Date	7 November 2022
Team ID	PNT2022TMID40489
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,

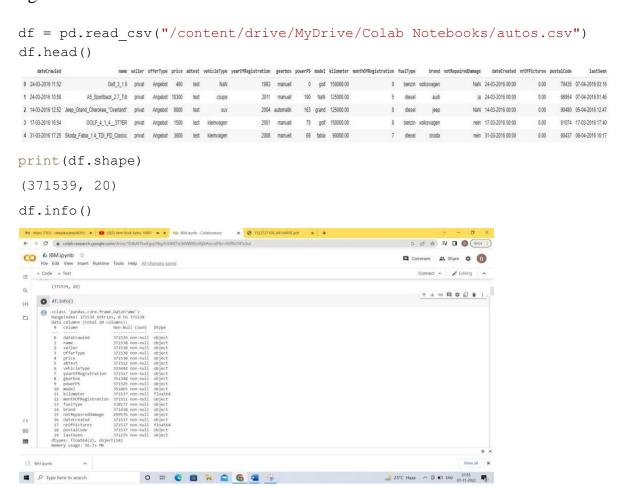
import pandas as po import numpy as np import matplotlib a from sklearn.prepro import pickle																			
<pre>df * pd.read_csv("/ df.head() dateCracked</pre>	content/drive/MyOrive/Colab No							Anna Maria						12000					
			affertype					geartox	powerPS			monthOfRegistration	fuelType	brand	notRepairedDamage		mrOfPictures		lastSeen
0 24-03-2016 11:52	Gof_3_1.6	privat	Angebot	480	test	NaN	1993	manuell	0	golf	150000.00	0	benzin	volkswagen	NaN	24-03-2016 00:00	0.00	70435	07-04-2016 03:16
1 24-03-2016 10:58	A5_Sportback_2.7_Tdl	privat	Angebot	18300	test	coupe	2011	manuell	190	NaN	125000.00	5	desel	audi	ja	24-03-2016 00:00	0.00	66954	07-04-2016 01:46
				-	See a		2004	automatik	163	grand	125000.00		diesel	jeep	NAM	14-03-2016 00:00	0.00	90480	05-04-2016 12:47
2 14-03-2016 12:52	Jeep_Grand_Cherokee_"Overland"	privat	Angebot	9600	test	SUV	2004												
2 14-03-2016 12:52 3 17-03-2016 16:54	Jeep_Grand_Cherokee_"Overland" GOLF_4_1_43T7ER		Angebot	1500	test	kleiniragen	2001	manueli	75			6	benzin	volkswagen		17-03-2016 00:00			17-03-2016 17:40

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.





'gewerblich'] print(df.offerType.value counts())

df[df.offerType != 'Gesuch']

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                         O # C # 6 4 9
df['powerPS']=df['powerPS'].astype(int)
df=df[(df.powerPS > 50) & (df.powerPS < 900)]</pre>
print(df.shape)
df=df[df['yearOfRegistration'].str.isnumeric().fillna(False)]
df['yearOfReqistration']=df['yearOfReqistration'].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):
                             Non-Null Count
       Column
                                               Dtype
  _____
   0
      seller
                             308923 non-null
                                               object
     offerType
                             308923 non-null
                                               object
   1
   2
      price
                             308923 non-null
                                               object
   3
      vehicleType
                             297510 non-null
                                               object
   4
      yearOfRegistration
                             308923 non-null
                                               int64
   5
      gearbox
                             303629 non-null
                                               object
                             308923 non-null
                                               int64
   6
      powerPS
   7
                                               object
      model
                             297134 non-null
      kilometer
                             308923 non-null float64
      monthOfRegistration 308923 non-null
                                               object
   9
   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
ation',
'gearbox', 'powerPS', 'model', 'kilometer', 'monthOfRegistration', 'fue
'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()
 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 >= 100) & (new_df.price <= 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
     _____
     seller
                           278363 non-null object
 0
 1
   offerType
                           278363 non-null object
                           278363 non-null int64
 2
     price
                           278363 non-null object
 3
     vehicleType
     yearOfRegistration 278363 non-null int64
 4
 5
     gearbox
                           278363 non-null object
     powerPS
 6
                           278363 non-null int64
 7
     model
                           278363 non-null object
     kilometer
 8
                           278363 non-null int64
     monthOfRegistration 278363 non-null object
                           278363 non-null object
 10 fuelType
     brand
                           278363 non-null object
 11
 12 notRepairedDamage
                           278363 non-null object
dtypes: int64(4), object(9)
memory usage: 29.7+ MB
```

new df.head()

nev	v_df.hea	d()											
	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	Angehot	650	limousine	1995	manual	102	3er	150000	10	netrol	hmw	Vec