

Professional Readiness for Innovation, Employability, and Entrepreneurship

PROJECT REPORT

Title : Car Resale Value Prediction
Team ID : PNT2022TMID17930
Team Lead : Nabil Ahamed H (9517201904097)
Members : Avinash R (9517201904025)
Gobhinath I (9517201904040)
Suhaanjith TS (9517201904160)

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1. INTRODUCTION

1.1 Project Overview

This system “Car Resale Value Prediction” aims to build a regression model to predict used cars' resale value based on multiple aspects, including vehicle mileage, year of manufacturing, fuel consumption, transmission, road tax, fuel type, and engine size. This model can benefit sellers, buyers, and car manufacturers in the used cars market. Upon completion, it can output a relatively accurate price prediction based on the information that user's input. Various regression methods, including linear regression, polynomial regression, support vector regression, decision tree regression, and random forest regression, were applied in the research to achieve the highest accuracy.

This system was implemented as a web application where the user enters the details of the car to get an estimation of the car's resale value.

1.2 Purpose

Car resale value prediction helps the user to predict the resale value of the car depending upon various features like kilometers driven, fuel type, etc. The purpose of this system is of commercial interest to sellers/financer to be able to predict the resale value of cars with better accuracy. The most essential elements for forecast are brand and model, period use of vehicle, mileage of vehicle, gear type and fuel type utilized in the vehicle just as fuel utilization per mile profoundly influences cost of a vehicle because of continuous changes in the cost of a fuel. In view of the differing highlights and factors, and furthermore with the assistance of master information the vehicle resale value forecast has been done precisely.

2. LITERATURE SURVEY

2.1 Existing problem

With difficult economic conditions, it is likely that sales of second-hand imported (reconditioned) cars and used cars will increase. In many developed countries, it is common to lease a car rather than buying it outright. After the lease period is over, the buyer has the possibility to buy the car at its residual value, i.e., its expected resale value. Thus, it is of commercial interest to sellers/financers to be able to predict the salvage value (residual value) of cars with accuracy.

2.2 References

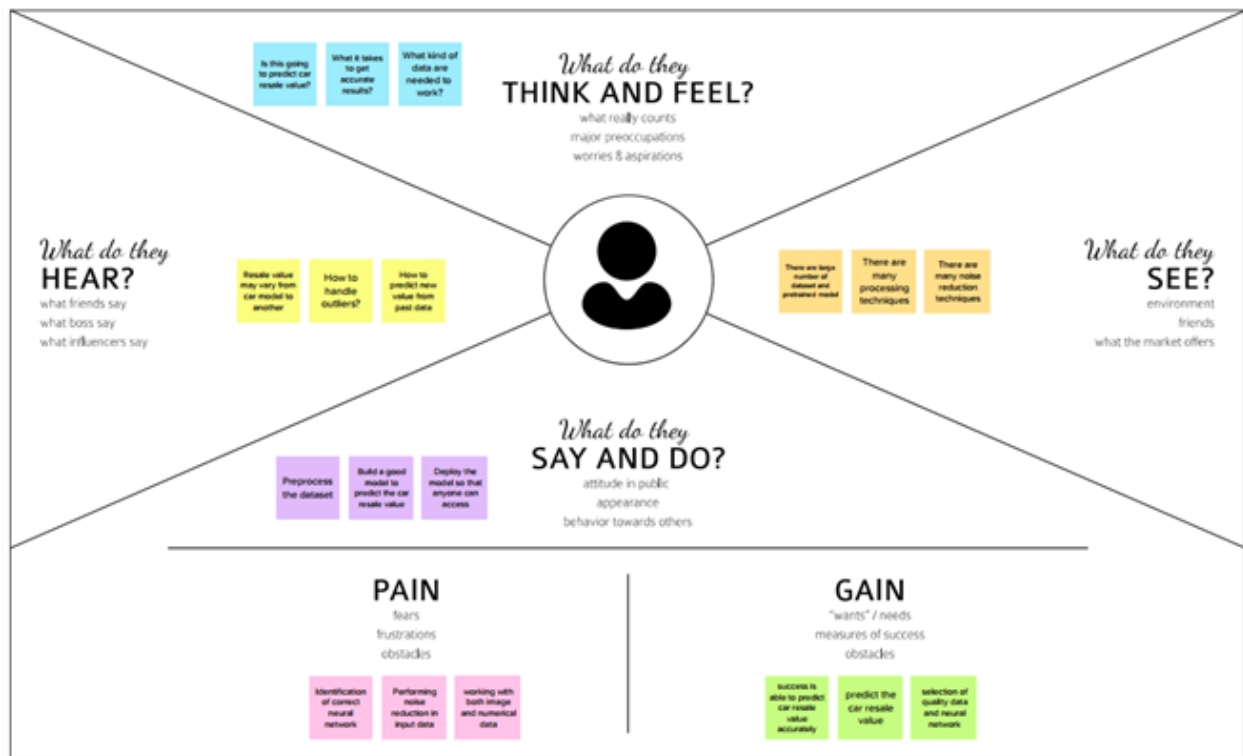
Project Title	Author	Abstract
Price Prediction of Used Cars Using Machine Learning	Chuyang Jin	This work aims to build a model to predict used cars' reasonable prices based on multiple aspects. Various regression methods, including linear regression, polynomial regression, support vector regression, decision tree regression, and random forest regression, were applied in the work to obtain highest accuracy. Compared to previous research, the resulting model includes more aspects of used cars while also having a higher prediction accuracy.
Prediction of Prices for Used Car by using Regression Models (2018)	Nitis Monburinon, Prajak Chertchom, Thongchai Kaewkiriya, Suwat Rungpheung, Sabir Buya, Pitchayakit Boonpou.	In this work, a model to evaluate price based on big data analysis is proposed. It takes advantage of vehicle data and vehicle transaction data to analyze the price data for each type of vehicles. The work uses optimized Back Propagation neural network algorithm.
Car Price Prediction Using Machine Learning (2019)	Enis gegic, Becir Isakovic, Dino Keco, Zerina Masetic, Jasmin Kevric.	In this work, several distinct attributes are analyzed for the reliable and accurate prediction. The work is to build a model to predict the resale price of cars in Bosnia and Herzegovina
Used Car price prediction (2021)	Praful Rane, Deep Pandya, Dhawal Kotak.	In this work, machine learning models that can accurately predict the price of a used car based on its features was built. They have implemented and evaluated various learning methods on dataset consisting of the sale prices of different models.
Prediction of Used Car Price Based on Supervised Learning Algorithm (2021)	Feng Wang, Xusong Zhang; Qiang Wang	In this work, Extra Trees Regressor, Random Forest Regressor was used. Finally, the algorithm was optimized by using the hyperparameter function. The results show that $R^2 = 0.9807$ obtained from extreme random numbers is the best performance. The algorithm was obtained and validated with new data to derive the final algorithm model.

2.3 Problem Statement Definition

It is easy for any company to price their new cars based on the manufacturing and marketing cost it involves. But when it comes to a used car it is quite difficult to define a price because it involves it is influenced by various parameters like car brand, manufactured year etc. The goal of our system is to predict the best price for a used car in the based on the previous data related to sold cars using machine learning.

3. IDEATION & PROPOSED SOLUTION


3.1 Empathy Map Canvas



3.2 Ideation & Brainstorming

Step-1: Team Gathering, Collaboration and Select the Problem Statement

Template



Brainstorm & idea prioritization

Use this template in your own brainstorming sessions so your team can unleash their imagination and start shaping concepts even if you're not sitting in the same room.

⌚ 10 minutes to prepare
🕒 1 hour to collaborate
👥 2-8 people recommended

[Share template feedback](#)

➦

Before you collaborate

A little bit of preparation goes a long way with this session. Here's what you need to do to get going.

⌚ 10 minutes

➦

Team gathering

Define who should participate in the session and send an invite. Share relevant information or pre-work ahead.

➦

Set the goal

Think about the problem you'll be focusing on solving in the brainstorming session.

➦

Learn how to use the facilitation tools

Use the Facilitation Superpowers to run a happy and productive session.

[Open article](#)

1

Define your problem statement

The problem statement is to consider the main factors which would affect the resale value of a vehicle and to build a regression model that would give the nearest resale value of the vehicle.

⌚ 5 minutes

PROBLEM

How might we predict the resale value of a car?

Key rules of brainstorming

➦ Stay in topic.

➦ Encourage wild ideas.

➦ Defer judgment.

➦ Listen to others.

➦ Go for volume.

➦ If possible, be visual.

Step-2: Brainstorm, Idea Listing and Grouping

Brainstorm

Write down any ideas that come to mind that address your problem statement.

🕒 10 minutes

Nabil Ahamed H

Understand the problem	Check for similar datasets	Identify improving techniques
Analyze the UI	Implement deployment processes	Can the accuracy be increased
Understand the visualization techniques	Explore the working procedures of APIs	Discuss with our friends to get an idea

Gobhinath I

To find appropriate dataset and understand it	Explore different use case for solutions	Identify the need for the model and building it
Test and evaluate the model for different cases	Implement the model	Check for accuracy in accuracy
Visualize the results done by the model from friendly way	Analyze other related APIs and what	Implement it for that results to show off a better way

Suhasrjith TS

Understand the problem statement defined	Understand the different type of algorithms	Explore and test the suitable dataset
Propose the dataset using suitable techniques	Explore about data model	Identify different APIs that can be used
Increase or decrease the value of hyper-parameter settings	Explore about model testing and deployment process	Analyze the result and visualize it

Avinash R

Analyze the problem Statement	Study APIs model and get a good idea	Check for other similar problem statements
Input the Suitable dataset	To implement the Model	Train the model
To check different ways to select hyper	Analyze the different case of hyper	Check for some improvements

Group ideas

Take turns sharing your ideas while clustering similar or related notes as you go. Once all sticky notes have been grouped, give each cluster a sentence-like label. If a cluster is bigger than six sticky notes, try and see if you can break it up into smaller sub-groups.

