

```

!pip install pydantic
!pip install PyYAML
!pip install jinja2
!pip install visions
!pip install htmlmin
!pip install phik
!pip install requests
!pip install tqdm
!pip install seaborn
!pip install multimethod
!pip install statsmodels
!pip install typeguard
!pip install imagehash
!pip install wordcloud
!pip install dacite
!pip install numba

Requirement already satisfied: pydantic in /usr/local/lib/python3.10/dist-packages (1.10.13)
Requirement already satisfied: typing-extensions>=4.2.0 in /usr/local/lib/python3.10/dist-packages (from pydantic) (4.5.0)
Requirement already satisfied: PyYAML in /usr/local/lib/python3.10/dist-packages (6.0.1)
Requirement already satisfied: jinja2 in /usr/local/lib/python3.10/dist-packages (3.1.2)
Requirement already satisfied: MarkupSafe>=2.0 in /usr/local/lib/python3.10/dist-packages (from jinja2) (2.1.3)
Collecting visions
  Downloading visions-0.7.5-py3-none-any.whl (102 kB)
    102.7/102.7 kB 2.6 MB/s eta 0:00:00
Requirement already satisfied: numpy in /usr/local/lib/python3.10/dist-packages (from visions) (1.23.5)
Requirement already satisfied: pandas>=0.25.3 in /usr/local/lib/python3.10/dist-packages (from visions) (1.5.3)
Requirement already satisfied: attrs>=19.3.0 in /usr/local/lib/python3.10/dist-packages (from visions) (23.1.0)
Requirement already satisfied: networkx>=2.4 in /usr/local/lib/python3.10/dist-packages (from visions) (3.2)
Collecting tangled-up-in-unicode>=0.0.4 (from visions)
  Downloading tangled_up_in_unicode-0.2.0-py3-none-any.whl (4.7 MB)
    4.7/4.7 MB 58.1 MB/s eta 0:00:00
Collecting multimethod>=1.4 (from visions)
  Downloading multimethod-1.10-py3-none-any.whl (9.9 kB)
Requirement already satisfied: python-dateutil>=2.8.1 in /usr/local/lib/python3.10/dist-packages (from pandas>=0.25.3->visions) (2.8.1)
Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.10/dist-packages (from pandas>=0.25.3->visions) (2023.3.post1)
Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.10/dist-packages (from python-dateutil>=2.8.1->pandas>=0.25.3->visions) (1.16.0)
Installing collected packages: tangled-up-in-unicode, multimethod, visions
Successfully installed multimethod-1.10 tangled-up-in-unicode-0.2.0 visions-0.7.5
Collecting htmlmin
  Downloading htmlmin-0.1.12.tar.gz (19 kB)
  Preparing metadata (setup.py) ... done
Building wheels for collected packages: htmlmin
  Building wheel for htmlmin (setup.py) ... done
  Created wheel for htmlmin: filename=htmlmin-0.1.12-py3-none-any.whl size=27081 sha256=9f745e5d089b7707e22823d2c337f1fe29d549ca6d87c
  Stored in directory: /root/.cache/pip/wheels/dd/91/29/a79cecb328d01739e64017b6fb9a1ab9d8cb1853098ec5966d
Successfully built htmlmin
Installing collected packages: htmlmin
Successfully installed htmlmin-0.1.12
Collecting phik
  Downloading phik-0.12.3-cp310-cp310-manylinux_2_17_x86_64.manylinux2014_x86_64.whl (679 kB)
    679.5/679.5 kB 6.4 MB/s eta 0:00:00
Requirement already satisfied: numpy>=1.18.0 in /usr/local/lib/python3.10/dist-packages (from phik) (1.23.5)
Requirement already satisfied: scipy>=1.5.2 in /usr/local/lib/python3.10/dist-packages (from phik) (1.11.3)
Requirement already satisfied: pandas>=0.25.1 in /usr/local/lib/python3.10/dist-packages (from phik) (1.5.3)
Requirement already satisfied: matplotlib>=2.2.3 in /usr/local/lib/python3.10/dist-packages (from phik) (3.7.1)
Requirement already satisfied: joblib>=0.14.1 in /usr/local/lib/python3.10/dist-packages (from phik) (1.3.2)
Requirement already satisfied: contourpy>=1.0.1 in /usr/local/lib/python3.10/dist-packages (from matplotlib>=2.2.3->phik) (1.1.1)
Requirement already satisfied: cycler>=0.10 in /usr/local/lib/python3.10/dist-packages (from matplotlib>=2.2.3->phik) (0.12.1)
Requirement already satisfied: fonttools>=4.22.0 in /usr/local/lib/python3.10/dist-packages (from matplotlib>=2.2.3->phik) (4.43.1)
Requirement already satisfied: kiwisolver>=1.0.1 in /usr/local/lib/python3.10/dist-packages (from matplotlib>=2.2.3->phik) (1.4.5)
Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.10/dist-packages (from matplotlib>=2.2.3->phik) (23.2)
Requirement already satisfied: pillow>=6.2.0 in /usr/local/lib/python3.10/dist-packages (from matplotlib>=2.2.3->phik) (9.4.0)
Requirement already satisfied: pyparsing>=2.3.1 in /usr/local/lib/python3.10/dist-packages (from matplotlib>=2.2.3->phik) (3.1.1)
Requirement already satisfied: python-dateutil>=2.7 in /usr/local/lib/python3.10/dist-packages (from matplotlib>=2.2.3->phik) (2.8.2)
Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.10/dist-packages (from pandas>=0.25.1->phik) (2023.3.post1)
Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.10/dist-packages (from python-dateutil>=2.7->matplotlib>=2.2.3->phik) (1.16.0)
Installing collected packages: phik
Successfully installed phik-0.12.3
Requirement already satisfied: requests in /usr/local/lib/python3.10/dist-packages (2.31.0)
Requirement already satisfied: charset-normalizer<4,>=2 in /usr/local/lib/python3.10/dist-packages (from requests) (3.3.1)
Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.10/dist-packages (from requests) (3.4)
Requirement already satisfied: urllib3<3,>=1.21.1 in /usr/local/lib/python3.10/dist-packages (from requests) (2.0.7)
Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.10/dist-packages (from requests) (2023.7.22)

