight, volumes

```
[61]: data['vol'] = (data['width']*data['length']*data['height'])/(12.54**3)
data.drop(['width','length','height'],axis=1,inplace=True)
```

[63]: data.select_dtypes(include='number').head(2)

```
[63]:
         symboling normalized-losses
                                         wheel-base
                                                      curb-weight
                                                                    engine-size
                                                                                 bore
                  3
      0
                                  122.0
                                               88.6
                                                              2548
                                                                                 3.47
                                                                            130
      1
                  3
                                  122.0
                                               88.6
                                                              2548
                                                                            130
                                                                                 3.47
         stroke
                  compression-ratio
                                     horsepower
                                                 peak-rpm
                                                               price
                                                                       city-L/100km
           2.68
      0
                                 9.0
                                           111.0
                                                     5000.0
                                                             13495.0
                                                                          11.190476
```

111.0

5000.0

16500.0

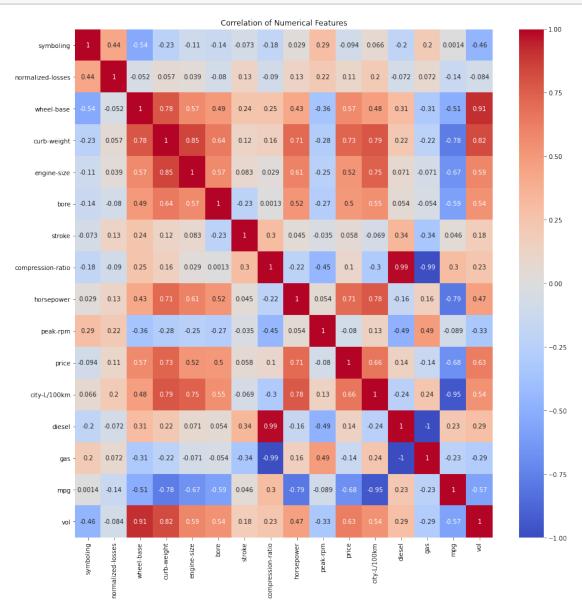
11.190476

9.0

diesel gas mpg vol 0 0 1 24.0 0.017871 1 0 1 24.0 0.017871

1

2.68

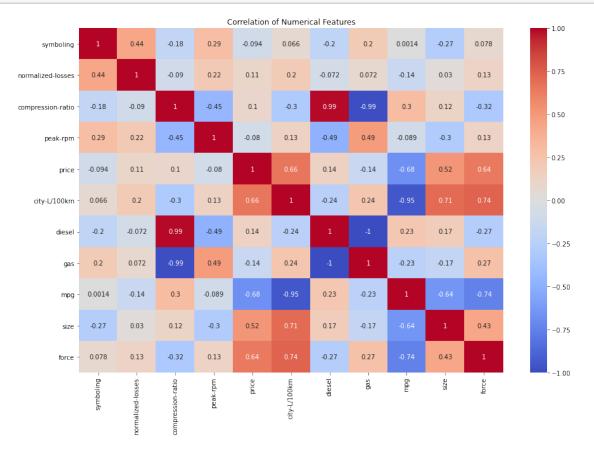


```
[66]: # Feature Engineering with other features

data['weights']=(data['wheel-base']*data['curb-weight']*data['engine-size'])/3
data.drop(['wheel-base','curb-weight','engine-size'],axis=1,inplace=True)
```

```
[67]: data['size'] = (data['vol']*data['weights'])/9.81
data.drop(['vol','weights'],axis=1,inplace=True)
```

```
[68]: data['force'] = (data['horsepower']/(data['bore']*data['stroke']))
data.drop(['horsepower','bore','stroke'],axis=1,inplace=True)
```



```
[70]: data.columns
```