🕒 20 minutes

Working with Model

Use the accuracy be increased	Explore the working procedures of APIs	Identify the need for the model and building it
Test and evaluate the model for different cases	Implement the model	Identify different APIs that can be used
Increase or decrease the value of hyper-parameter settings	Explore about model testing and deployment process	Study APIs model and get a good idea
To implement the Model	Train the model	To check different ways to select hyper

Dataset

Check for similar datasets	To find appropriate dataset and understand it	Explore and test the suitable dataset
Propose the dataset using suitable techniques	Input the Suitable dataset	

Visualization

Analyze the result and visualize it	Understand the visualization techniques	Visualize the results done by the model in user-friendly way
Analyze the UI		

Accuracy

Use the accuracy be increased	Check for accuracy in accuracy	Increase or decrease the value of hyper-parameter settings
-------------------------------	--------------------------------	--

Step-3: Idea Prioritization

4

Prioritize

Your team should all be on the same page about what's important moving forward. Place your ideas on this grid to determine which ideas are important and which are feasible.

20 minutes



3.3 Proposed Solution

S. No:	Parameter	Description
1.	Problem Statement (Problem to be solved)	With difficult economic conditions, it is likely that sales of second-hand imported (reconditioned) cars and used cars will increase. In many developed countries, it is common to lease a car rather than buying it outright. After the lease period is over, the buyer has the possibility to buy the car at its residual value, i.e., its expected resale value. Thus, it is of commercial interest to sellers/financers to be able to predict the salvage value (residual value) of cars with accuracy.
2.	Idea / Solution description	In order to predict the resale value of the car, we proposed an intelligent, flexible, and effective system that is based on using regression algorithms. Considering the main factors which would affect the resale value of a vehicle a regression model is to be built that would give the nearest resale value of the vehicle. We will be using various regression algorithms and algorithm with the best accuracy will be taken as a solution, then it will be integrated to the web-based application where the user is notified with the status of his product.
3.	Novelty / Uniqueness	Car resale value price data frequently resides in several locations from various sources, such as industries or private persons, to various source systems. The organization as a whole contributes to the data. This data becomes accessible and usable when it is combined into a single, central system, such as an enterprise data warehouse (EDW).

4.	Social Impact / Customer Satisfaction	1. Enhanced resale value accuracy 2. Improved relationships with customers 3. Leads to increased quality of products and it's related after sales service
5.	Business Model (Revenue Model)	This business plan addresses all relevant concerns by presenting a comprehensive account of a month-by-month marketing strategy coupled with an extensive report on all aspects of the needs of a successful used car center.
6.	Scalability of the Solution	A variety of institutions must store, evaluate, and take action on the massive amounts of data being produced by the car resale industries as it expands quickly. India is a vast, culturally varied nation with a sizable population that is increasingly able to access centralized resale services.

3.4 Problem Solution fit

<u>1. Customer Segments</u> + Car mechanic + Customer	<u>1. Customer Limitation</u> Proper information about the car is to be known by the customer to find the resale value.	<u>5. Available Solution</u> To predict the resale value of the car, we use an intelligent, flexible, and effective system with web application.
<u>1. Problems</u> Customer should know the details of their car in web application.	<u>9. Problem root cause</u> <ul style="list-style-type: none"> No Proper platform for car resale value prediction. 	<u>2. Behavior</u> Customers are supposed to enter the car details in the web application to find the resale price of the car.

	<ul style="list-style-type: none"> No awareness of resale price of a used car. 	
<u>3. Triggers to Act</u> 1) When customers decided to sell their car. 2) When car mechanic decides to buy a used car.	<u>10. YourSolution</u> Using predictive modellingto predict the resale value of car.	<u>8. Channels ofBehavior</u> 1. Online: car details to be entered in web application.
<u>4. Emotions</u> Customers get an awareness of the resale price of their own car.		2. Offline: customers are supposed to collect the details of their car with the help of a car mechanic.

4. REQUIREMENT ANALYSIS

4.1 Functional Requirements

Following are the functional requirements of the proposed solution.

FR No.	Functional Requirement (Epic)	Sub Requirement (Story / Sub-Task)
FR-1	User Registration	Registration through Website
FR-2	User Confirmation	Confirmation via website
FR-3	Car Registration	Registering the car details
FR-4	Value Prediction	Predicting the car resale value

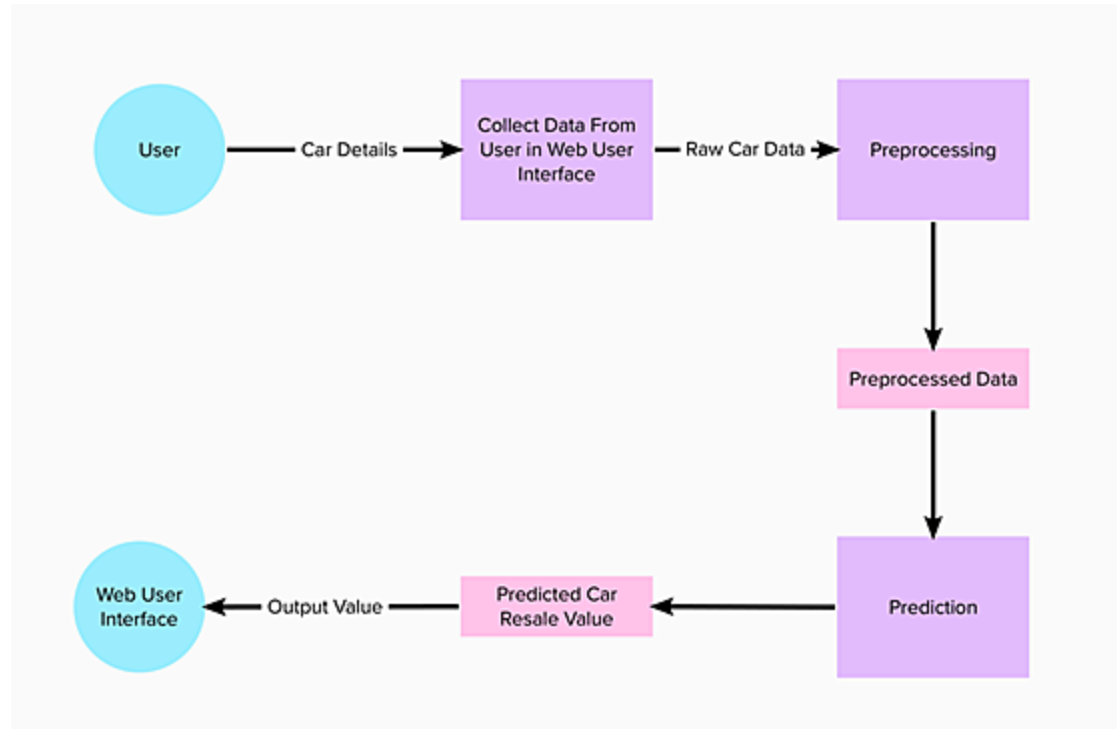
4.2 Non-Functional requirements

Following are the non-functional requirements of the proposed solution.

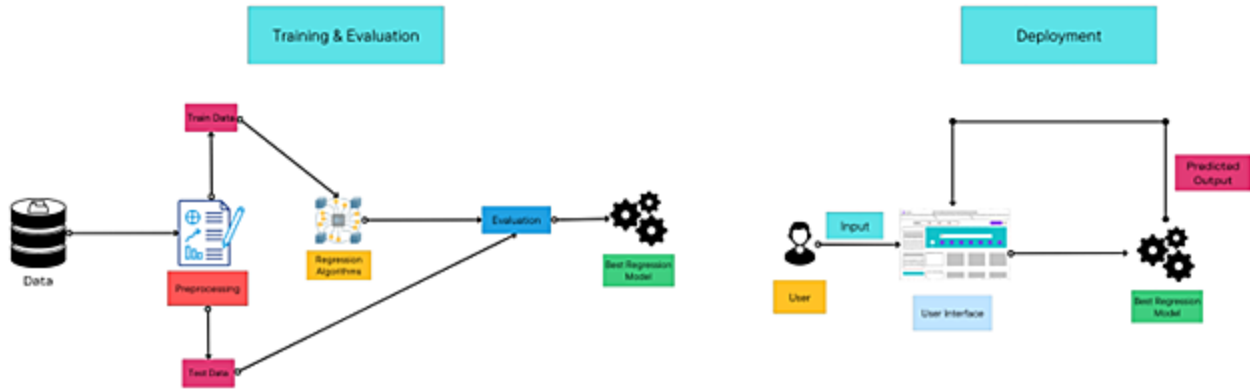
FR No.	Non-Functional Requirement	Description
NFR-1	Usability	Predicting the resale value
NFR-2	Security	Providing security to the website
NFR-3	Reliability	Providing high reliability by predicting values for different types of cars
NFR-4	Performance	Providing high performance by using some machine learning techniques
NFR-5	Availability	It is used for all types of cars
NFR-6	Scalability	Predicting values for different types of cars

5. PROJECT DESIGN

5.1 Data Flow Diagrams



5.2 Solution & Technical Architecture



5.3 User Stories

User Type	Functional Requirement (Epic)	User Story Number	User Story / Task	Acceptance criteria	Priority	Release
Customer (Desktop user)	Home Page	USN-1	As a user, I can view the home page of the web application.	I can view the homepage	Low	Sprint-1
Customer (Desktop user)	Data Entry	USN-2	As a user, I can enter my car details in the application.	I can enter the car details	Medium	Sprint-2
Customer (Desktop user)	View car Resale value	USN-3	As a user, I can view the resale value of my car.	I can view my car's resale value	Medium	Sprint-3
Customer (Desktop user)	Resale Value Prediction	USN-4	As a user, I expect the application to predict the resale value of my car.	I expect the application to predict my car resale price	High	Sprint-4