```

```

import pandas as pd
import numpy as np
import seaborn as sns

```

```
import matplotlib.pyplot as plt
```

```
! python --version
```

```
Python 3.10.12
```

```
import sys
```

```
!{sys.executable} -m pip install -U ydata-profiling
```

```
!jupyter nbextension enable --py widgetsnbextension
```

```

  Downloading imagehash-4.3.1-py2.py3-none-any.whl (290 kB)
    296.5/296.5 kB 9.2 MB/s eta 0:00:00
Requirement already satisfied: wordcloud>=1.9.1 in /usr/local/lib/python3.10/dist-packages (from ydata-profiling) (1.9.2)
Collecting dacite>=1.8 (from ydata-profiling)
  Downloading dacite-1.8.1-py3-none-any.whl (14 kB)
Requirement already satisfied: numba<0.59.0,>=0.56.0 in /usr/local/lib/python3.10/dist-packages (from ydata-profiling) (0.56.4)
Requirement already satisfied: PyWavelets in /usr/local/lib/python3.10/dist-packages (from imagehash==4.3.1->ydata-profiling) (1.4.1)
Requirement already satisfied: pillow in /usr/local/lib/python3.10/dist-packages (from imagehash==4.3.1->ydata-profiling) (9.4.0)
Requirement already satisfied: attrs>=19.3.0 in /usr/local/lib/python3.10/dist-packages (from visions[type_image_path]==0.7.5->ydata-profiling) (23.1.0)
Requirement already satisfied: networkx>=2.4 in /usr/local/lib/python3.10/dist-packages (from visions[type_image_path]==0.7.5->ydata-profiling) (3.1)
Collecting tangled-up-in-unicode>=0.0.4 (from visions[type_image_path]==0.7.5->ydata-profiling)
  Downloading tangled_up_in_unicode-0.2.0-py3-none-any.whl (4.7 MB)
    4.7/4.7 MB 15.5 MB/s eta 0:00:00
Requirement already satisfied: MarkupSafe>=2.0 in /usr/local/lib/python3.10/dist-packages (from jinja2<3.2,>=2.11.1->ydata-profiling) (2.1.2)
Requirement already satisfied: contourpy>=1.0.1 in /usr/local/lib/python3.10/dist-packages (from matplotlib<=3.7.3,>=3.2->ydata-profiling) (1.0.7)
Requirement already satisfied: cycler>=0.10 in /usr/local/lib/python3.10/dist-packages (from matplotlib<=3.7.3,>=3.2->ydata-profiling) (0.11.0)
Requirement already satisfied: fonttools>=4.22.0 in /usr/local/lib/python3.10/dist-packages (from matplotlib<=3.7.3,>=3.2->ydata-profiling) (4.22.0)
Requirement already satisfied: kiwisolver>=1.0.1 in /usr/local/lib/python3.10/dist-packages (from matplotlib<=3.7.3,>=3.2->ydata-profiling) (1.4.5)
Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.10/dist-packages (from matplotlib<=3.7.3,>=3.2->ydata-profiling) (23.1)
Requirement already satisfied: pyparsing>=2.3.1 in /usr/local/lib/python3.10/dist-packages (from matplotlib<=3.7.3,>=3.2->ydata-profiling) (3.1.0)
Requirement already satisfied: python-dateutil>=2.7 in /usr/local/lib/python3.10/dist-packages (from matplotlib<=3.7.3,>=3.2->ydata-profiling) (2.8.2)
Requirement already satisfied: llvmlite<0.40,>=0.39.0dev0 in /usr/local/lib/python3.10/dist-packages (from numba<0.59.0,>=0.56.0->ydata-profiling) (0.40.0)
Requirement already satisfied: setuptools in /usr/local/lib/python3.10/dist-packages (from numba<0.59.0,>=0.56.0->ydata-profiling) (68.0.0)
Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.10/dist-packages (from pandas!=1.4.0,<2.1,>1.1->ydata-profiling) (2023.3)
Requirement already satisfied: joblib>=0.14.1 in /usr/local/lib/python3.10/dist-packages (from phik<0.13,>=0.11.1->ydata-profiling) (1.3.2)
