Date	7 November 2022
Team ID	PNT2022TMIDO1748
Project Name	PROJECT-CAR RESALES VALUE PREDICTION
Maximum Marks	2 Marks

Collect dataset:

Machine Learning has become a tool used in almost every task that requires estimation. So we need to build a model to estimate the price of used cars. The model should take car-related parameters and output a selling price. On sprint-1 the selling price of a used car depends on certain features datasets are collected from different open sources like kaggle.com, data.gov, UCI machine learning repository, the dataset which contains a set of features through which the resale price of the car can be identified is to be collected as

- seller
- offerType
- price
- vehicleType
- yearOfRegistration
- gearbox
- powerPS
- model
- kilometer
- monthOfRegistration
- fuelType
- brand
- notRepairedDamage

ML is a data hunger technology, it depends heavily on data, without data, it is impossible. It is the most crucial aspect that makes algorithm training possible. Collects Data, Import necessary packages, Pre-process images, and passes on to Network Model and Saves Model Weights. The libraries can be imported,

import pandas as pd import numpy as np import matplotlib a from sklearm.prepro import pickle																			
<pre>df = pd.read_csv("/ df.head()</pre>	content/drive/MyOrive/Colab No	otebooks/	nutos.csv")																
dateCrauled	nane	seller	offerType	price a	btest	vehicleType	yearOfRegistration	gearbox	powerPS	nodel	kilometer	monthOfRegistration	fuelType	brand	notRepairedDamage	dateCreated	nrOfPictures	postalCode	lastSeen
dateCrawled 0 24-03-2016 11:52	name Got_3_1.6		offerType Angebot	price a	btest	vehicleType NaN	yearOfRegistration			model golf				brand volkswagen	- 4 - 5	dateCreated 24-03-2016 00:00			1astSeen 07-04-2016 03:16
		privat						manuell	0					volkswagen	NaN		0.00	70435	
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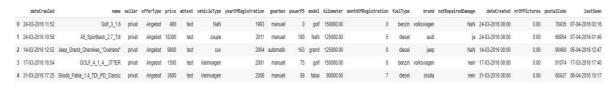
Pre-Process The Data:

Pre-processing the dataset that includes:

- Handling the null values.
- Handling the categorical values if any.
- Normalize the data if required.
- Identify the dependent and independent variables.

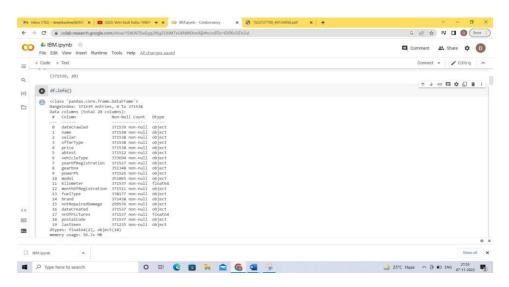
Data cleaning and wrangling methods are applied on the *used cars* data file. Before making data cleaning, some explorations and data visualizations were applied on data set. This gave some idea and guide about how to deal with missing values and extreme values. After data cleaning, data exploration was applied again in order to understand cleaned version of the data.

df = pd.read_csv("/content/drive/MyDrive/Colab Notebooks/autos.csv")
df.head()



print(df.shape)
(371539, 20)