6. PROJECT PLANNING & SCHEDULING

6.1 Sprint Planning & Estimation

Title	Description	Date
Literature Survey and Information Gathering	Gathering Information by referring the technical papers, research publications etc.	3 September 2022
Prepare Empathy Map	To capture user pain and gains Prepare List of Problem Statement	10 September 2022
Ideation	Prioritize a top 3 ideas based on feasibility and Importance	17 September 2022
Proposed Solution	Solution include novelty, feasibility, business model, social impact and scalability of solution	24 September 2022
Problem Solution Fit	Solution fit document	1 October 2022
Solution Architecture	Solution Architecture	1 October 2022
Customer Journey	To Understand User Interactions and experiences with application	8 October 2022
Functional Requirement	Prepare functional Requirement	12 October 2022
Data flow Diagrams	Data flow diagram	12 October 2022
Technology Architecture	Technology Architecture diagram	12 October 2022
Milestone & sprint delivery plan	Activity what we done & further plans	22 October 2022
Project Development- Delivery of sprint 1,2,3 &4	Develop and submit the developed code by testing it	24 October 2022 – 19 November 2022

6.2 Sprint Delivery Schedule

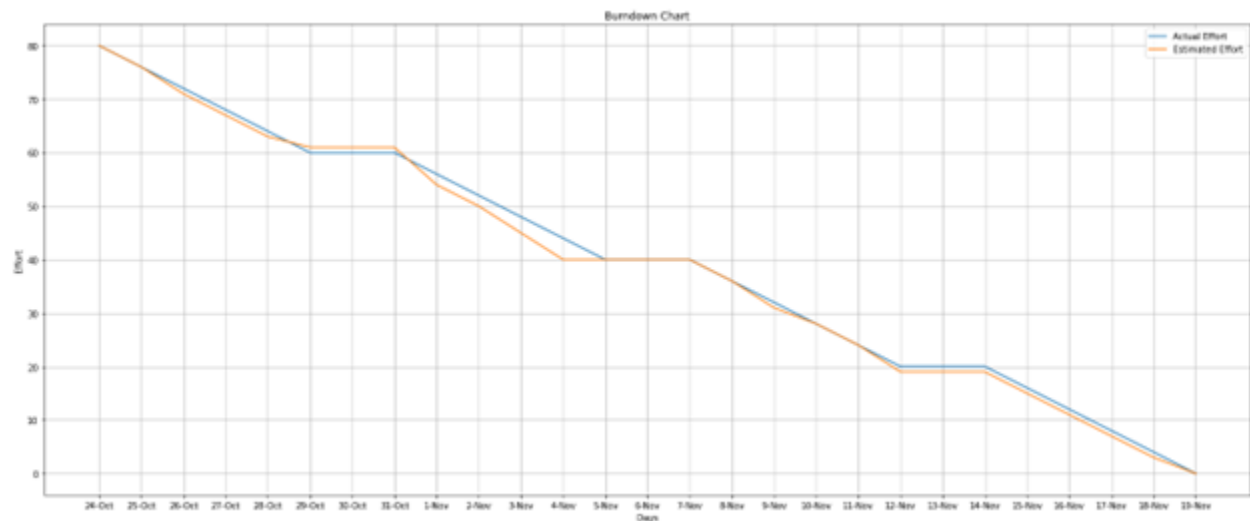
Sprint	Functional Requirement (Epic)	User Story Number	User Story / Task	Story Points	Priority	Team Members
Sprint-1	Home Page	USN-1	As a user, I can view the home page of the web application.	20	Low	Suhaanjith TS
Sprint-2	Data Entry	USN-2	As a user, I can enter my car details in the application.	20	Medium	Gobhinath I

Sprint-3	Car resale value display	USN-3	As a user, I can view the resale value of my car.	20	Medium	Nabil Ahamed H
Sprint-4	Resale Value Prediction	USN-4	As a user, I expect the application to predict the resale value of my car.	20	Medium	Avinash R

6.3 Project Tracker

Sprint	Total Story Points	Duration	Sprint Start Date	Sprint End Date (Planned)	Story Points Completed (as on Planned End Date)	Sprint Release Date (Actual)
Sprint-1	20	6 Days	24 Oct 2022	29 Oct 2022	20	29 Oct 2022
Sprint-2	20	6 Days	31 Oct 2022	05 Nov 2022	20	05 Nov 2022
Sprint-3	20	6 Days	07 Nov 2022	12 Nov 2022	20	12 Nov 2022
Sprint-4	20	6 Days	14 Nov 2022	19 Nov 2022	20	19 Nov 2022

6.4 Burndown Chart



6.5 Reports from JIRA

	OCT							NOV							NOV							NOV							NOV						
	24	25	26	27	28	29	30	31	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24			
Sprints	CRVP Sprint 1							CRVP Sprint 2							CRVP Sprint 3							CRVP Sprint 4													
CRVP-1 Home Page	[Progress Bar]																																		
CRVP-2 Data Entry								[Progress Bar]																											
CRVP-3 Car Resale Value Display															[Progress Bar]																				
CRVP-4 Resale Value Prediction																						[Progress Bar]													

7. CODING & SOLUTIONING

7.1 Home Page

Displays the home page of the application.

Code:

1) car.html

```
<!DOCTYPE html>
<html lang="en" dir="ltr">
  <head>
    <meta charset="utf-8">
    <title>Car Resale Value Predicting Application</title>
    <link rel="icon" type="image/x-icon" href="../static/Images/favicon.ico">
    <link rel="stylesheet" href="../static/css/style.css">
    <link rel="stylesheet" href="https://cdn.jsdelivr.net/npm/font-awesome/4.7.0/css/font-awesome.min.css">
  </head>
  <body>
    <section class="header">
      <nav>
        <a href="/"></a>

      </nav>
      <div class="text-box">
        <h1>Car resale value Predictor</h1>
        <p>Best system to predict the amount of resale value based on the parameters provided
by the user .</p>
        <a href="/predict_page" class="visit-btn ">Check price</a>
      </div>
    </section>

  </body>
</html>
```

2) style.css

```
*{
  margin: 0;
  padding: 0;
}

.header{
  min-height: 100vh;
  width: 100%;
  background-image: linear-
gradient(rgba(25,30,30,0.7),rgba(25,30,30,0.7)),url(../Images/car1.png);
```

```

    background-position: center;
    background-size: cover;
    position: relative;
}
nav{
    display: flex;
    padding: 2% 6%;
    justify-content: space-between;
    align-items: center;
}
.nav-links{
    flex: 1;
    text-align: right;
}
.nav-links ul li{
    list-style: none;
    display: inline-block;
    padding: 8px 12px;
    position: relative;
}
.nav-links ul li a{
    color: white;
    text-decoration: none;
    font-size: 13px;
}
.text-box{
    text-align: center;
    position: relative;
    color: #FFE4C4;
    top: 50%;
}
.text-box h1{
    margin-top: 50px;
    font-size: 55px;
}
.text-box p{
    margin: 10px 0 40px;
    font-size: 15px;
}
.visit-btn{
    display: inline;
    border: 3px solid #fff;
    padding: 10px 14px;
    font-size: 15px;
    background: transparent;

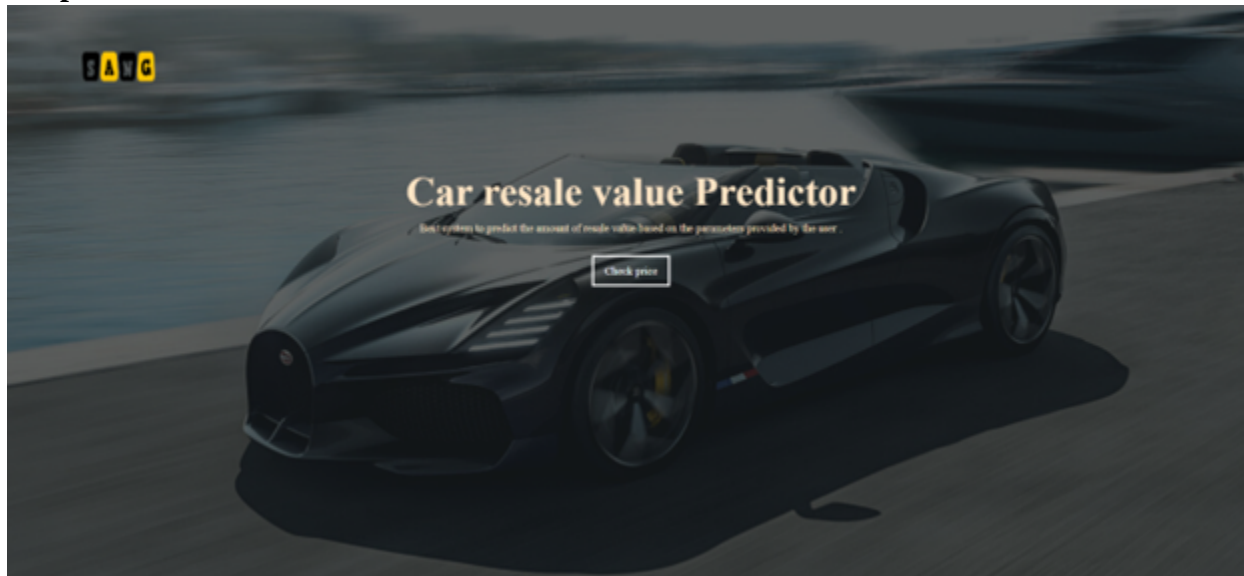
```

```

color: white;
text-decoration:none;
}

```

Output:



7.2 Data Entry Page

Allows user to enter the details about the car for which the resale value is to be predicted.