Collecting annotated-types>=0.4.0 (from pydantic>=2->ydata-profiling)
  Downloading annotated_types-0.6.0-py3-none-any.whl (12 kB)
Collecting pydantic-core==2.10.1 (from pydantic>=2->ydata-profiling)
  Downloading pydantic_core-2.10.1-cp310-cp310-manylinux_2_17_x86_64.manylinux2014_x86_64.whl (2.0 MB)
    2.0/2.0 MB 25.5 MB/s eta 0:00:00
Collecting typing-extensions>=4.6.1 (from pydantic>=2->ydata-profiling)
  Downloading typing_extensions-4.8.0-py3-none-any.whl (31 kB)
Requirement already satisfied: charset-normalizer<4,>=2 in /usr/local/lib/python3.10/dist-packages (from requests<3,>=2.24.0->ydata-profiling) (3.2.0)
Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.10/dist-packages (from requests<3,>=2.24.0->ydata-profiling) (3.4)
Requirement already satisfied: urllib3<3,>=1.21.1 in /usr/local/lib/python3.10/dist-packages (from requests<3,>=2.24.0->ydata-profiling) (2.0.7)
Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.10/dist-packages (from requests<3,>=2.24.0->ydata-profiling) (2023.7.22)
Requirement already satisfied: patsy>=0.5.2 in /usr/local/lib/python3.10/dist-packages (from statsmodels<1,>=0.13.2->ydata-profiling) (0.5.2)
Requirement already satisfied: six in /usr/local/lib/python3.10/dist-packages (from patsy>=0.5.2->statsmodels<1,>=0.13.2->ydata-profiling) (1.16.0)
Building wheels for collected packages: htmlmin
  Building wheel for htmlmin (setup.py) ... done
  Created wheel for htmlmin: filename=htmlmin-0.1.12-py3-none-any.whl size=27081 sha256=93e1cd9576d35f4d7bbcd3fa92d80dadad6b558e086fa
  Stored in directory: /root/.cache/pip/wheels/dd/91/29/a79cecb328d01739e64017b6fb9a1ab9d8cb1853098ec5966d
Successfully built htmlmin
Installing collected packages: htmlmin, typing-extensions, tangled-up-in-unicode, multimethod, dacite, annotated-types, typeguard, py
Attempting uninstall: typing-extensions
  Found existing installation: typing_extensions 4.5.0
  Uninstalling typing_extensions-4.5.0:
    Successfully uninstalled typing_extensions-4.5.0
ERROR: Operation cancelled by user
Enabling notebook extension jupyter-js-widgets/extension...
Paths used for configuration of notebook:
  /root/.jupyter/nbconfig/notebook.json
Paths used for configuration of notebook:

- Validating: OK
Paths used for configuration of notebook:
  /root/.jupyter/nbconfig/notebook.json

```

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

Choose Files used_cars.csv

- used_cars.csv(text/csv) - 611832 bytes, last modified: 9/25/2023 - 100% done

Saving used_cars.csv to used_cars.csv

```
data_file = "used_cars.csv"
df= pd.read_csv(data_file)
```

	brand	model	model_year	milage	fuel_type	engine	transmission	ext_
0	Ford	Utility Police Interceptor Base	2013	51,000 mi.	E85 Flex Fuel	300.0HP 3.7L V6 Cylinder Engine Flex Fuel Capa...	6-Speed A/T	Bl
1	Hyundai	Palisade SEL	2021	34,742 mi.	Gasoline	3.8L V6 24V GDI DOHC	8-Speed Automatic	Moonl Cl
2	Lexus	RX 350 RX 350	2022	22,372 mi.	Gasoline	3.5 Liter DOHC	Automatic	E
3	INFINITI	Q50 Hybrid Sport	2015	88,900 mi.	Hybrid	354.0HP 3.5L V6 Cylinder Engine Gas/Electric H...	7-Speed A/T	Bl
4	Audi	Q3 45 S line Premium Plus	2021	9,835 mi.	Gasoline	2.0L I4 16V GDI DOHC Turbo	8-Speed Automatic	Gla W Met
...

