In [1]:	Imports
	<pre>import pandas as pd import seaborn as sns from matplotlib import pyplot as plt</pre>
In [2]:	Loading and exploring the dataset 1. Load the dataset named carsales.csv and store it in a dataframe called raw_df. # Insert your code below # ====================================
In [3]:	<pre># ====================================</pre>
	<pre># ====================================</pre>
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	3 230.0 6.0 2.0 4 230.0 6.0 2.0 Market Categories Vehicle Size Vehicle Style \ O Factory Tuner, Luxury, High-Performance Compact Coupe Luxury, Performance Compact Convertible Luxury, High-Performance Compact Coupe
	Luxury, Performance Compact Coupe Luxury Compact Convertible Miles Per Gallon (MPG) Retail Price 1 19 46135 1 19 40650 2 20 36350 3 18 29450
In [27]:	1. Which columns in raw_df contains missing values? # Insert your code below # ===================================
	<pre>print(f"The columns which contain missing values are: \n{raw_df.isna().sum()}") The columns which contain missing values are: Unnamed: 0</pre>
	Fuel 3 Horse Power 69 Engine Cylinders 27 Number of Doors 6 Market Categories 2731 Vehicle Size 0 Vehicle Style 0
	Miles Per Gallon (MPG) 0 Retail Price 0 dtype: int64 4. Create a copy of raw_df named df . Remove any rows containing NaN values in the new dataframe. What is the shape of df before and after removing the NaN values? How many rows got removed? Use df in all following tasks unless otherwise is stated
In [5]:	<pre># Insert your code below # ====================================</pre>
	<pre>df = raw_df.dropna() # Using dropna() to remove all NaN print(f"\nThe shape after we drop all NaN: {df.shape}\n") # Shape after The shape before we drop all NaN: (10257, 13)</pre> The shape after we drop all NaN: (7441, 13)
In [6]:	<pre># Insert your code below # ====================================</pre>
In [7]:	<pre>print(f"There are {df.Brand.nunique()} unique brand values, while there are only {df.Fuel.nunique()} unique fu There are 47 unique brand values, while there are only 8 unique fuel values. 6. Which car brand has the most cars in the dataset? # Insert your code below</pre>
	# ====================================
	7. Find the average Retail Price per vehicle style. The output should be in the following format: Vehicle Style: [style], Average Price: [msrp] Vehicle Style: [style], Average Price: [msrp]
In [8]:	<pre>Vehicle Style: [style], Average Price: [msrp] # Insert your code below # ====================================</pre>
	<pre>print(f"\nBelow are the different vehicle styles and their respective mean prices:\n") for car, mean_price in grouped_cars.items(): print(f"Vechile Style: [{car}], Average Price: [{mean_price}]") # Printing out each vehicle style and the Below are the different vehicle styles and their respective mean prices:</pre>
	<pre>Vechile Style: [2dr Hatchback], Average Price: [22904.551136363636] Vechile Style: [2dr SUV], Average Price: [42031.1111111111] Vechile Style: [4dr Hatchback], Average Price: [23684.59713375796] Vechile Style: [4dr SUV], Average Price: [42588.92898272553] Vechile Style: [Cargo Minivan], Average Price: [22964.0] Vechile Style: [Cargo Van], Average Price: [30724.70588235294] Vechile Style: [Convertible], Average Price: [102362.01605136436]</pre>
	Vechile Style: [Convertible SUV], Average Price: [47975.0] Vechile Style: [Coupe], Average Price: [106314.17291414752] Vechile Style: [Crew Cab Pickup], Average Price: [39032.566860465115] Vechile Style: [Extended Cab Pickup], Average Price: [32238.53846153846] Vechile Style: [Passenger Minivan], Average Price: [29838.157894736843] Vechile Style: [Passenger Van], Average Price: [35963.15] Vechile Style: [Regular Cab Pickup], Average Price: [28136.792899408283]
	<pre>Vechile Style: [Sedan], Average Price: [56723.13185530922] Vechile Style: [Wagon], Average Price: [36176.68831168831] 8. Filter out non-gasoline cars. Remove rows where Fuel == electric or Fuel == diesel and print out the shape of the new dataframe. Save the results to df_gasoline</pre>
In [9]:	<pre># Insert your code below # ====================================</pre>
	Below are all the cars which are not electric nor diesel cars: Unnamed: 0 Brand Model Year Fuel \ 0 0 BMW 1 Series M 2011 premium unleaded (required) 1 1 BMW 1 Series 2011 premium unleaded (required) 2 2 BMW 1 Series 2011 premium unleaded (required) 3 3 BMW 1 Series 2011 premium unleaded (required) 4 BMW 1 Series 2011 premium unleaded (required) 5 premium unleaded (required) 6 premium unleaded (required)
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	Horse Power Engine Cylinders Number of Doors \ 0
	10253 300.0 6.0 4.0 10254 300.0 6.0 4.0 10255 300.0 6.0 4.0 10256 221.0 6.0 4.0 Market Categories Vehicle Size Vehicle Style \ 0 Factory Tuner, Luxury, High-Performance Compact Coupe
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	10255 Crossover, Hatchback, Luxury Midsize 4dr Hatchback 10256 Luxury Midsize Sedan Miles Per Gallon (MPG) Retail Price 0 19 46135 1 19 40650 2 20 36350
	3 18 29450 4 18 34500 10252 16 46120 10253 16 56670 10254 16 50620 10255 16 50920 10256 17 28995
	[7284 rows x 13 columns] C:\Users\Arterx\AppData\Local\Temp/ipykernel_4344/4068372095.py:5: UserWarning: Boolean Series key will be reindexed to match DataFrame index. df_gasoline = no_diesel[df.Fuel != "electric"] # New dataframe called df_gasoline where there are only gasoline cars present 9. Convert miles per galon to liters per 10 km.
	Add a new column to gasoline_df with the fuel consumption measured in liters fuel used per 10km driven. Save the results in a new column named liters_per_10km. Below is a function for converting miles per gallon to litres per 10km. Use this function to convert the values.
In [10]:	<pre>def mpg_to_liters_per_10km(mpg): """Returns miles per gallon converted to liters per 10km Args: mpg (int): Fuel efficiency measured in miles per gallon Returns:</pre>
In [11]:	float: Fuel efficency measured in liters fuel used per 10 km driven return 23.5 / mpg # Insert your code below # ====================================
	<pre>df_gasoline["liters_per_10km"] = mpg_to_liters_per_10km(df["Miles Per Gallon (MPG)"]) print(df_gasoline) Unnamed: 0</pre>
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