Code:

1) value.html

```

<!DOCTYPE html>
<html lang="en" dir="ltr">
<head>
<link rel="stylesheet" href="../static/css/value.css">
<title>Car Resale Value Predicting Application</title>
<link rel="icon" type="image/x-icon" href="../static/Images/favicon.ico">
<script src="https://kit.fontawesome.com/b9b6bac803.js" crossorigin="anonymous"></script>
<link rel="stylesheet" href="https://cdn.jsdelivr.net/npm/font-awesome/4.7.0/css/font-awesome.min.css">
<style>
    table, th, td {
        padding: 10px;
    }
</style>
</head>
<body>
<div class="container">
    <div class="header">
        <h1>Get the Accurate Resale Value of Your Car</h1>
    </div>

```

```

<form action="http://localhost:5000/predict" class="form">
<div class="form-control">
<label for="year" padding:10px>Registration year : </label>
<input id="year" maxlength="50" name="regyear" type="text" autocomplete="off"/>
<i class="fas fa-check-circle"></i>
<i class="fas fa-exclamation-circle"></i>
<span></span>
</div>
<div class="form-control">
<label for="month">Registration Month : </label>
<input id="month" maxlength="50" name="regmonth" type="text" autocomplete="off"/>
<i class="fas fa-check-circle"></i>
<i class="fas fa-exclamation-circle"></i>
<span></span>
</div>
<div class="form-control">
<label for="power">Power of car in PS: </label>
<input id="power" maxlength="50" name="powerps" type="text" autocomplete="off"/>
<i class="fas fa-check-circle"></i>
<i class="fas fa-exclamation-circle"></i>
<span></span>
</div>
<div class="form-control">
<label for="kilometer">Kilometers that car have driven : </label>
<input id="kilometer" maxlength="50" name="kms" type="text" autocomplete="off"/>
<i class="fas fa-check-circle"></i>
<i class="fas fa-exclamation-circle"></i>
<span></span>
</div>
<div class="form-control">
    <h3>Gear Type</h3>
    <table style="width:50%">
        <tr>
            <th></th>
            <th></th>
        </tr>
        <tr>
            <td>Manual</td>
            <td><input type="radio" name="geartype" value="manual" id="manual" /></td>
        </tr>
        <tr>
            <td>Automatic</td>
            <td><input type="radio" name="geartype" value="automatic" id="automatic"
/></td>
        </tr>
        <tr>
            <td>Not mentioned</td>

```

```

        <td><input type="radio" name="geartype" value="not-declared" id="not" /></td>
    </tr>
</table>
<i class="fas fa-check-circle"></i>
<i class="fas fa-exclamation-circle"></i>
<span></span>
</div>
<div class="form-control">
    <h3>Your car is repaired or damaged :</h3>
    <table style="width:50%">
        <tr>
            <th></th>
            <th></th>
        </tr>
        <tr>
            <td>Yes</td>
            <td><input type="radio" name="damage" value="yes" id="yes"/></td>
        </tr>
        <tr>
            <td>No</td>
            <td><input type="radio" name="damage" value="no" id="no"/></td>
        </tr>
        <tr>
            <td>Not Declared</td>
            <td><input type="radio" name="damage" value="not-declared"
id="notdec"/></td>
        </tr>
    </table>
    <i class="fas fa-check-circle"></i>
    <i class="fas fa-exclamation-circle"></i>
    <span></span>
</div>
<div class="form-control">
    <label for="model">Model Type : </label>
    <select name="model" id="model">
        <option value="" disabled selected hidden>Choose Model Name...</option>
        <option value="golf">Golf </option>
        <option value="grand">Grand </option>
        <option value="fabia">Fabia </option>
        <option value="3er">3er </option>
        <option value="2_reihe">2 Reihe </option>
        <option value="andere">Andere </option>
        <option value="c_max">C Max </option>
        <option value="3_reihe">3 Reihe </option>
        <option value="passat">Passat </option>
        <option value="navara">Navara </option>
        <option value="ka">Ka </option>
    </select>
</div>

```

<option value="polo">Polo </option>
<option value="twingo">Twingo </option>
<option value="a_klasse">A klasse </option>
<option value="scirocco">Scirocco </option>
<option value="5er">5er </option>
<option value="meriva">Meriva </option>
<option value="arosa">Arosa </option>
<option value="c4">C4 </option>
<option value="civic">Civic </option>
<option value="transporter">Transporter </option>
<option value="punto">Punto </option>
<option value="e_klasse">E Klasse </option>
<option value="clio">Clio </option>
<option value="kadett">Kadett </option>
<option value="kangoo">Kangoo </option>
<option value="corsa">Corsa </option>
<option value="one">One </option>
<option value="fortwo">Fortwo </option>
<option value="1er">1er </option>
<option value="b_klasse">B Klasse </option>
<option value="signum">Signum </option>
<option value="astra">Astra </option>
<option value="a8">A8 </option>
<option value="jetta">Jetta </option>
<option value="fiesta">Fiesta </option>
<option value="c_klasse">C Klasse </option>
<option value="micra">Micra </option>
<option value="vito">Vito </option>
<option value="sprinter">Sprinter </option>
<option value="156">156 </option>
<option value="escort">Escort </option>
<option value="forester">Forester </option>
<option value="xc_reihe">Xc Reihe </option>
<option value="scenic">Scenic </option>
<option value="a4">A4 </option>
<option value="a1">A1 </option>
<option value="insignia">Insignia </option>
<option value="combo">Combo </option>
<option value="focus">Focus </option>
<option value="tt">Tt </option>
<option value="a6">A6 </option>
<option value="jazz">Jazz </option>
<option value="omega">Omega </option>
<option value="slk">Slk </option>
<option value="7er">7er </option>
<option value="80">80 </option>
<option value="147">147 </option>
<option value="glk">Glk </option>

<option value="100">100 </option>
<option value="z_reihe">Z Reihe </option>
<option value="sportage">Sportage </option>
<option value="sorento">Sorento </option>
<option value="v40">V40 </option>
<option value="5er">5er </option>
<option value="ibiza">Ibiza </option>
<option value="3er">3er </option>
<option value="mustang">Mustang </option>
<option value="eos">Eos </option>
<option value="touran">Touran </option>
<option value="getz">Getz </option>
<option value="a3">A3 </option>
<option value="almera">Almera </option>
<option value="megane">Megane </option>
<option value="7er">7er </option>
<option value="1er">1er </option>
<option value="lupo">Lupo </option>
<option value="r19">R19 </option>
<option value="zafira">Zafira </option>
<option value="caddy">Caddy </option>
<option value="2_reihe">2 Reihe </option>
<option value="mondeo">Mondeo </option>
<option value="cordoba">Cordoba </option>
<option value="colt">Colt </option>
<option value="impreza">Impreza </option>
<option value="vectra">Vectra </option>
<option value="berlingo">Berlingo </option>
<option value="80">80 </option>
<option value="m_klasse">M Klasse </option>
<option value="tiguan">Tiguan </option>
<option value="i_reihe">I Reihe </option>
<option value="espace">Espace </option>
<option value="sharan">Sharan </option>
<option value="6_reihe">6 Reihe </option>
<option value="panda">Panda </option>
<option value="up">Up </option>
<option value="seicento">Seicento </option>
<option value="ceed">Ceed </option>
<option value="5_reihe">5 Reihe </option>
<option value="yeti">Yeti </option>
<option value="octavia">Octavia </option>
<option value="mii">Mii </option>
<option value="rx_reihe">Rx Reihe </option>
<option value="6er">6er </option>
<option value="modus">Modus </option>
<option value="fox">Fox </option>
<option value="matiz">Matiz </option>

<option value="beetle">Beetle </option>
<option value="c1">C1 </option>
<option value="rio">Rio </option>
<option value="touareg">Touareg </option>
<option value="logan">Logan </option>
<option value="spider">Spider </option>
<option value="cuore">Cuore </option>
<option value="s_max">S Max </option>
<option value="a2">A2 </option>
<option value="x_reihe">X Reihe </option>
<option value="a5">A5 </option>
<option value="galaxy">Galaxy </option>
<option value="c3">C3 </option>
<option value="viano">Viano </option>
<option value="s_klasse">S Klasse </option>
<option value="1_reihe">1 Reihe </option>
<option value="avensis">Avensis </option>
<option value="sl">SL </option>
<option value="roomster">Roomster </option>
<option value="q5">Q5 </option>
<option value="kaefer">Kaefer </option>
<option value="santa">Santa </option>
<option value="cooper">Cooper </option>
<option value="leon">Leon </option>
<option value="4_reihe">4 Reihe </option>
<option value="500">500 </option>
<option value="laguna">Laguna </option>
<option value="ptcruiser">Ptcruiser </option>
<option value="clk">Clk </option>
<option value="primera">Primera </option>
<option value="exeo">Exeo </option>
<option value="159">159 </option>
<option value="transit">Transit </option>
<option value="juke">Juke </option>
<option value="qashqai">Qashqai </option>
<option value="carisma">Carisma </option>
<option value="accord">Accord </option>
<option value="corolla">Corolla </option>
<option value="lanos">Lanos </option>
<option value="phaeton">Phaeton </option>
<option value="boxster">Boxster </option>
<option value="verso">Verso </option>
<option value="swift">Swift </option>
<option value="rav">Rav </option>
<option value="kuga">Kuga </option>
<option value="picanto">Picanto </option>
<option value="kalos">Kalos </option>
<option value="superb">Superb </option>

<option value="stilo">Stilo </option>
<option value="alhambra">Alhambra </option>
<option value="911">911 </option>
<option value="mx_reihe">Mx Reihe </option>
<option value="m_reihe">M Reihe </option>
<option value="roadster">Roadster </option>
<option value="ypsilon">Ypsilon </option>
<option value="cayenne">Cayenne </option>
<option value="galant">Galant </option>
<option value="justy">Justy </option>
<option value="90">90 </option>
<option value="sirion">Sirion </option>
<option value="crossfire">Crossfire </option>
<option value="6_reihe">6 Reihe </option>
<option value="agila">Agila </option>
<option value="duster">Duster </option>
<option value="cr_reihe">Cr Reihe </option>
<option value="v50">V50 </option>
<option value="discovery">Discovery </option>
<option value="c_reihe">C Reihe </option>
<option value="v_klasse">V Klasse </option>
<option value="yaris">Yaris </option>
<option value="c5">C5 </option>
<option value="aygo">Aygo </option>
<option value="cc">Cc </option>
<option value="carnival">Carnival </option>
<option value="fusion">Fusion </option>
<option value="bora">Bora </option>
<option value="forfour">Forfour </option>
<option value="100">100 </option>
<option value="cl">Cl </option>
<option value="tigra">Tigra </option>
<option value="156">156 </option>
<option value="300c">300c </option>
<option value="100">100 </option>
<option value="147">147 </option>
<option value="q3">Q3 </option>
<option value="spark">Spark </option>
<option value="v70">V70 </option>
<option value="x_type">X Type </option>
<option value="5_reihe">5 Reihe </option>
<option value="ducato">Ducato </option>
<option value="s_type">S Type </option>
<option value="x_trail">X Trail </option>
<option value="toledo">Toledo </option>
<option value="altea">Altea </option>
<option value="7er">7er </option>
<option value="voyager">Voyager </option>

