

INTRODUCTION AND OVERVIEW

- Machine learning is an application of artificial intelligence(AI) that provides systems the ability to automatically learn and improve from experience without being explicitly programmed.
- Determining whether the listed price of a used car is a challenging task, due to the many factors that drive a used vehicle's price on the market. The focus of this project is developing machine learning models that can accurately predict the price of a used car based on its features, in order to make informed purchases.

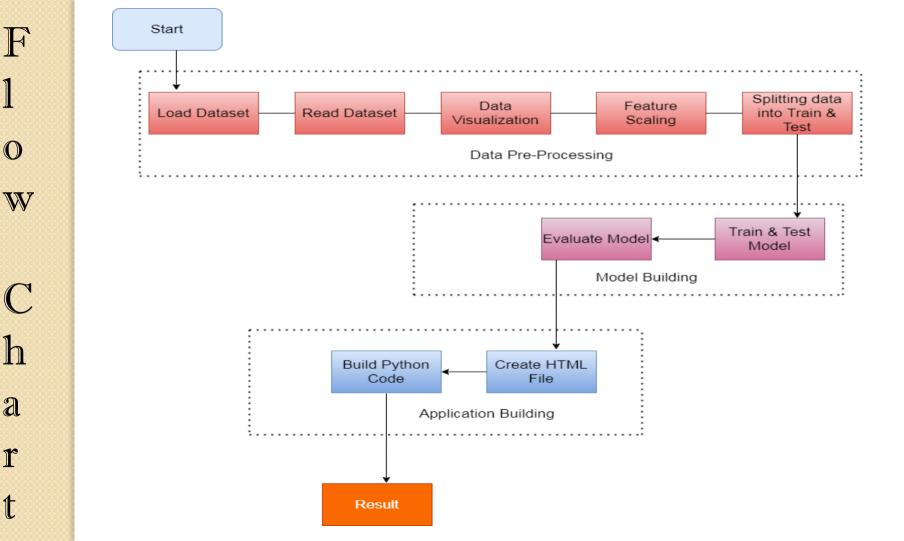
- We implement and evaluate various learning methods on a dataset consisting of the sale prices of different makes and models. Our results show that Random Forest model regression yield the best results.
- Multiple regression algorithms like Simple Linear, Multi Linear, Polynomial, Decision Tree, Random Forest were compared and evaluated based on their accuracy to select the best vehicle.

PROJECT DESCRIPTION:

We investigate the application of supervised machine learning techniques to predict the price of used cars. The predictions are based on the data. Different techniques like simple linear regression, multi linear regression, polynomial regression, random tree regression and decision trees have been used to make the predictions. The predictions are then evaluated and compared in order to find those which provide the best performances.

PROPOSED SOLUTION:

In this Project, we discuss the results obtained from applying our selected regressor on the pre-processsed dataset.. All supervised machine learning algorithms applied in this study have been developed using the open-source, object-oriented programming language Python 3.0 and its many packages. The cross-validated accuracy score and cross-validated AUC scores for all the regression algorithms we used on the vehicle resale data.



Software Design:

- Jupyter Notebook Environment
- Spyder Ide
- Machine learning algorithms.
- Python (pandas, numpy, matplotlib, seaborn, sklearn)
- HTML
- Flask

EXPERIMENTAL INVESTIGATION:

| dateCrawl | name | seller | offerType | e price | abtest | vehicleTyp ye | earOfReg | gearbox | powerPS | model | kilometer | monthOfR | fuelType | brand | notRepaire | dateCreaten | rOfPictur | postalCod | lastSeen |
|-----------|-------------|----------|-----------|---------|---------|---------------|----------|-----------|---------|----------|-----------|----------|----------|-----------|------------|-------------|-----------|-----------|----------|
| нинанина | Golf_3_1. | f privat | Angebot | 480 | test | | 1993 | manuell | 0 | golf | 150000 | 0 | benzin | volkswage | n | ******* | 0 | 70435 | ####### |
| ******* | A5_Sportb | privat | Angebot | 18300 | test | coupe | 2011 | manuell | 190 | | 125000 | 5 | diesel | audi | ja | ******* | 0 | 66954 | ####### |
| ******* | Jeep_Gran | privat | Angebot | 9800 | test | suv | 2004 | automatik | 163 | grand | 125000 | 8 | diesel | jeep | | ******* | 0 | 90480 | ####### |
| ******* | GOLF_4_1 | privat | Angebot | 1500 | test | kleinwager | 2001 | manuell | 75 | golf | 150000 | 6 | benzin | volkswage | nein | ******* | 0 | 91074 | ####### |
| ******* | Skoda_Fal | privat | Angebot | 3600 | test | kleinwager | 2008 | manuell | 69 | fabia | 90000 | 7 | diesel | skoda | nein | ******* | 0 | 60437 | ####### |
| ******* | BMW_316 | privat | Angebot | 650 | test | limousine | 1995 | manuell | 102 | 3er | 150000 | 10 | benzin | bmw | ja | ****** | 0 | 33775 | ####### |
| ******* | Peugeot_2 | 2 privat | Angebot | 2200 | test | cabrio | 2004 | manuell | 109 | 2_reihe | 150000 | 8 | benzin | peugeot | nein | ******* | 0 | 67112 | ####### |
| ******* | VW_Derb | y privat | Angebot | 0 | test | limousine | 1980 | manuell | 50 | andere | 40000 | 7 | benzin | volkswage | nein | ****** | 0 | 19348 | ####### |
| ******* | Ford_C_ | privat | Angebot | 14500 | control | bus | 2014 | manuell | 125 | c_max | 30000 | 8 | benzin | ford | | ****** | 0 | 94505 | ####### |
| ******* | VW_Golf_ | privat | Angebot | 999 | test | kleinwager | 1998 | manuell | 101 | golf | 150000 | 0 | | volkswage | n | ******* | 0 | 27472 | ####### |
| ******* | Mazda_3 | privat | Angebot | 2000 | control | limousine | 2004 | manuell | 105 | 3_reihe | 150000 | 12 | benzin | mazda | nein | ****** | 0 | 96224 | ####### |
| ******* | Volkswage | privat | Angebot | 2799 | control | kombi | 2005 | manuell | 140 | passat | 150000 | 12 | diesel | volkswage | ja | ******* | 0 | 57290 | ####### |
| ******** | VW_Passa | privat | Angebot | 999 | control | kombi | 1995 | manuell | 115 | passat | 150000 | 11 | benzin | volkswage | n | ****** | 0 | 37269 | ####### |
| ******** | VW_PASS | privat | Angebot | 2500 | control | kombi | 2004 | manuell | 131 | passat | 150000 | 2 | | volkswage | nein | ****** | 0 | 90762 | ####### |
| ******** | Nissan_Na | privat | Angebot | 17999 | control | suv | 2011 | manuell | 190 | navara | 70000 | 3 | diesel | nissan | nein | ****** | 0 | 4177 | ****** |
| ******** | KA_Luftha | privat | Angebot | 450 | test | kleinwager | 1910 | | 0 | ka | 5000 | 0 | benzin | ford | | ******* | 0 | 24148 | ####### |
| ******** | Polo_6n_3 | 1 privat | Angebot | 300 | test | | 2016 | | 60 | polo | 150000 | 0 | benzin | volkswage | n | ******* | 0 | 38871 | ****** |
| ******** | Renault_T | privat | Angebot | 1750 | control | kleinwager | 2004 | automatik | 75 | twingo | 150000 | 2 | benzin | renault | nein | ****** | 0 | 65599 | ***** |
| | Ford_C_N | l privat | Angebot | 7550 | test | bus | 2007 | manuell | 136 | c_max | 150000 | 6 | diesel | ford | nein | ******* | 0 | 88361 | ****** |
| ******** | Mercedes | privat | Angebot | 1850 | test | bus | 2004 | manuell | 102 | a_klasse | 150000 | 1 | benzin | mercedes | nein | ******* | 0 | 49565 | ****** |
| ******** | Volkswage | privat | Angebot | 10400 | control | coupe | 2009 | manuell | 160 | scirocco | 100000 | 4 | benzin | volkswage | nein | ******* | 0 | 75365 | ***** |
| ******** | BMW_530 | privat | Angebot | 3699 | test | limousine | 2002 | automatik | 231 | 5er | 150000 | 7 | benzin | bmw | nein | ******* | 0 | 68309 | ****** |
| ******** | Opel_Mer | i privat | Angebot | 2900 | test | | 2018 | manuell | 90 | meriva | 150000 | 5 | benzin | opel | nein | ******* | 0 | 49716 | ****** |
| | Stadtflitze | privat | Angebot | 450 | test | kleinwager | 1997 | manuell | 50 | arosa | 150000 | 5 | benzin | seat | nein | ******* | 0 | 9526 | ****** |
| ******** | MERCEDE | privat | Angebot | 500 | test | limousine | 1990 | manuell | 118 | andere | 150000 | 10 | benzin | mercedes | ja | ******* | 0 | 35390 | ***** |
| ********* | BMW 530 | orivat | Angebot | 2500 | control | kombi | 2002 | automatik | 193 | 5er | 150000 | 9 | diesel | bmw | ja | ******* | 0 | 73765 | ******* |

HEAT MAP:



Model Building and Evaluation:

Here random forest regression is used to predict. Accuracy is evaluated.

```
In [51]: # #random forest regression
            from sklearn.ensemble import RandomForestRegressor
            rf=RandomForestRegressor(n estimators=100,criterion='mse',random state=0)
In [52]: M rf.fit(x train,y train)
   Out[52]: RandomForestRegressor(random state=0)
In [53]: M joblib.dump(rf, "random.save")
   Out[53]: ['random.save']
In [54]: M y pred=rf.predict(x test)
            y pred
   Out[54]: array([ 928.15125 , 2177.1600974 , 6564.15457131, ..., 566.04522143,
                   7661.12527882, 831.72857143])
In [55]: N y test
   Out[55]: array([ 500, 3850, 6500, ..., 500, 7850, 1890], dtype=int64)
In [56]: M from sklearn.metrics import r2 score
            r2 score(y test,y pred)
   Out[56]: 0.8128590913835658
```

CONCLUSION:

- The main objective of this research was to develop a prototype of the system that can be used by the buyer.
- Multiple machine learning algorithms were developed and used for this research. Random forest regression proved to best-fit for development of the vehicle resale system when compared by giving highest accuracy.

THANK YOU