| Date | 7 November 2022 |
|---------------|--------------------------------------|
| Team ID | PNT2022TMID01544 |
| 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 po import numpy as np import matplotlib a from sklearn.prepro import pickle | | | | | | | | | | | | | | | | | | | |
|--|--|----------------------------|----------------------|----------------------|----------------|--------------------|--------------------|---------------------------------|--------------|----------------------|-------------------------------------|-------------------------------|----------------------------|---------------------|-------------------|--------------------------------------|--------------|-------------------------|--------------------------------------|
| df = pd.read_csv("; df.head() | /content/drive/MyDrive/Colab No | stebooks/s | nutos.csv") | | | | | | | | | | | | | | | | |
| dateCrawled | name | seller | offerType | price : | abtest | vehicleType | yearOfRegistration | gearbox | powerPS | nodel | kilometer | monthOfRegistration | fuelType | brand | notRepairedDamage | dateCreated | nrOfPictures | postalCode | lastSeen |
| | | | offerType Angebot | | abtest test | vehicleType NaN | yearOfRegistration | | powerPS 0 | | kilometer 150000.00 | monthOfRegistration | | brand volkswagen | | dateCreated 24-03-2016 00:00 | nrOfPictures | | 1astSeen 07-04-2016 03:16 |
| dateCrawled | Goff_3_1.6 | privat | Angebot | 480 | | | | manuell | | golf | 150000.00 | monthOfRegistration 0 5 | | | NaN | | | 70435 | |
| dateCrawled 0 24-03-2016 11:52 1 24-03-2016 10:58 | Goff_3_1.6 | privat privat | Angebot | 480 18300 | test | NaN | 1993 2011 | manuell | 190 | golf NaN | 150000.00 | monthOfRegistration 0 5 | benzin | volkswagen | NaN ja | 24-03-2016 00:00 | 0.00 | 70435 66954 | 07-04-2016 03:16 07-04-2016 01:46 |
| dateCrawled 0 24-03-2016 11:52 1 24-03-2016 10:58 | Golf_3_1.6 A5_Sportback_2.7_Tdi Jeep_Grand_Cherokee_"Overland" | privat privat privat | Angebot Angebot | 480 18300 9800 | test | NaN coupe | 1993 2011 | manuell manuell automatik | 190 | golf NaN grand | 150000.00 125000.00 125000.00 | monthOfRegistration 0 5 8 | benzin diesel diesel | volkswagen audi | NaN ja NaN | 24-03-2016 00:00 24-03-2016 00:00 | 0.00 | 70435 66954 90480 | 07-04-2016 03:16 |

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.

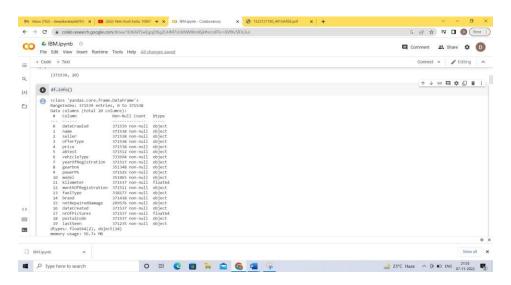
Data cleaning and wrangling methods are applied on the *used cars* data file. Before making data cleaning, some explorations and data visualizations were applied on data set. This gave some idea and guide about how to deal with missing values and extreme values. After data cleaning, data exploration 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', '139', '99', '113', '218', '122', '129', '70', '366', '98', '61', '177', '80', '170', '55', '143', '64', '286', '232', '150', '156', '82', '264', '155', '54', '188', '87', '180', '86', '84', '303', '224', '235', '200', '178', '265', '77', '110', '144', '120', '164', '164', '165', '88', '194', '305', '197', '179', '256', '45', '313', '41', '165', '98', '130', '114', '211', '56', '326', '201', '213', '58', '107', '83', '174', '100', '220', '68', '66', '299', '74', '52', '510', '147', '65', '310', '71', '97', '239', '295', '233', '5', '300', '103', '245', '258', '292', '320', '63', '81', '148', '354', '44', '145', '230', '280', '260', '457', '104', '400', '188', '333', '186', '117', '141', '56', '132', '224', '158', '39', '727', '92', '51', '135', '53', '435', '260', '43', '146', '66', '710', '166', '716', '64', '716', '714', '713', '714', '713', '714', '713', '714', '713', '714', '714', '713', '714', '714', '713', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714', '714
           odf['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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      • studf[(df.poserPS + 50) & (df.poserPS + 500)]
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                                                                                                                                                                                                                                         Type here to search
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=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
   0
                              308923 non-null
                                                object
     offerType
                             308923 non-null
                                                object
   1
                              308923 non-null
                                                object
   2
      price
   3
      vehicleType
                              297510 non-null
                                                object
      yearOfRegistration
                             308923 non-null int64
   4
   5
      gearbox
                              303629 non-null
                                                object
   6
      powerPS
                              308923 non-null int64
   7
      model
                              297134 non-null
                                                object
      kilometer
                             308923 non-null float64
      monthOfRegistration 308923 non-null
                                                object
   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=
new df.vehicleType.replace(('kleinwagen', 'cabrio', 'kombi', 'andere'),
 ('small car', 'convertible', 'combination', 'others'), inplace=True)
```

df['yearOfRegistration']=df['yearOfRegistration'].astype(int)

```
new df['price'].unique()
 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):
     Column
                            Non-Null Count
                                             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
 5
     gearbox
                            278363 non-null object
                            278363 non-null int64
 6
     powerPS
                            278363 non-null object
 7
     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.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 |