<option value="calibra">Calibra </option>
 <option value="bravo">Bravo </option>
 <option value="range_rover">Range Rover </option>
 <option value="antara">Antara </option>
 <option value="tucson">Tucson </option>
 <option value="q7">Q7 </option>
 <option value="citigo">Citigo </option>
 <option value="jimny">Jimny </option>
 <option value="cx_reihe">Cx Reihe </option>
 <option value="wrangler">Wrangler </option>
 <option value="lybra">Lybra </option>
 <option value="range_rover_sport">Range Rover Sport </option>
 <option value="lancer">Lancer </option>
 <option value="159">159 </option>
 <option value="freelander">Freelander </option>
 <option value="captiva">Captiva </option>
 <option value="c2">C2 </option>
 <option value="500">500 </option>
 <option value="range_rover_evoque">Range Rover Evoque </option>
 <option value="sander0">Sander0 </option>
 <option value="note">Note </option>
 <option value="900">900 </option>
 <option value="147">147 </option>
 <option value="defender">Defender </option>
 <option value="cherokee">Cherokee </option>
 <option value="clubman">Clubman </option>
 <option value="samara">Samara </option>
 <option value="2_reihe">2 Reihe </option>
 <option value="1er">1er </option>
 <option value="3er">3er </option>
 <option value="601">601 </option>
 <option value="3_reihe">3 Reihe </option>
 <option value="4_reihe">4 Reihe </option>
 <option value="5er">5er </option>
 <option value="6_reihe">6 Reihe </option>
 <option value="legacy">Legacy </option>
 <option value="pajero">Pajero </option>
 <option value="auris">Auris </option>
 <option value="niva">Niva </option>
 <option value="5_reihe">5 Reihe </option>
 <option value="s60">S60 </option>
 <option value="nubira">Nubira </option>
 <option value="vivaro">Vivaro </option>
 <option value="g_klasse">G Klasse </option>
 <option value="lodgy">Lodgy </option>
 <option value="850">850 </option>
 <option value="serie_2">Serie 2 </option>
 <option value="6er">6er </option>

```

<option value="charade">Charade </option>
<option value="croma">Croma </option>
<option value="outlander">Outlander </option>
<option value="gl">Gl </option>
<option value="doblo">Doblo </option>
<option value="musa">Musa </option>
<option value="amarok">Amarok </option>
<option value="156">156 </option>
<option value="move">Move </option>
<option value="9000">9000 </option>
<option value="v60">V60 </option>
<option value="145">145 </option>
<option value="aveo">Aveo </option>
<option value="200">200 </option>
<option value="300c">300c </option>
<option value="b_max">B Max </option>
<option value="delta">Delta </option>
<option value="terios">Terios </option>
<option value="rangerover">RangeRover </option>
<option value="90">90 </option>
<option value="materia">Materia </option>
<option value="kalina">Kalina </option>
<option value="elefantino">Elefantino </option>
<option value="i3">I3 </option>
<option value="kappa">Kappa </option>
<option value="serie_3">Serie 3 </option>
<option value="48429">48429 </option>
<option value="serie_1">Serie 1 </option>
<option value="discovery_sport">Discovery Sport </option>
</select>
<i class="fas fa-check-circle"></i>
<i class="fas fa-exclamation-circle"></i>
<span></span>
</div>
<div class="form-control">
<label for="brand">Brand :</label>
<select name="brand" id="brand">
<option value="" disabled selected hidden>Choose Brand Name...</option>
<option value="volkswagen">Volkswagen </option>
<option value="audi">Audi </option>
<option value="jeep">Jeep </option>
<option value="skoda">Skoda </option>
<option value="bmw">Bmw </option>
<option value="peugeot">Peugeot </option>
<option value="ford">Ford </option>
<option value="mazda">Mazda </option>
<option value="nissan">Nissan </option>
<option value="renault">Renault </option>

```

```

<option value="mercedes_benz">Mercedes Benz </option>
<option value="opel">Opel </option>
<option value="seat">Seat </option>
<option value="citroen">Citroen </option>
<option value="honda">Honda </option>
<option value="fiat">Fiat </option>
<option value="mini">Mini </option>
<option value="smart">Smart </option>
<option value="hyundai">Hyundai </option>
<option value="sonstige_autos">Sonstige Autos </option>
<option value="alfa_romeo">Alfa Romeo </option>
<option value="subaru">Subaru </option>
<option value="volvo">Volvo </option>
<option value="mitsubishi">Mitsubishi </option>
<option value="kia">Kia </option>
<option value="suzuki">Suzuki </option>
<option value="lancia">Lancia </option>
<option value="porsche">Porsche </option>
<option value="toyota">Toyota </option>
<option value="chevrolet">Chevrolet </option>
<option value="dacia">Dacia </option>
<option value="daihatsu">Daihatsu </option>
<option value="trabant">Trabant </option>
<option value="saab">Saab </option>
<option value="chrysler">Chrysler </option>
<option value="jaguar">Jaguar </option>
<option value="daewoo">Daewoo </option>
<option value="rover">Rover </option>
<option value="land_rover">Land Rover </option>
<option value="lada">Lada </option>
</select>
<i class="fas fa-check-circle"></i>
<i class="fas fa-exclamation-circle"></i>
<span></span>
</div>
<div class="form-control">
<label for="fuelType">Fuel Type </label>
<select name="fuelType" id="fuel">
<option value="" disabled selected hidden>Choose Fuel Type...</option>
<option value="petrol"> Petrol </option>
<option value="diesel"> Diesel </option>
<option value="not-declared"> Not Declared </option>
<option value="lpg">LPG </option>
<option value="cng">CNG </option>
<option value="hybrid">Hybrid </option>
<option value="others">Others </option>
<option value="electric">Electric </option>
</select>

```

```

<i class="fas fa-check-circle"></i>
<i class="fas fa-exclamation-circle"></i>
<span></span>
</div>
<div class="form-control">
<label for="vehicletype">Vehicle type: </label>
<select name="vehicletype" id="vehicle" >
<option value="" disabled selected hidden>Choose Vehicle Type...</option>
<option value="coupe">Coupe </option>
<option value="suv">SUV </option>
<option value="kleinwagen">Kleinwagen </option>
<option value="limousine">Limousine </option>
<option value="cabrio">Cabrio </option>
<option value="bus">Bus </option>
<option value="kombi">Kombi </option>
<option value="andere">Andere </option>
<option value="volkswagen">Volkswagen </option>
</select>
<i class="fas fa-check-circle"></i>
<i class="fas fa-exclamation-circle"></i>
<span></span>
</div>
<input type="submit" id="submit"></input>
</form>
</div>
</body>
</html>

```

2) value.css

```

*{
padding:0px;
margin:0;
box-sizing:border-box;
font-family: cursive;
font-weight: bold;
color: #E74C3C;
}

body{
background-image: linear-
gradient(rgba(25,30,30,0.7),rgba(25,30,30,0.7)),url(../Images/car2.png);
min-height:100vh;
display:flex;
justify-content:center;
align-items:center;
}

.header{

```

```

    color:Black;
    text-align:center;
    padding:10px 0px 10px 100px;
}

#model{
    width:500px;
    color: black;
}

#brand{
    width:500px;
    color: black;
}

#fuel{
    width:500px;
    color: black;
}

#vehicle{
    width:500px;
    color: black;
}

.form{
    padding:30px 40px;
}

.form-control{
    margin-bottom:10px;
    padding-bottom: 20px;
    position:relative;
    margin-left: 100px;
}

.form-control label{
    display:block;
    margin-bottom:5px;
}

.form-control input{
    border: 2px solid #f0f0f0;
    width:80%;
    font-size :.8rem;
    padding:5px;
    display:inline-table;
}

```

```

.form-control i{
    position:absolute;
    right:20px;
    top:35px;
    visibility:hidden;
}

.form-control span{
    position:absolute;
    left:0;
    bottom:0;
    visibility:hidden;
    font-weight:bolder;
    font-style:italic;
    font-size:1rem;
}

.form-control.success input{
    border-color:#2ecc71;
}

.form-control.error input{
    border-color:#e74c3c;
}

.form-control.error span{
    color:red;
    visibility:visible;
}

.form-control.success i.fa-check-circle {
    border-color:#2ecc71;
    visibility:visible;
}

.form-control.error i.fa-exclamation-circle {
    border-color:#e74c3c;
    visibility:visible;
}

.form #submit{
    background-color:#E74C3C;
    border:none;
    outline:none;
    color:white;
    width:500px;
    border-radius:4px;
}

```



```
padding:10px;
cursor:pointer;
transition:all .5s;
font-size:1rem;
margin-left: 100px;
}

.form #submit:hover{
background-color:#6441a5;
}

.form-control #manual{
padding-bottom: 20%;
}
```

Output

Get the Accurate Resale Value of Your Car

Registration year :

Registration Month :