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```
print(df.dtypes)
```

```
brand      object
model      object
model_year  int64
milage     object
fuel_type  object
engine     object
transmission object
ext_col    object
int_col    object
accident   object
clean_title object
price      object
dtype: object
```

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 4009 entries, 0 to 4008
Data columns (total 12 columns):
#   Column          Non-Null Count  Dtype
---  -
0   brand           4009 non-null  object
1   model           4009 non-null  object
2   model_year      4009 non-null  int64
3   milage          4009 non-null  object
4   fuel_type       3839 non-null  object
5   engine          4009 non-null  object
6   transmission    4009 non-null  object
7   ext_col         4009 non-null  object
8   int_col         4009 non-null  object
9   accident        3896 non-null  object
10  clean_title     3413 non-null  object
11  price           4009 non-null  object
dtypes: int64(1), object(11)
memory usage: 376.0+ KB
```

```
df['milage'] = df['milage'].str.replace(r'\D', '', regex=True)
df['milage'] = df['milage'].astype(float)
df['price'] = df['price'].str.replace(r'\D', '', regex=True)
df['price'] = df['price'].astype(float)
df
```

	brand	model	model_year	milage	fuel_type	engine	transmission	ext_
0	Ford	Utility Police Interceptor Base	2013	51000.0	E85 Flex Fuel	300.0HP 3.7L V6 Cylinder Engine Flex Fuel Capa...	6-Speed A/T	B
1	Hyundai	Palisade SEL	2021	34742.0	Gasoline	3.8L V6 24V GDI DOHC	8-Speed Automatic	Moon C
2	Lexus	RX 350 RX 350	2022	22372.0	Gasoline	3.5 Liter DOHC	Automatic	I
3	INFINITI	Q50 Hybrid Sport	2015	88900.0	Hybrid	354.0HP 3.5L V6 Cylinder Engine Gas/Electric H...	7-Speed A/T	B
4	Audi	Q3 45 S line Premium Plus	2021	9835.0	Gasoline	2.0L I4 16V GDI DOHC Turbo	8-Speed Automatic	Gle V Mel
...

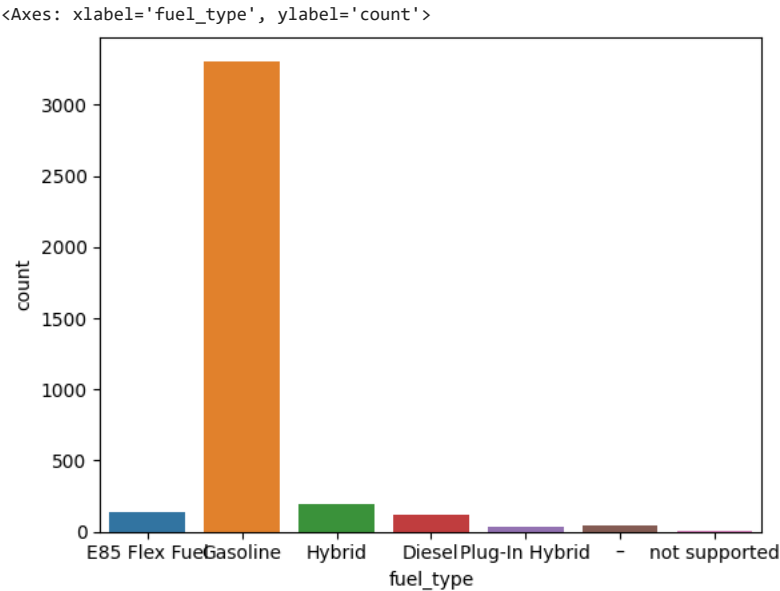
```
df.isna().sum()
#df.dtypes

brand      0
model      0
model_year 0
milage     0
fuel_type  170
engine     0
transmission 0
ext_col    0
int_col    0
accident   113
clean_title 596
price      0
dtype: int64

df['accident'] = df['accident'].replace({'At least 1 accident or damage reported' : 'Yes',
'None reported': 'No'})
df['clean_title'] = df['clean_title'].fillna('No')
#this last part is done by me
df['accident'] = df['accident'].fillna('No')
df
```

	brand	model	model_year	milage	fuel_type	engine	transmission	ext_
0	Ford	Utility Police Interceptor Base	2013	51000.0	E85 Flex Fuel	300.0HP 3.7L V6 Cylinder Engine Flex Fuel	6-Speed A/T	B

```
sns.countplot(x = 'fuel_type', data = df)
```



```
df['fuel_type'] = df['fuel_type'].fillna('Gasoline')
df
```