df.info()



```
df['powerPS'].unique()
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                  df[powerPs].unique()

array(['0', '190', '163', '75', '69', '102', '109', '50', '125', '101', '105', '140', '115', '131', '60', '136', '160', '231', '90', '118', '193', '99', '113', '218', '122', '129', '70', '306', '95', '61', '177', '80', '170', '55', '143', '64', '286', '232', '150', '156', '82', '204', '155', '54', '185', '87', '180', '86', '84', '303', '224', '235', '200', '178', '265', '77', '110', '144', '120', '164', '184', '126', '88', '194', '305', '197', '179', '250', '45', '313', '41', '165', '98', '130', '114', '211', '56', '326', '201', '213', '58', '197', '83', '174', '100', '226', '85', '73', '192', '68', '66', '299', '74', '52', '510', '147', '65', '310', '71', '97', '239', '295', '233', '5', '300, '103', '245', '258', '292', '320', '63', '81', '148', '334', '44', '145', '230', '288', '266', '457', '104', '400', '188', '333', '186', '117', '141', '55', '132', '244', '158', '39', '272', '92', '51', '135', '53', '455', '26', '37', '30', '104', '400', '188', '333', '186', '117', '141', '55', '26', '273', '131', '141', '56', '131', '141', '56', '131', '141', '56', '131', '141', '56', '131', '141', '56', '131', '141', '56', '131', '141', '56', '131', '141', '56', '131', '141', '56', '131', '141', '56', '131', '141', '56', '131', '141', '56', '131', '141', '56', '131', '141', '56', '131', '141', '156', '131', '141', '156', '131', '141', '156', '131', '141', '156', '131', '141', '156', '131', '141', '156', '131', '141', '156', '131', '141', '156', '131', '141', '156', '131', '141', '156', '131', '141', '156', '131', '141', '156', '131', '141', '156', '131', '141', '156', '131', '141', '156', '131', '141', '156', '131', '141', '156', '131', '141', '156', '131', '141', '156', '131', '141', '156', '131', '141', '156', '131', '141', '156', '131', '141', '156', '131', '141', '156', '131', '141', '156', '131', '141', '156', '131', '141', '156', '156', '156', '156', '156', '156', '156', '156', '156', '156', '156', '156', '156', '156', '156', '156', '156', '156', '156', '156', '156', '156', '156', '156', '156
            o df['powerPS'].unique()
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df=df[df['powerPS'].str.isnumeric().fillna(False)]
print(df.seller.value counts())
df[df.seller != 'gewerblich']
print(df.offerType.value counts())
df[df.offerType != 'Gesuch']
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df['powerPS']=df['powerPS'].astype(int)
df=df[(df.powerPS > 50) & (df.powerPS < 900)]
print(df.shape)
df=df[df['yearOfRegistration'].str.isnumeric().fillna(False)]
```

```
df['yearOfRegistration']=df['yearOfRegistration'].astype(int)
df=df[(df.yearOfRegistration > 1950) & (df.yearOfRegistration < 2017)]</pre>
print (df.shape)
df.drop(['name', 'abtest', 'dateCrawled', 'nrOfPictures', 'lastSeen', '
postalCode', 'dateCreated'], axis='columns', inplace=True)
df.info()
  <class 'pandas.core.frame.DataFrame'>
  Int64Index: 308923 entries, 1 to 371538
  Data columns (total 13 columns):
       Column
                              Non-Null Count
                                                 Dtype
       -----
  ---
       seller
                              308923 non-null
                                                object
   0
     offerType
                              308923 non-null
                                                object
   1
                                                object
   2
      price
                              308923 non-null
   3
      vehicleType
                              297510 non-null
                                                object
                                                int64
      yearOfRegistration
                              308923 non-null
   4
   5
      gearbox
                              303629 non-null
                                                object
   6
      powerPS
                              308923 non-null int64
   7
      model
                              297134 non-null
                                                object
                              308923 non-null float64
      kilometer
     monthOfRegistration 308923 non-null
                                                obiect
   10 fuelType
                              293046 non-null
                                                object
   11 brand
                              308923 non-null
                                                object
   12 notRepairedDamage
                              265507 non-null
                                                object
  dtypes: float64(1), int64(2), object(10)
  memory usage: 33.0+ MB
new df=df.copy()
new df = new df.drop duplicates(['price', 'vehicleType', 'yearOfRegistr
'gearbox', 'powerPS', 'model', 'kilometer', 'monthOfRegistration', 'fue
lType',
'notRepairedDamage'])
new df.gearbox.replace(('manuell', 'automatik'), ('manual', 'automatic'
), inplace=True)
new df.fuelType.replace(('benzin', 'andere', 'elektro'), ('petrol', 'ot
hers', 'electric'), inplace=True)
new df.notRepairedDamage.replace(('ja', 'nein'),('Yes', 'No'), inplace=
True)
new df.vehicleType.replace(('kleinwagen', 'cabrio', 'kombi', 'andere'),
 ('small car', 'convertible', 'combination', 'others'), inplace=True)
```

```
new df['price'].unique()
array(['18300', '9800', '1500', ..., '18429', '24895', '10985'],
      dtype=object)
new df['price'] = new df['price'].astype(int)
new df = new df[(new df.price \geq 100) & (new df.price \leq 150000)]
new df['fuelType'].fillna (value='not-declared', inplace=True)
new df['gearbox'].fillna (value='not-declared', inplace=True)
new df['notRepairedDamage'].fillna (value='not-declared', inplace=True)
new df[ 'vehicleType'].fillna (value='not-declared', inplace=True)
new df['model'].fillna (value='not-declared', inplace=True)
new df['kilometer']=new df['kilometer'].astype(int)
new df.info()
<class 'pandas.core.frame.DataFrame'>
Int64Index: 278363 entries, 1 to 371538
Data columns (total 13 columns):
                            Non-Null Count
     Column
                                             Dtype
     -----
                               ------
_ _ _
 0
     seller
                            278363 non-null object
 1 offerType
                            278363 non-null object
                           278363 non-null int64
 2
    price
 3
     vehicleType
                            278363 non-null object
    yearOfRegistration 278363 non-null int64
 4
                            278363 non-null object
 5
     gearbox
     powerPS
                            278363 non-null int64
 6
 7
                            278363 non-null object
     model
     kilometer
                            278363 non-null int64
 8
     monthOfRegistration 278363 non-null object
 10 fuelType
                            278363 non-null object
 11 brand
                            278363 non-null object
 12 notRepairedDamage
                           278363 non-null object
dtypes: int64(4), object(9)
memory usage: 29.7+ MB
```

new df['price'].unique()

new_df.head()

new_df.head()

	seller	offerType	price	vehicleType	yearOfRegistration	gearbox	powerPS	model	kilometer	monthOfRegistration	fuelType	brand	notRepairedDamage
1	privat	Angebot	18300	coupe	2011	manual	190	not-declared	125000	5	diesel	audi	Yes
2	privat	Angebot	9800	suv	2004	automatic	163	grand	125000	8	diesel	jeep	not-declared
3	privat	Angebot	1500	small car	2001	manual	75	golf	150000	6	petrol	volkswagen	No
4	privat	Angebot	3600	small car	2008	manual	69	fabia	90000	7	diesel	skoda	No
5	privat	Angebot	650	limousine	1995	manual	102	3er	150000	10	petrol	bmw	Yes