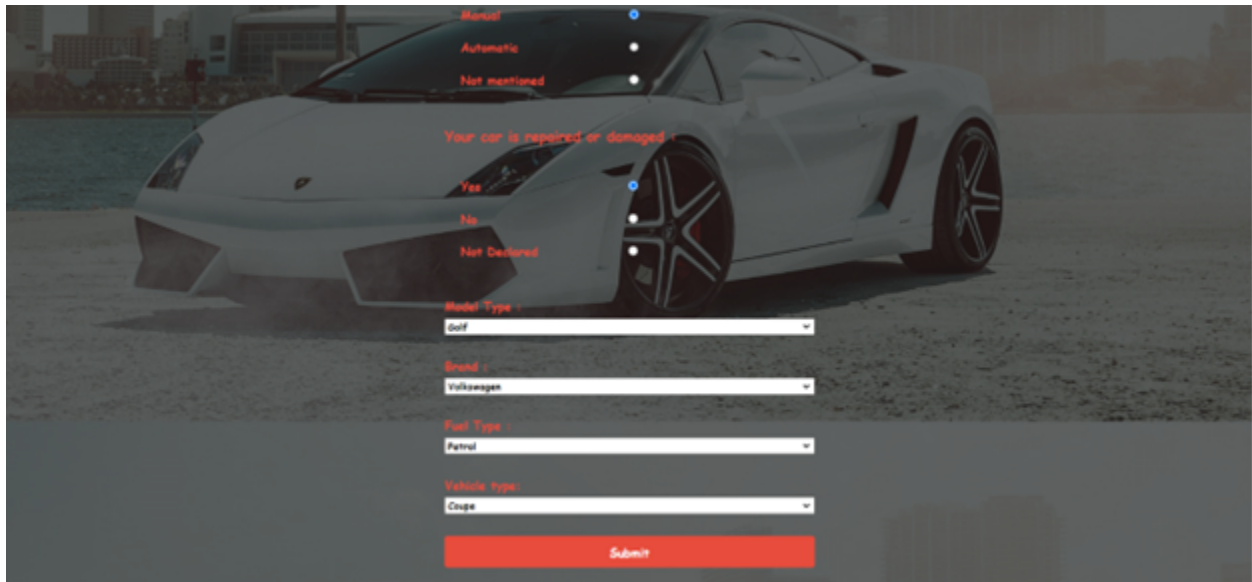
Power of car in PS:

Kilometers that car have driven :

Gear Type

☒ Manual
☐ Automatic
☐ Not mentioned

Your car is repaired or damaged :
☒ Yes



7.3 Output Display Page

The predicted resale car value is displayed in this page.

Code

1) predict.html

```
<!DOCTYPE html>
<html lang="en">
<head>
  <meta charset="UTF-8">
  <meta http-equiv="X-UA-Compatible" content="IE=edge">
  <meta name="viewport" content="width=device-width, initial-scale=1.0">
  <link rel="stylesheet" href="../static/css/predict.css">
  <title>Car Resale Value Predicting Application</title>
  <link rel="icon" type="image/x-icon" href="../static/Images/favicon.ico">
</head>
<body>
  <section class="header">
    <nav>
      <a href="/"></a>
    </nav>
    <div class="text-box">
      <h1>The Predicted Car Resale Value is </h1>
      <h1>{{predict}}</h1>
    </div>
  </section>

</body>
</html>
```

2) predict.css

```
.header{
  min-height: 100vh;
  width: 100%;
  background-image: linear-
gradient(rgba(25,30,30,0.7),rgba(25,30,30,0.7)),url(../Images/car3.jpg);
  background-position: center;
  background-size: cover;
  position: relative;
}

.text-box{
  text-align: center;
  position: relative;
  color: #FFE4C4;
  top:50%;
}

.text-box h1{
  margin-top: 50px;
  font-size: 55px;
}

.text-box p{
  margin: 10px 0 40px;
  font-size: 15px;
}

body{
  margin: 0;
}

nav{
  display:flex;
  padding: 2% 6%;
  justify-content: space-between;
  align-items: center;
}
```

Output



7.4 Model Selection & Hyperparameter Tuning

Code

```
import pandas as pd
import numpy as np
from sklearn.preprocessing import LabelEncoder
from sklearn.model_selection import train_test_split, GridSearchCV
from sklearn.metrics import mean_absolute_error, mean_squared_error, r2_score
import pickle
import wandb

#regression models
from sklearn.ensemble import BaggingRegressor, RandomForestRegressor,
HistGradientBoostingRegressor, ExtraTreesRegressor
from xgboost.sklearn import XGBRegressor
from lightgbm import LGBMRegressor

wandb.login(key='b75e0564aba32dce859c60044418df71ce7389a8')

data = pd.read_csv('../input/naalaiya-thiran/Preprocessed/autos_preprocessed.csv', header=0,
sep=',', encoding='Latin1')

labels = ['gearbox', 'notRepairedDamage', 'model', 'brand', 'fuelType', 'vehicleType']

mapper = {}
for i in labels:
    mapper[i] = LabelEncoder()
    mapper[i].fit(data[i])
    tr = mapper[i].transform(data[i])
    np.save(str('classes'+i+'.npy'), mapper[i].classes_)
```

```

data.loc[:, i+'_labels'] = pd.Series(tr, index=data.index)

labeled = data[['price', 'yearOfRegistration', 'powerPS', 'kilometer', 'monthOfRegistration']
               +[x+"_labels" for x in labels]]

print(labeled.columns)

def find_scores(Y_actual, Y_pred, X_train):
    mae = mean_absolute_error(Y_actual, Y_pred)
    mse = mean_squared_error(Y_actual, Y_pred)
    rmse = np.sqrt(mse)
    rmsle = np.log(rmse)
    r2 = r2_score(Y_actual, Y_pred)
    n, k = X_train.shape
    adj_r2_score = 1 - ((1-r2)*(n-1)/(n-k-1))

    wandb.log({"mae": mae, "mse": mse, 'rmse':rmse, 'rmsle':rmsle, 'r2':r2, 'adj_r2':adj_r2_score})

def bagging_regressor():
    config_defaults = {
        'n_estimators':100,
        'max_samples':0.4,
        'bootstrap':True,
        'random_state':42
    }
    wandb.init(config=config_defaults)
    config = wandb.config

    X = labeled.iloc[:,1:].values
    Y = labeled.iloc[:,0].values.reshape(-1,1)

    X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.4, random_state=42)

    model = BaggingRegressor(
        n_estimators=config.n_estimators,
        bootstrap=config.bootstrap,
        max_samples=config.max_samples,
        random_state = config.random_state)

    model.fit(X_train, Y_train)

    Y_pred = model.predict(X_test)

    find_scores(Y_test, Y_pred, X_train)

```

```

bagging_regressor_configs = {
    "name": 'BaggingRegressor',
    "method": "grid",
    "metric": {
        "name": "adj_r2",
        "goal": "maximize"
    },
    "parameters": {
        "n_estimators": {
            "values": [100, 200, 300]
        },
        "max_samples": {
            "values": [0.4, 0.5, 0.6]
        }
    }
}

sweep_id = wandb.sweep(sweep=bagging_regressor_configs, project="car_resale_value")
wandb.agent(sweep_id=sweep_id, function=bagging_regressor)

def random_forest_regressor():
    config_defaults = {
        'n_estimators': 100,
        'max_samples': 0.4,
        'criterion': 'squared_error',
        'bootstrap': True,
        'random_state': 42
    }
    wandb.init(config=config_defaults)
    config = wandb.config

    X = labeled.iloc[:, 1:].values
    Y = labeled.iloc[:, 0].values.reshape(-1, 1)

    X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.4, random_state=42)

    model = RandomForestRegressor(
        n_estimators=config.n_estimators,
        criterion=config.criterion,
        bootstrap=config.bootstrap,
        max_samples=config.max_samples,
        random_state=config.random_state)

    model.fit(X_train, Y_train)

```

```

Y_pred = model.predict(X_test)

find_scores(Y_test, Y_pred, X_train)

random_forest_configs = {
    "name": 'RandomForestRegressor',
    "method": "grid",
    "metric": {
        "name": "adj_r2",
        "goal": "maximize"
    },
    "parameters": {
        "n_estimators": {
            "values": [100, 200, 300]
        },
        "max_samples": {
            "values": [0.4, 0.5, 0.6]
        }
    }
}

sweep_id = wandb.sweep(sweep=random_forest_configs, project="car_resale_value")
wandb.agent(sweep_id=sweep_id, function=random_forest_regressor)

def hist_gradient_boost_regressor():
    config_defaults = {
        'loss': 'squared_error',
        'learning_rate': 0.1,
        'max_iter': 100,
        'random_state': 42
    }
    wandb.init(config=config_defaults)
    config = wandb.config

    X = labeled.iloc[:, 1:].values
    Y = labeled.iloc[:, 0].values.reshape(-1, 1)

    X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.4, random_state=42)

    model = HistGradientBoostingRegressor(
        loss=config.loss,
        learning_rate=config.learning_rate,
        max_iter=config.max_iter,
        random_state=config.random_state)

```

```

model.fit(X_train, Y_train)

Y_pred = model.predict(X_test)

find_scores(Y_test, Y_pred, X_train)

hist_gradient_boost_configs = {
    "name": 'HistGradientBoostingRegressor',
    "method": "grid",
    "metric": {
        "name": "adj_r2",
        "goal": "maximize"
    },
    "parameters": {
        "loss": {
            "values": ['squared_error', 'absolute_error']
        },
        "learning_rate": {
            "values": [0.01, 0.03, 0.05, 0.07]
        },
        "max_iter": {
            "values": [100, 200, 300]
        },
        "random_state": {
            "values": [42]
        }
    }
}

sweep_id = wandb.sweep(sweep=hist_gradient_boost_configs, project="car_resale_value")
wandb.agent(sweep_id=sweep_id, function=hist_gradient_boost_regressor)

def extra_tree_regressor():
    config_defaults = {
        'criterion': 'squared_error',
        'max_samples': 0.4,
        'bootstrap': True,
        'random_state': 42
    }
    wandb.init(config=config_defaults)
    config = wandb.config

    X = labeled.iloc[:, 1:].values
    Y = labeled.iloc[:, 0].values.reshape(-1, 1)

