


▼ Importing the libraries

```
1 import numpy as np
2 import pandas as pd
3 from sklearn.impute import SimpleImputer
```

▼ Importing the dataset

```
1 dataset = pd.read_csv("car_ad.csv", encoding = "cp1252")
```

```
1 dataset.head()
```



	car	price	body	mileage	engV	engType	registration	year	model	di
0	Ford	15500.0	crossover	68	2.5	Gas	yes	2010	Kuga	
1	Mercedes-Benz	20500.0	sedan	173	1.8	Gas	yes	2011	E-Class	
2	Mercedes-Benz	35000.0	other	135	5.5	Petrol	yes	2008	CL 550	
3	Mercedes-	17000.0		160	1.8	Diesel		2010	B 160	

Double-click (or enter) to edit

▼ Dropping the drive values having NaN

```
1 dataset = dataset.dropna(subset=['drive'])
```

```
1 dataset.describe()
```

price mileage engV year



▼ Taking care of missing data

```
1 imputer = SimpleImputer(missing_values=np.nan, strategy='mean')
2 imputer.fit(dataset[['engV']])
3 dataset[['engV']] = imputer.transform(dataset[['engV']])
```

```
1 imputer = SimpleImputer(missing_values=0, strategy='mean')
2 imputer.fit(dataset[['mileage']])
3 dataset[['mileage']] = imputer.transform(dataset[['mileage']])
4 print(dataset)
5 imputer.fit(dataset[['price']])
6 dataset[['price']] = imputer.transform(dataset[['price']])
7 print(dataset[['price']])
```

	car	price	body	mileage	engV	engType	registration	\
0	Ford	15500.0	crossover	68.0	2.5	Gas	yes	
1	Mercedes-Benz	20500.0	sedan	173.0	1.8	Gas	yes	
2	Mercedes-Benz	35000.0	other	135.0	5.5	Petrol	yes	
3	Mercedes-Benz	17800.0	van	162.0	1.8	Diesel	yes	
5	Nissan	16600.0	crossover	83.0	2.0	Petrol	yes	
...	
9571	Hyundai	14500.0	crossover	140.0	2.0	Gas	yes	
9572	Volkswagen	2200.0	vagon	150.0	1.6	Petrol	yes	
9573	Mercedes-Benz	18500.0	crossover	180.0	3.5	Petrol	yes	
9574	Lexus	16999.0	sedan	150.0	3.5	Gas	yes	
9575	Audi	22500.0	other	71.0	3.6	Petrol	yes	

	year	model	drive
0	2010	Kuga	full
1	2011	E-Class	rear
2	2008	CL 550	rear
3	2012	B 180	front
5	2013	X-Trail	full
...
9571	2011	Tucson	front
9572	1986	Passat B2	front
9573	2008	ML 350	full
9574	2008	ES 350	front
9575	2007	Q7	full

[9065 rows x 10 columns]

	price
0	15500.0
1	20500.0
2	35000.0
3	17800.0
5	16600.0
...	...
9571	14500.0
9572	2200.0
9573	18500.0
9574	16999.0
9575	22500.0

[9065 rows x 1 columns]

```
1 print(dataset.describe(include='all'))
```

	car	price	body	mileage	engV	engType	\
count	9065	9065.000000	9065	9065.000000	9065.000000	9065	
unique	83	NaN	6	NaN	NaN	4	
top	Volkswagen	NaN	sedan	NaN	NaN	Petrol	
freq	879	NaN	3444	NaN	NaN	4181	
mean	NaN	16229.235289	NaN	144.568958	2.588607	NaN	
std	NaN	24202.479898	NaN	94.391987	5.318369	NaN	
min	NaN	259.350000	NaN	1.000000	0.100000	NaN	
25%	NaN	5500.000000	NaN	80.000000	1.600000	NaN	
50%	NaN	9900.000000	NaN	136.000000	2.000000	NaN	
75%	NaN	16800.000000	NaN	195.000000	2.588607	NaN	
max	NaN	547800.000000	NaN	999.000000	99.990000	NaN	

	registration	year	model	drive
count	9065	9065.000000	9065	9065
unique	2	NaN	863	3
top	yes	NaN	E-Class	front
freq	8542	NaN	185	5188
mean	NaN	2006.638941	NaN	NaN
std	NaN	7.001318	NaN	NaN
min	NaN	1953.000000	NaN	NaN
25%	NaN	2004.000000	NaN	NaN
50%	NaN	2008.000000	NaN	NaN
75%	NaN	2012.000000	NaN	NaN
max	NaN	2016.000000	NaN	NaN