	brand	model	model_year	milage	fuel_type	engine	transmission	ext_
0	Ford	Utility Police Interceptor Base	2013	51000.0	E85 Flex Fuel	300.0HP 3.7L V6 Cylinder Engine Flex Fuel Capa...	6-Speed A/T	B
1	Hyundai	Palisade SEL	2021	34742.0	Gasoline	3.8L V6 24V GDI DOHC	8-Speed Automatic	Moon C
2	Lexus	RX 350 RX 350	2022	22372.0	Gasoline	3.5 Liter DOHC	Automatic	I
3	INFINITI	Q50 Hybrid Sport	2015	88900.0	Hybrid	354.0HP 3.5L V6 Cylinder Engine Gas/Electric H...	7-Speed A/T	B
4	Audi	Q3 45 S line Premium Plus	2021	9835.0	Gasoline	2.0L I4 16V GDI DOHC Turbo	8-Speed Automatic	Gle V Mel
...

```
df.isna().sum()
#df.dtypes
```

```
brand      0
model      0
model_year 0
milage     0
fuel_type  0
engine     0
transmission 0
```

```
ext_col      0
int_col      0
accident     0
clean_title  0
price        0
dtype: int64

df.dtypes

brand      object
model      object
model_year  int64
milage     float64
fuel_type  object
engine     object
transmission object
ext_col    object
int_col    object
accident   object
clean_title object
price      float64
dtype: object

Current_Year = 2023
df['age'] = Current_Year - df['model_year']
df['age'] = df['age'].astype(np.int64)
df

   brand  model  model_year  milage  fuel_type  engine  transmission  ext_
0  Ford  Utility Police Interceptor Base  2013  51000.0  E85 Flex Fuel  300.0HP 3.7L V6 Cylinder Engine Flex Fuel Capa...  6-Speed A/T  B
1  Hyundai  Palisade SEL  2021  34742.0  Gasoline  3.8L V6 24V GDI DOHC  8-Speed Automatic  Moon C
2  Lexus  RX 350 RX 350  2022  22372.0  Gasoline  3.5 Liter DOHC  Automatic  I
3  INFINITI  Q50 Hybrid Sport  2015  88900.0  Hybrid  354.0HP 3.5L V6 Cylinder Engine Gas/Electric H...  7-Speed A/T  B
4  Audi  Q3 45 S line Premium Plus  2021  9835.0  Gasoline  2.0L I4 16V GDI DOHC Turbo  8-Speed Automatic  Gle V Mel
...  ...  ...  ...  ...  ...  ...  ...  ...
5  ...  ...  ...  ...  ...  ...  ...  ...
```

```
df_new = df.drop(['model_year'], axis=1)
df_new
```

	brand	model	milage	fuel_type	engine	transmission	ext_col	int_col
0	Ford	Utility Police Interceptor Base	51000.0	E85 Flex Fuel	300.0HP 3.7L V6 Cylinder Engine Flex Fuel Capa...	6-Speed A/T	Black	Black
1	Hyundai	Palisade SEL	34742.0	Gasoline	3.8L V6 24V GDI DOHC	8-Speed Automatic	Moonlight Cloud	Gray

```
plt.figure(figsize=(20,8))
plt.subplot(1,2,1)
plt.title('Car Selling price Distribution Plot')
sns.distplot(df_new['price'])
sns.set_style('darkgrid')
```

```
plt.subplot(1,2,2)
plt.title('Car Selling price Spread')
sns.boxplot(y=df_new['price'])
sns.set_style('darkgrid')
```

```
plt.show()
```

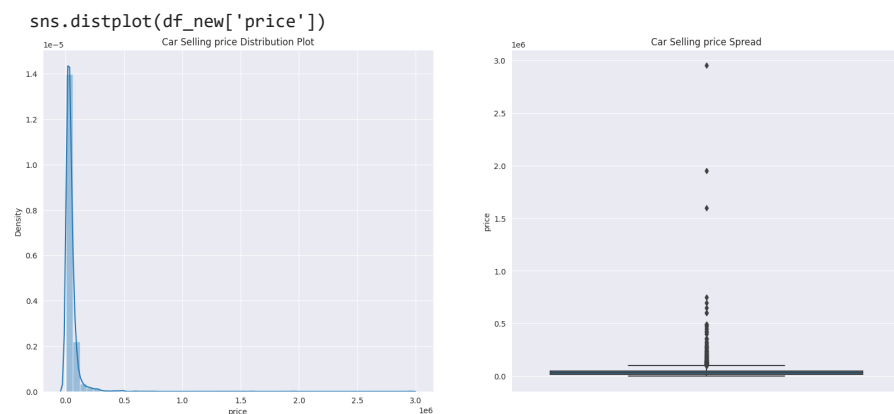
<ipython-input-17-69c6c010fec7>:4: UserWarning:

`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see

<https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>



```
# plotting the target-age scatter graph
sns.scatterplot(data=df_new, x="age", y="price")
sns.set_style('darkgrid')
plt.title("Selling price by age", size=12)
plt.ylabel("Selling price (Thousand bucks)", size=10)
```

```
plt.xlabel("Age", size=10)
plt.show()
```



```
#Distribution
plt.figure(figsize=(20,8))
plt.subplot(1,2,1)
plt.title('Car Milage Plot')
sns.distplot(df_new.milage, color='green')
```

```
#Spread
plt.subplot(1,2,2)
plt.title('Car milage Spread')
sns.boxplot(y=df_new.milage)

plt.show()
```



```
<ipython-input-19-cf6510de259a>:5: UserWarning:
```

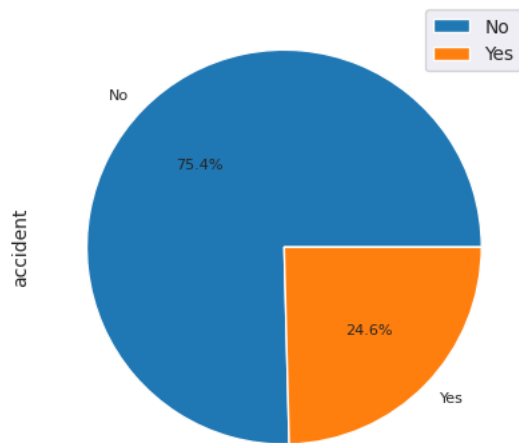
```
`distplot` is a deprecated function and will be removed in seaborn v0.14.0.
```

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

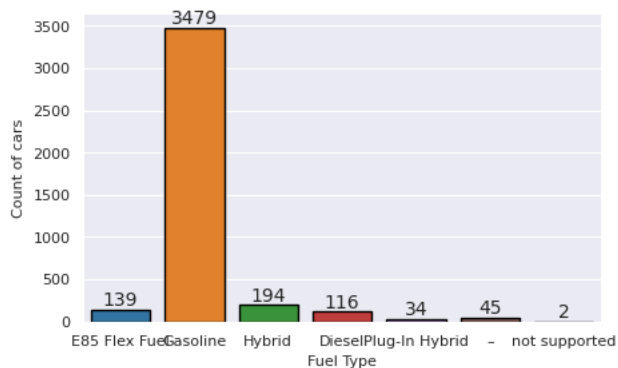
For a guide to updating your code to use the new functions, please see

<https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>

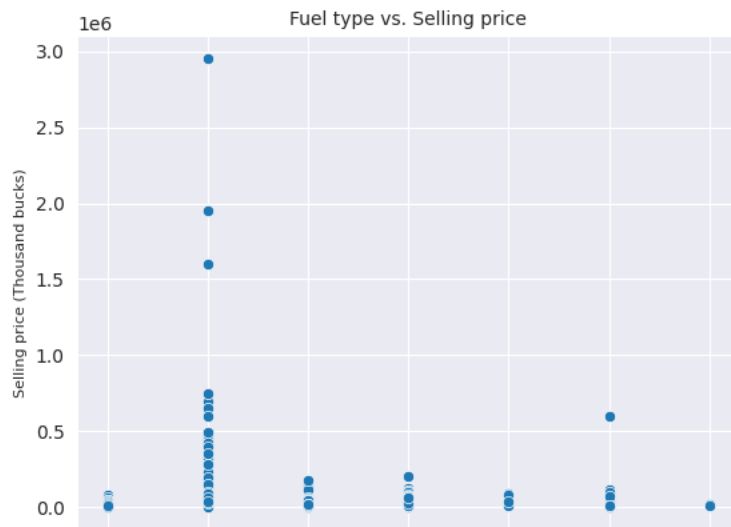
```
df_sym = pd.DataFrame(df_new['accident'].value_counts())
df_sym.plot.pie(subplots=True, labels = df_sym.index.values, autopct='%1.1f%%', fontsize=8)
# Unsquish the pie.
plt.gca().set_aspect('equal')
plt.show()
```



```
# Count of cars by fuel_type
plt.figure(figsize = (5, 3))
ax=sns.countplot(data=df_new, x=df.fuel_type, ec='black')
sns.set_style('darkgrid')
for cont in ax.containers:
    ax.bar_label(cont)
plt.ylabel('Count of cars', size=8)
plt.yticks(size=8)
plt.xlabel('Fuel Type', size=8)
plt.xticks(size=8)
plt.show()
```



```
# plotting the target-Fuel type scatter graph
sns.scatterplot(data=df_new, x="fuel_type", y="price")
sns.set_style('darkgrid')
plt.title("Fuel type vs. Selling price", size=10)
plt.ylabel("Selling price (Thousand bucks)", size=8)
plt.xlabel("Fuel type", size=8)
plt.show()
```



```
obj_cols = [col for col in df_new.columns if df_new[col].dtypes == 'O']
print('Number of Qualitative Variable: ', len(obj_cols))
```

```
def bar_charts(data, obj_cols):
    col_counter = 0
    data = df_new.copy()
    for col in obj_cols:
        data[col].value_counts().plot(kind = "bar",figsize=(10,2),fontsize=10)
        plt.xlabel(col)
        plt.title(col)
        plt.show()
        col_counter += 1
    print(col_counter, "variables have been plotted")

bar_charts(df_new, obj_cols)
```