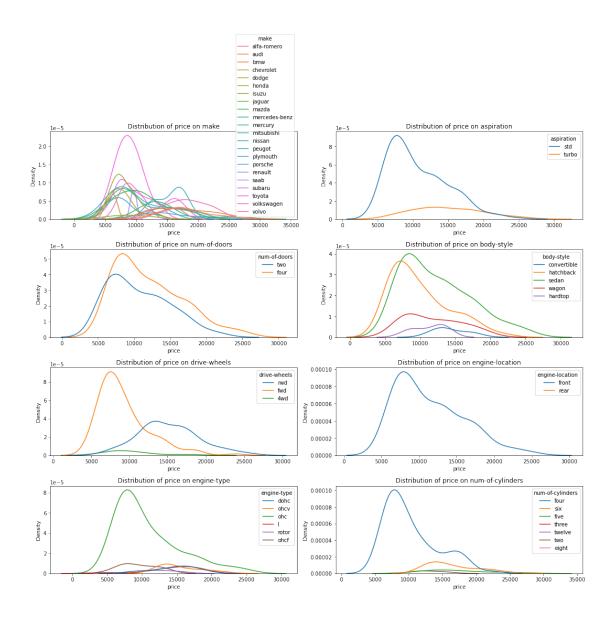
```
[73]: # Analysing categorical features

data.select_dtypes(exclude='number').head(2)
```

[73]: make aspiration num-of-doors body-style drive-wheels \
0 alfa-romero std two convertible rwd
1 alfa-romero std two convertible rwd

engine-location engine-type num-of-cylinders fuel-system horsepower-binned 0 front dohc four mpfi Medium 1 front dohc four mpfi Medium

```
[74]: cat_col = data.select_dtypes(exclude='number').columns
[82]: fig,axes = plt.subplots(4,2)
      fig.set_figheight(15)
      fig.set_figwidth(15)
      for ax,col in zip(axes.flatten(),cat_col):
          sns.kdeplot(ax=ax,data=data,x='price',hue=col)
          ax.set_title('Distribution of price on '+str(col))
          plt.tight_layout()
     C:\Users\87548\anaconda3\lib\site-packages\seaborn\distributions.py:306:
     UserWarning: Dataset has 0 variance; skipping density estimate.
       warnings.warn(msg, UserWarning)
     C:\Users\87548\anaconda3\lib\site-packages\seaborn\distributions.py:306:
     UserWarning: Dataset has 0 variance; skipping density estimate.
       warnings.warn(msg, UserWarning)
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     C:\Users\87548\anaconda3\lib\site-packages\seaborn\distributions.py:306:
     UserWarning: Dataset has 0 variance; skipping density estimate.
       warnings.warn(msg, UserWarning)
     C:\Users\87548\anaconda3\lib\site-packages\seaborn\distributions.py:306:
     UserWarning: Dataset has 0 variance; skipping density estimate.
       warnings.warn(msg, UserWarning)
```



```
[76]: fig,axes = plt.subplots(4,2)
fig.set_figheight(15)
fig.set_figwidth(15)

for ax,col in zip(axes.flatten(),cat_col):
    sns.kdeplot(ax=ax,data=data,x='mpg',hue=col)
    ax.set_title('Distribution of price on '+str(col))
    plt.tight_layout()
```

C:\Users\87548\anaconda3\lib\site-packages\seaborn\distributions.py:306:
UserWarning: Dataset has 0 variance; skipping density estimate.
warnings.warn(msg, UserWarning)

C:\Users\87548\anaconda3\lib\site-packages\seaborn\distributions.py:306:

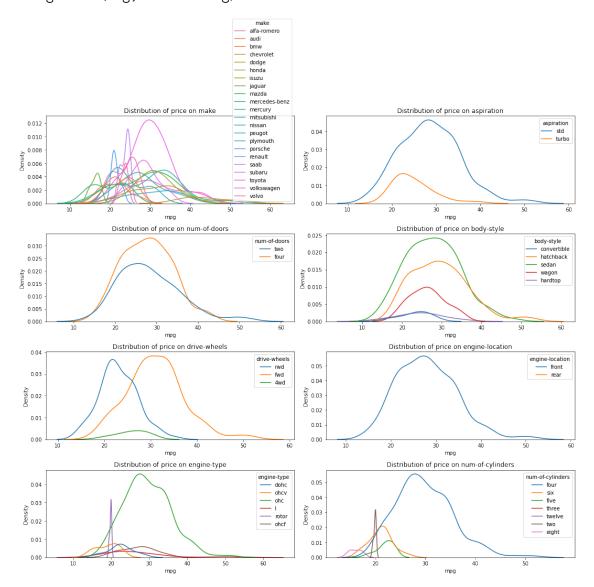
UserWarning: Dataset has 0 variance; skipping density estimate. warnings.warn(msg, UserWarning)

C:\Users\87548\anaconda3\lib\site-packages\seaborn\distributions.py:306:
UserWarning: Dataset has 0 variance; skipping density estimate.
warnings.warn(msg, UserWarning)

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warnings.warn(msg, UserWarning)



[]:[