

CAR RESALE VALUE PREDICTION

TEAM ID: PNT2022TMID12591

```
In [141]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import pickle
from sklearn.preprocessing import LabelEncoder
from sklearn.model_selection import train_test_split, cross_val_score
from sklearn.ensemble import RandomForestRegressor
from sklearn.metrics import r2_score
```

```
In [142]: df=pd.read_csv('autos.csv',header=0,sep=',',encoding='Latin1')
```

c:\New folder\lib\site-packages\IPython\core\interactiveshell.py:3340: DtypeWarning: Columns (11) have mixed types.Specify dtype option on import or set low_memory=False.
exec(code_obj, self.user_global_ns, self.user_ns)

```
In [143]: df.head()
```

```
Out[143]:
```

	dateCrawled		name	seller	offerType	price	abtest	vehicleType	year
0	24-03-2016 11:52		Golf_3_1.6	privat	Angebot	480.0	test	NaN	
1	24-03-2016 10:58	A5_Sportback_2.7_Tdi		privat	Angebot	18300.0	test	coupe	
2	14-03-2016 12:52	Jeep_Grand_Cherokee_"Overland"		privat	Angebot	9800.0	test	suv	
3	17-03-2016 16:54	GOLF_4_1_4__3TÜRER		privat	Angebot	1500.0	test	kleinwagen	
4	31-03-2016 17:25	Skoda_Fabia_1.4_TDI_PD_Classic		privat	Angebot	3600.0	test	kleinwagen	

In [147]: `df.head()`

Out[147]:

	price	vehicleType	yearOfRegistration	gearbox	powerPS	model	kilometer	monthOfRegistration
0	480.0	NaN	1993.0	manuell	0.0	golf	150000	
1	18300.0	coupe	2011.0	manuell	190.0	NaN	125000	
2	9800.0	suv	2004.0	automatik	163.0	grand	125000	
3	1500.0	kleinwagen	2001.0	manuell	75.0	golf	150000	
4	3600.0	kleinwagen	2008.0	manuell	69.0	fabia	90000	

In [148]: `df.shape`

Out[148]: (371539, 11)

In [149]: `df=df[(df['powerPS']>50) & (df['powerPS']<90)]`

In [150]: `df=df[(df['yearOfRegistration']>1949) & (df['yearOfRegistration']<2017)]`
`df=df[(df['price']>=100) & (df['price']<=150000)]`

In [151]: `df.shape`

Out[151]: (78684, 11)

In [152]: `df.columns`

Out[152]: Index(['price', 'vehicleType', 'yearOfRegistration', 'gearbox', 'powerPS',
'model', 'kilometer', 'monthOfRegistration', 'fuelType', 'brand',
'notRepairedDamage'],
dtype='object')

In [153]: `new_df=df.copy()`
`new_df=new_df.drop_duplicates(['price', 'vehicleType', 'yearOfRegistration', 'gearbox'])`

In [154]: `new_df['gearbox'].replace(('manuell','automatik'),('manual','automatic'),inplace=True)`
`new_df['fuelType'].replace(('benzin','andere','elektro'),('petrol','others','electric'),inplace=True)`
`new_df['vehicleType'].replace(('kleinwagen','cabrio','kombi','andere'),('small car','suv','coupe','sedan'),inplace=True)`
`new_df['notRepairedDamage'].replace(('ja','nein'),('yes','no'),inplace=True)`

In [155]: `new_df.to_csv('autos_preprocessed.csv')`

```
In [156]: labels=['gearbox','notRepairedDamage','model','brand','fuelType','vehicleType']
```

```
In [157]: dicti={}
for feature in labels:
    dicti[feature]=LabelEncoder()
    dicti[feature].fit(new_df[feature])
    temp=dicti[feature].transform(new_df[feature])
    np.save(str('classes'+feature+'.npy'),dicti[feature].classes_)
    print(feature,':',dicti[feature])
    new_df.loc[:,feature+'_Labels']=pd.Series(temp,index=new_df.index)
labeled=new_df[['price','yearOfRegistration','powerPS','kilometer','monthOfRegistration']]

gearbox : LabelEncoder()
notRepairedDamage : LabelEncoder()
model : LabelEncoder()
brand : LabelEncoder()
fuelType : LabelEncoder()
vehicleType : LabelEncoder()
```

```
In [158]: labeled.columns
```

```
Out[158]: Index(['price', 'yearOfRegistration', 'powerPS', 'kilometer',
               'monthOfRegistration', 'gearbox_Labels', 'notRepairedDamage_Labels',
               'model_Labels', 'brand_Labels', 'fuelType_Labels',
               'vehicleType_Labels'],
              dtype='object')
```

```
In [159]: X=labeled.iloc[:,1:].values
Y=labeled.iloc[:,0].values
```

```
In [160]: Y=Y.reshape(-1,1)
```

```
In [161]: X_train,X_test,y_train,y_test=train_test_split(X,Y,test_size=.3)
```

```
In [162]: regressor=RandomForestRegressor(n_estimators=1000,max_depth=10)
regressor.fit(X_train,np.ravel(y_train,order='C'))
```

```
Out[162]: RandomForestRegressor(max_depth=10, n_estimators=1000)
```

```
In [163]: prediction=regressor.predict(X_test)
```

```
In [164]: r2_score(y_test,prediction)
```

```
Out[164]: 0.747891214137351
```

```
In [165]: file='model.sav'
pickle.dump(regressor,open(file,'wb'))
```

```
In [172]: from sklearn.linear_model import LogisticRegression
```

```
In [173]: lr = LogisticRegression()
```

```
In [174]: lr.fit(X_train, y_train)
```

```
c:\New folder\lib\site-packages\sklearn\utils\validation.py:993: DataConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n_samples, ), for example using ravel().
```

```
y = column_or_1d(y, warn=True)
```

```
c:\New folder\lib\site-packages\sklearn\linear_model\_logistic.py:814: ConvergenceWarning: lbfgs failed to converge (status=1):
STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
```

Increase the number of iterations (max_iter) or scale the data as shown in:

<https://scikit-learn.org/stable/modules/preprocessing.html> (<https://scikit-learn.org/stable/modules/preprocessing.html>)

Please also refer to the documentation for alternative solver options:

https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression (https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression)

```
n_iter_i = _check_optimize_result(
```

```
Out[174]: LogisticRegression()
```

```
In [175]: pred = lr.predict(X_test)
```

```
In [176]: r2_score(y_test, pred)
```

```
Out[176]: -0.5266927414549378
```

```
In [ ]:
```