brand



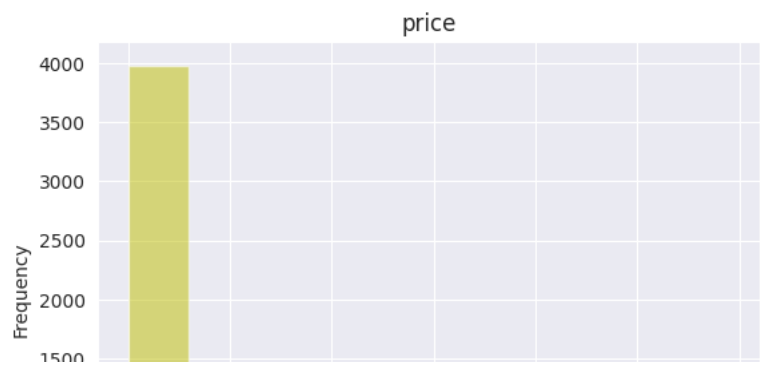
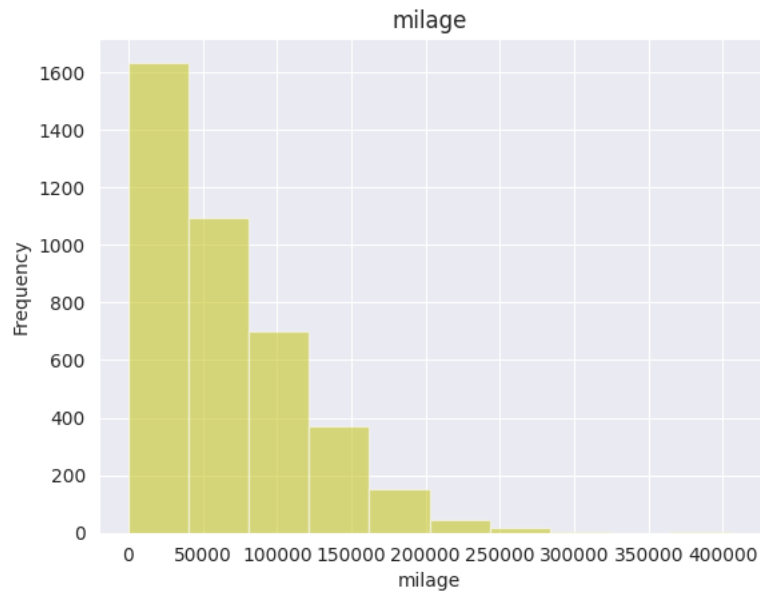
30




```
num_cols = [col for col in df_new.columns if df_new[col].dtypes != 'O']  
print('Number of Numerical Variable: ', len(num_cols))
```

```
def hist_for_nums(data, numeric_cols):  
    col_counter = 0  
    data = data.copy()  
    for col in numeric_cols:  
        data[col].plot.hist(alpha=0.5, color='y')  
        plt.xlabel(col)  
        plt.title(col)  
        plt.show()  
        col_counter += 1  
    print(col_counter, "variables have been plotted")  
hist_for_nums(df_new, num_cols)
```

Number of Numerical Variable: 3



```
import ydata_profiling
from ydata_profiling import ProfileReport
profile = ProfileReport(df)
```

5000

A horizontal bar chart showing the distribution of numerical variables. The x-axis is labeled '5000' and ranges from 0 to 40,000 with major ticks every 5,000. The y-axis is labeled 'profile' and ranges from 0 to 4,000 with major ticks every 500. The chart consists of 10 olive-green bars. The first bar (0-5,000) has the highest frequency, around 4,000. The frequency decreases as the price increases, with the last bar (35,000-40,000) having a frequency near zero.