▼ Encoding categorical data

```
1 from sklearn.preprocessing import LabelEncoder
2 LE_dataset = LabelEncoder()
3 dataset['car']= LE_dataset.fit_transform(dataset['car'])
4 dataset['body']= LE_dataset.fit_transform(dataset['body'])
5 dataset['engType']= LE_dataset.fit_transform(dataset['engType'])
6 dataset['registration']= LE_dataset.fit_transform(dataset['registration'])
7 dataset['registration']= LE_dataset.fit_transform(dataset['registration'])
8 dataset['drive']= LE_dataset.fit_transform(dataset['drive'])
9 print(dataset)
```

	car	price	body	mileage	engV	engType	registration	year	\
0	23	15500.0	0	68.0	2.5	1	1	2010	
1	50	20500.0	3	173.0	1.8	1	1	2011	
2	50	35000.0	2	135.0	5.5	3	1	2008	
3	50	17800.0	5	162.0	1.8	0	1	2012	
5	55	16600.0	0	83.0	2.0	3	1	2013	
...	
9571	33	14500.0	0	140.0	2.0	1	1	2011	
9572	77	2200.0	4	150.0	1.6	3	1	1986	
9573	50	18500.0	0	180.0	3.5	3	1	2008	
9574	43	16999.0	3	150.0	3.5	1	1	2008	

```

9575      4  22500.0      2      71.0   3.6          3          1  2007

      model  drive
0      Kuga      1
1    E-Class      2
2    CL 550      2
3    B 180      0
5    X-Trail      1
...      ...      ...
9571    Tucson      0
9572  Passat B2      0
9573    ML 350      1
9574    ES 350      0
9575      Q7      1

[9065 rows x 10 columns]

```

▼ Getting values in variables

```

1 x = dataset[['car', 'body', 'mileage', 'engV', 'engType', 'registration', 'year', 'model', 'dri
2 print(x)
3 y = dataset[['price']]
4 print(y)

```

```

      car  body  mileage  engV  engType  registration  year      model  drive
0     23     0     68.0   2.5        1          1  2010      Kuga      1
1     50     3    173.0   1.8        1          1  2011    E-Class      2
2     50     2    135.0   5.5        3          1  2008    CL 550      2
3     50     5    162.0   1.8        0          1  2012     B 180      0
5     55     0     83.0   2.0        3          1  2013    X-Trail      1
...     ...     ...     ...     ...      ...      ...     ...      ...
9571   33     0    140.0   2.0        1          1  2011     Tucson      0
9572   77     4    150.0   1.6        3          1  1986  Passat B2      0
9573   50     0    180.0   3.5        3          1  2008     ML 350      1
9574   43     3    150.0   3.5        1          1  2008     ES 350      0
9575    4     2     71.0   3.6        3          1  2007       Q7      1

[9065 rows x 9 columns]

      price
0    15500.0
1    20500.0
2    35000.0
3    17800.0
5    16600.0
...     ...
9571  14500.0
9572   2200.0
9573  18500.0
9574  16999.0
9575  22500.0

[9065 rows x 1 columns]

```

▼ Splitting into test and train

```

1 from sklearn.model_selection import train_test_split
2 x_train, x_test, y_train, y_test = train_test_split(x, y, test_size = 0.2, random_state
3 print(x_test)
4 print(x_train)
5 print(y_train)
6 print(y_test)

```

	car	body	mileage	engV	engType	registration	year	\
6159	59	2	1.000000	1.5	0	1	2007	
2959	39	3	73.000000	1.6	3	1	2008	
2474	74	3	170.000000	3.0	3	1	2004	
5256	4	3	370.000000	2.6	2	1	1997	
8328	56	1	160.000000	1.6	3	1	2005	
...
2119	70	4	144.568958	2.0	0	1	2016	
1028	74	0	144.568958	4.5	0	1	2016	
6161	12	3	103.000000	1.5	3	1	2005	
7144	76	3	67.000000	1.6	3	1	2012	
762	59	3	310.000000	1.4	3	1	2003	

	model	drive
6159	Kangoo ãđóç.	0
2959	Cerato	0
2474	Camry	0
5256	A6	0
8328	Astra H	0
...
2119	Outback	1
1028	Land Cruiser 200	1
6161	Aveo	0
7144	2115	0
762	Symbol	0

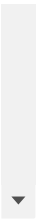
[1813 rows x 9 columns]

	car	body	mileage	engV	engType	registration	year	model	\
2967	23	1	71.0	1.0	3	1	2013	Fiesta	
9494	23	1	70.0	1.6	3	1	2006	Focus	
1605	58	0	115.0	4.8	3	1	2010	Cayenne	
9221	59	4	100.0	1.5	0	1	2009	Kangoo ïàññ.	
5785	55	5	90.0	1.6	1	1	2008	Note	
...
3046	5	0	200.0	3.0	0	1	2009	X5	
8243	56	4	290.0	1.4	3	1	1994	Astra F	
947	77	3	87.0	1.6	3	1	2011	Polo	
5476	33	0	90.0	2.7	1	1	2008	Santa FE	
248	67	3	109.0	1.8	3	1	2012	Octavia A5	

	drive
2967	0
9494	0
1605	1
9221	0
5785	0
...	...
3046	1
8243	0
947	0
5476	1

```
248      0

[7252 rows x 9 columns]
      price
2967 11500.000000
9494  7100.000000
```



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