price Range	Frequency
0 - 5,000	~4,000
5,000 - 10,000	~3,500
10,000 - 15,000	~3,000
15,000 - 20,000	~2,500
20,000 - 25,000	~2,000
25,000 - 30,000	~1,500
30,000 - 35,000	~1,000
35,000 - 40,000	~500

profile

Summarize dataset: 38/38 [00:08<00:00, 2.96it/s, 100% Completed]

Generate report structure: 100% 1/1 [00:12<00:00, 12.98s/it]

Render HTML: 100% 1/1 [00:01<00:00, 1.30s/it]

Overview

```
from sklearn import preprocessing
le = preprocessing.LabelEncoder()
for x in df_new.columns:
    if df_new[x].dtypes=='object':
        df_new[x]=le.fit_transform(df_new[x].astype(str))
corr = df_new.corr()
corr
```

	brand	model	milage	fuel_type	engine	transmission	ext_col
brand	1.000000	-0.070170	-0.012389	0.033300	-0.066116	-0.005099	-0.002001
model	-0.070170	1.000000	0.031513	0.004079	-0.037443	-0.024244	-0.008342
milage	-0.012389	0.031513	1.000000	-0.096195	-0.227913	-0.043796	0.000891
fuel_type	0.033300	0.004079	-0.096195	1.000000	0.080890	0.094140	-0.010056
engine	-0.066116	-0.037443	-0.227913	0.080890	1.000000	-0.011988	-0.037665
transmission	-0.005099	-0.024244	-0.043796	0.094140	-0.011988	1.000000	0.001548
ext_col	-0.002001	-0.008342	0.000891	-0.010056	-0.037665	0.001548	1.000000
int_col	0.008545	0.040801	-0.051394	0.013986	0.023628	-0.030224	0.085077
accident	-0.023373	0.000537	0.301174	-0.038539	-0.098442	0.021412	-0.004037
clean_title	0.013011	-0.039634	0.253614	-0.004947	0.024433	-0.038643	0.014161
price	0.030957	-0.033313	-0.305528	0.008496	0.285172	0.036943	0.004035
age	0.001070	0.028237	0.617720	0.075813	0.148065	0.064506	0.036160

```
#df.drop(['model'], axis = 1)
df_new
```

	brand	model	milage	fuel_type	engine	transmission	ext_col	int_col	accident
0	14	1743	51000.0	1	581	16	29	14	1
1	19	1182	34742.0	2	566	32	185	71	1
2	27	1325	22372.0	2	541	40	38	14	0
3	20	1242	88900.0	3	724	23	29	14	0
4	3	1225	9835.0	2	200	32	120	14	0
...
4004	5	484	714.0	2	1060	33	50	75	0
4005	3	1464	10900.0	2	714	59	29	14	0
4006	43	1677	2116.0	2	1133	40	29	14	0
4007	14	666	33000.0	2	917	38	38	14	0
4008	4	1790	43000.0	2	356	38	128	31	1

```

X = df_new.iloc[:, list(range(10)) + [-1]]
y = df_new.iloc[:, -2]

X
#y

from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size= 0.2, random_state= 1)

print(X_train.shape)
print(X_test.shape)
print(y_train.shape)
print(y_test.shape)

(3207, 11)
(802, 11)
(3207,)
(802,)

from sklearn.linear_model import LinearRegression
# Create an instance of the LinearRegression class
reg = LinearRegression()

# Fit the model to the data
reg.fit(X_train, y_train)
score_LR = reg.score(X_test, y_test)

print(score_LR)

0.31915467345097737

# Print the coefficients and intercept of the model
print(reg.coef_)
print('Intercept: ', reg.intercept_)

[ 2.11754287e+02 -1.51749000e+00 -3.82413150e-01 -4.68630937e+03
 6.12280708e+01 1.86939042e+02 5.33937334e+00 6.73130085e+01
-2.26895985e+03 -3.41477166e+03 1.45479648e+02]
Intercept: 31951.619637876604

```

Double-click (or enter) to edit

```

y_pred = reg.predict(X_test)

import math
from sklearn.metrics import mean_absolute_error, mean_squared_error, r2_score

mae = mean_absolute_error(y_true=y_test, y_pred=y_pred)
#squared True returns MSE value, False returns RMSE value.
mse = mean_squared_error(y_true=y_test, y_pred=y_pred) #default=True
rmse = mean_squared_error(y_true=y_test, y_pred=y_pred, squared=False)
#rmse = math.sqrt(mse)

print("MAE:", mae)
print("MSE:", mse)
print("RMSE:", rmse)

MAE: 22922.87178300498
MSE: 1825516368.7970462
RMSE: 42726.0619387868

#Cross avlidation for Linear Regression

from sklearn.model_selection import KFold
from sklearn.model_selection import cross_val_score
from numpy import mean
from numpy import absolute
from numpy import sqrt

#Cross avlidation for Linear Regression
#define cross-validation method to use

```