```



```

X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.4, random_state=42)

model = ExtraTreesRegressor(
    criterion=config.criterion,
    bootstrap = config.bootstrap,
    max_samples=config.max_samples,
    random_state = config.random_state)

model.fit(X_train, Y_train)

Y_pred = model.predict(X_test)

find_scores(Y_test, Y_pred, X_train)

extra_tree_configs = {
    "name": 'ExtraTreesRegressor',
    "method": "grid",
    "metric": {
        "name": "adj_r2",
        "goal": "maximize"
    },
    "parameters": {
        "criterion": {
            "values": ['squared_error', 'absolute_error']
        },
        "max_samples": {
            "values": [0.4, 0.5, 0.6]
        }
    }
}

sweep_id = wandb.sweep(sweep=extra_tree_configs, project="car_resale_value")
wandb.agent(sweep_id=sweep_id, function=extra_tree_regressor)

def XGB_regressor():
    config_defaults = {
        'learning_rate': 0.1,
        'n_estimators': 500,
        'booster': 'gbtree',
        'eta': 0.01,
        'random_state': 42
    }
    wandb.init(config=config_defaults)
    config = wandb.config

```

```

X = labeled.iloc[:,1:].values
Y = labeled.iloc[:,0].values.reshape(-1,1)

X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.4, random_state=42)

model = XGBRegressor(
    learning_rate=config.learning_rate,
    n_estimators = config.n_estimators,
    random_state = config.random_state)

model.fit(X_train, Y_train)

Y_pred = model.predict(X_test)

find_scores(Y_test, Y_pred, X_train)

extra_tree_configs = {
    "name": 'XGBRegressor',
    "method": "grid",
    "metric": {
        "name": "adj_r2",
        "goal": "maximize"
    },
    "parameters": {
        "learning_rate": {
            "values": [0.01, 0.03, 0.05, 0.07]
        },
        "n_estimators": {
            "values": [100,200,300]
        },
        "booster": {
            "values": ['gbtree','gblinear']
        },
        "eta": {
            "values": [0.01, 0.03, 0.05, 0.07]
        }
    }
}

sweep_id = wandb.sweep(sweep=extra_tree_configs, project="car_resale_value")
wandb.agent(sweep_id=sweep_id, function=XGB_regressor)

def LGBM_regressor():
    config_defaults = {
        'objective': 'root_mean_squared_error',

```

```

        'reg_sqrt': True,
        'metric': 'rmse',
        'random_state': 42
    }
wandb.init(config=config_defaults)
config = wandb.config

X = labeled.iloc[:,1:].values
Y = labeled.iloc[:,0].values.reshape(-1,1)
X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.4, random_state=42)
model = LGBMRegressor(
    learning_rate=config.learning_rate,
    n_estimators = config.n_estimators,
    random_state = config.random_state)
model.fit(X_train, Y_train)
Y_pred = model.predict(X_test)
find_scores(Y_test, Y_pred, X_train)
lgbm_configs = {
    "name": 'LGBMRegressor',
    "method": "grid",
    "metric": {
        "name": "adj_r2",
        "goal": "maximize"
    },
    "parameters": {
        "learning_rate": {
            "values": [0.01, 0.03, 0.05, 0.07]
        },
        "objective": {
            "values": ['root_mean_squared_error']
        },
        "boosting_type": {
            "values": ['gbdt','dart','goss','rf']
        },
        "reg_sqrt": {
            "values": [True]
        },
        "metric": {
            "values": ['rmse']
        },
        "n_estimators": {
            "values": [100,200,300]
        },
        "random_state": {
            "values": [42]
        }
    }

```

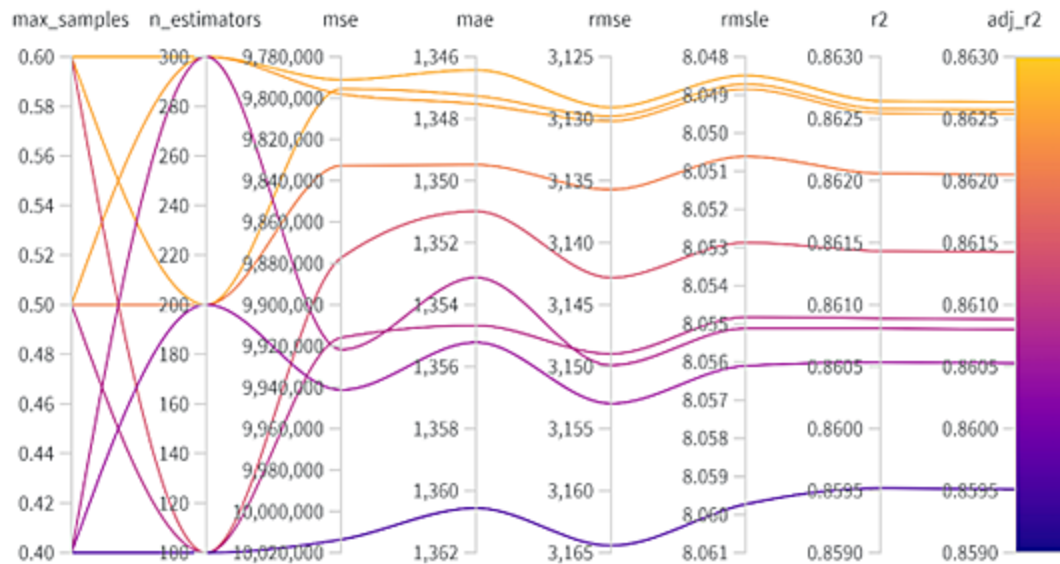
```

    }
  }
}
sweep_id = wandb.sweep(sweep=lgbm_configs, project="car_resale_value")
wandb.agent(sweep_id=sweep_id, function=LGBM_regressor)

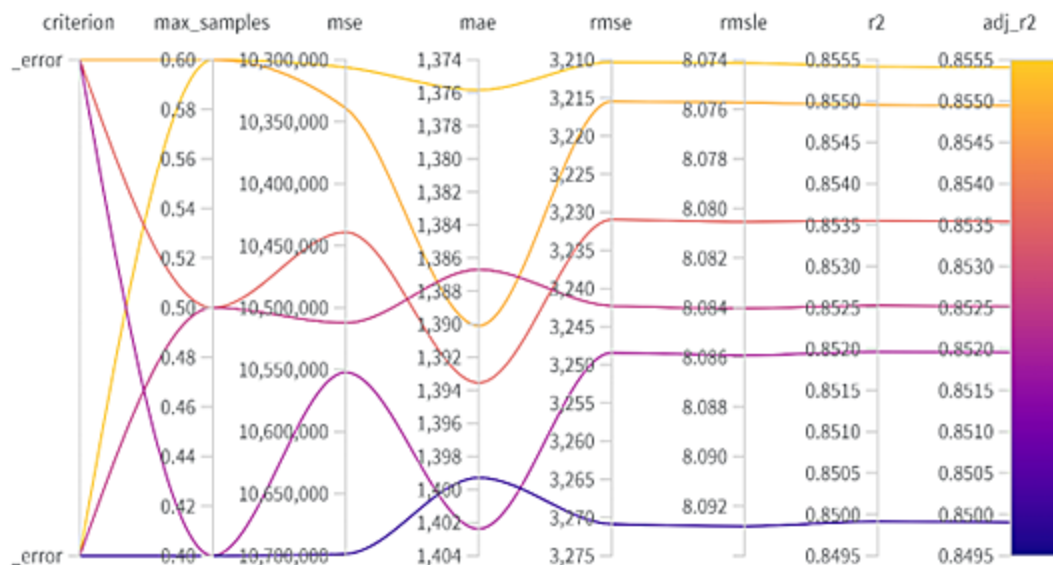
```

Output:

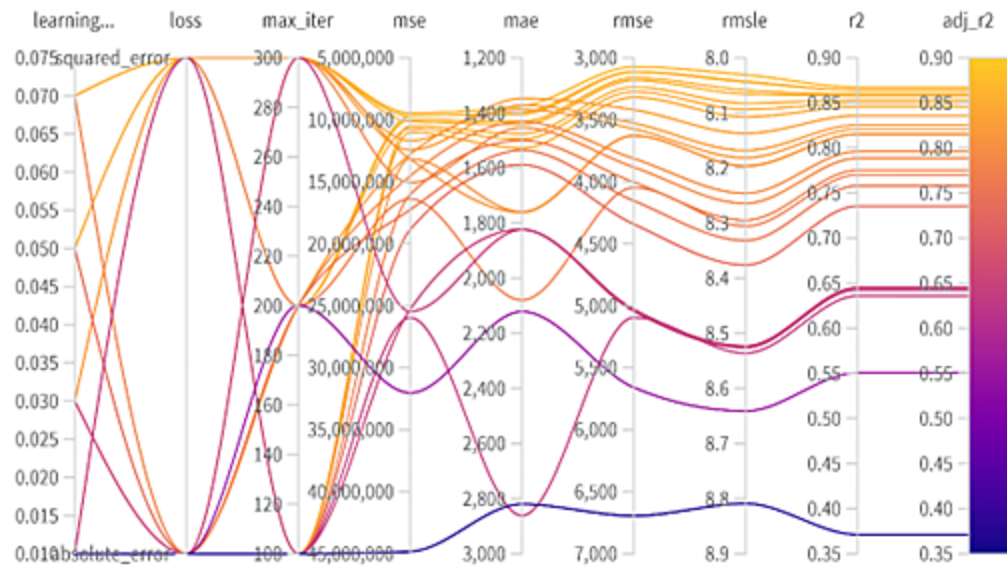
1) Bagging Regressor



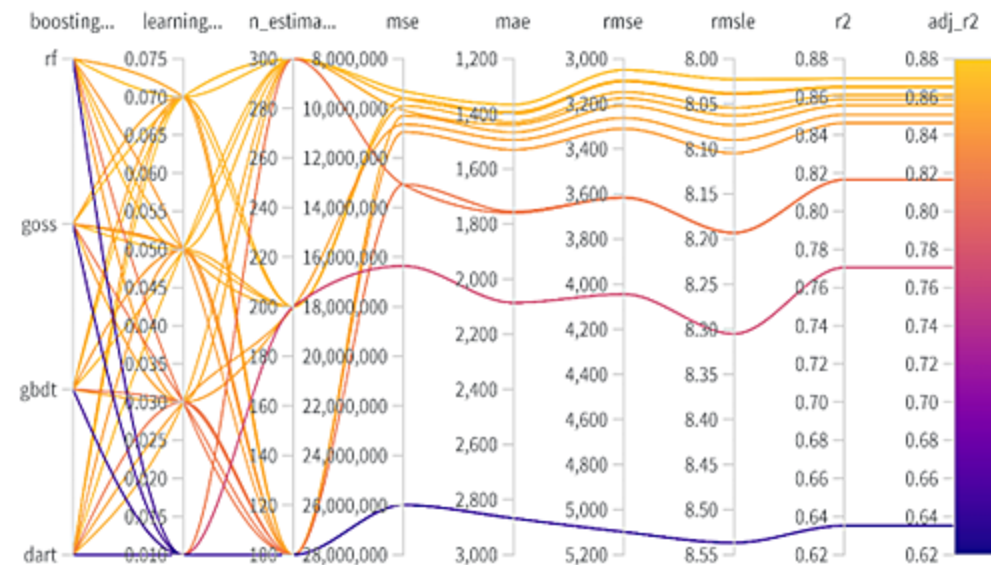
2) Extra Tree Regressor



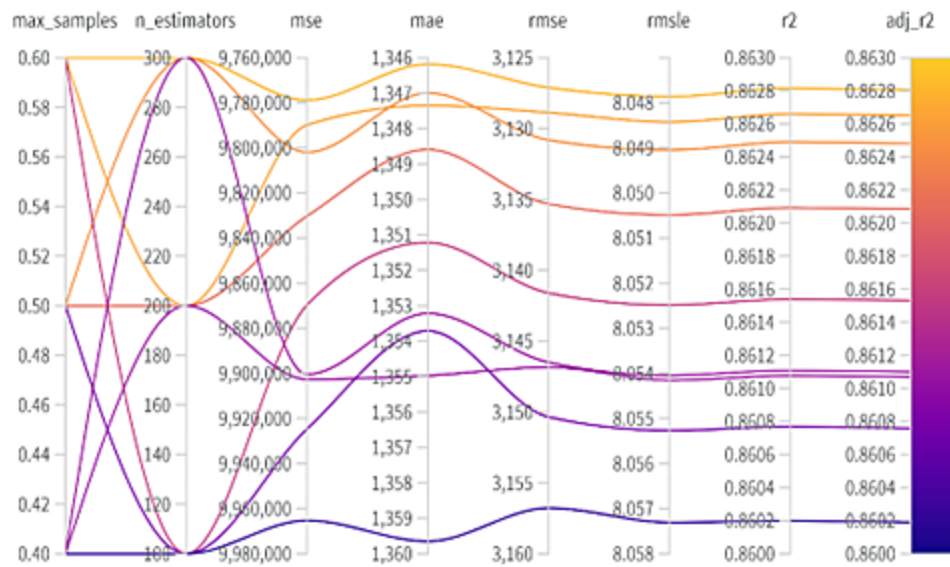
3) HOG Boosting Regressor



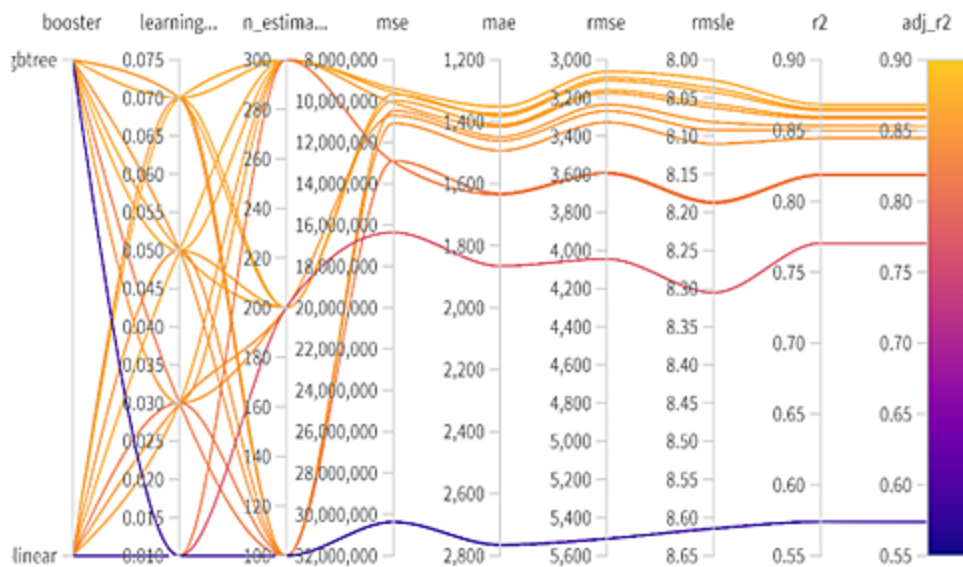
4) LGBM Regressor



5) Random Forest Regressor



6) XGB Regressor



7.5 Flask Integration

```
# Import Libraries
import pandas as pd
import numpy as np
from flask import Flask, render_template, Response, request
import pickle
from sklearn.preprocessing import LabelEncoder
import requests

# NOTE: you must manually set API_KEY below using information retrieved from your IBM
Cloud account.
API_KEY = "04ZW6LlRwAfofEU2VHPt69RKCWVc9U1o5LXkAU_66qA"
token_response = requests.post('https://iam.cloud.ibm.com/identity/token',
data={"apikey":API_KEY, "grant_type": 'urn:ibm:params:oauth:grant-type:apikey'})
mltoken = token_response.json()["access_token"]
header = {'Content-Type': 'application/json', 'Authorization': 'Bearer ' + mltoken}

app = Flask(__name__)#initiate flask app

def load_model(file='../Result/resale_model.sav'):#load the saved model
    return pickle.load(open(file, 'rb'))

@app.route('/')
def index():#main page
    return render_template('car.html')

@app.route('/predict_page')
def predict_page():#predicting page
    return render_template('value.html')

@app.route('/predict', methods=['GET','POST'])
def predict():
    reg_year = int(request.args.get('regyear'))
    powerps = float(request.args.get('powerps'))
    kms= float(request.args.get('kms'))
    reg_month = int(request.args.get('regmonth'))

    gearbox = request.args.get('geartype')
    damage = request.args.get('damage')
    model = request.args.get('model')
    brand = request.args.get('brand')
    fuel_type = request.args.get('fuelType')
    veh_type = request.args.get('vehicletype')
```

```

new_row = {'yearOfReg':reg_year, 'powerPS':powerps, 'kilometer':kms,
           'monthOfRegistration':reg_month, 'gearbox':gearbox,
           'notRepairedDamage':damage,
           'model':model, 'brand':brand, 'fuelType':fuel_type,
           'vehicletype':veh_type}

print(new_row)

new_df = pd.DataFrame(columns=['vehicletype','yearOfReg','gearbox',
                              'powerPS','model','kilometer','monthOfRegistration','fuelType',
                              'brand','notRepairedDamage'])
new_df = new_df.append(new_row, ignore_index=True)
labels = ['gearbox','notRepairedDamage','model','brand','fuelType','vehicletype']
mapper = {}

for i in labels:
    mapper[i] = LabelEncoder()
    mapper[i].classes = np.load('../Result/'+str('classes'+i+'.numpy'), allow_pickle=True)
    transform = mapper[i].fit_transform(new_df[i])
    new_df.loc[:,i+'_labels'] = pd.Series(transform, index=new_df.index)

    labeled = new_df[['yearOfReg','powerPS','kilometer','monthOfRegistration'] +
[x+'_labels' for x in labels]]
    X = labeled.values.tolist()
    print('\n\n', X)
    #predict = reg_model.predict(X)

# NOTE: manually define and pass the array(s) of values to be scored in the next line
payload_scoring = {"input_data": [{"fields": [['yearOfReg', 'powerPS', 'kilometer',
'monthOfRegistration','gearbox_labels', 'notRepairedDamage_labels',
'model_labels','brand_labels', 'fuelType_labels', 'vehicletype_labels']], "values": X}]}

response_scoring = requests.post('https://us-
south.ml.cloud.ibm.com/ml/v4/deployments/c0f74260-1f5f-43ad-8d71-
eb12ef099507/predictions?version=2022-11-13',
                                json=payload_scoring,
headers={'Authorization': 'Bearer ' + mltoken})
predictions = response_scoring.json()
print(response_scoring.json())
predict = predictions['predictions'][0]['values'][0][0]
print("Final prediction :",predict)

return render_template('predict.html',predict=predict)
if __name__ == '__main__':
    reg_model = load_model()#load the saved model
    app.run(host='localhost', debug=True, threaded=False)

```


8. TESTING

8.1 Test Cases Scenarios

1	Verify user is able to see home page?
2	Verify user is able to navigate to data entry page?
3	Verify user is able to see data entry page?
4	Verify user is able to enter values in the fields?
5	Verify user is able to navigate to output display page?
6	Verify user is able to view the output display page?
7	Verify user is able to view the car resale value output in the output display page?

8.2 User Acceptance Testing

Test case ID	Feature Type	Component	Test Scenario	Pre-Requirement	Steps To Execute	Test Data	Expected Result	Actual Result	Status	Comments	TC for Automation(Y/N)	BUG ID	Executed By
HomePage_TC_001	UI	Home Page	Verify all the UI elements in Home page rendered properly		1.Enter URL and click go 2.Verify all the UI elements displayed or not	-	All the UI elements rendered properly	Working as expected	Pass		N		Suhasnith
HomePage_TC_002	Functional	Home Page	Verify the Data Entry page can be reachable		1.Enter URL and click go 2.Verify all the UI elements displayed or not	-	User should navigate to Data Entry Page	Working as expected	Pass		N		Gobhinath
DataEntryPage_TC_001	UI	Data Entry Page	Verify all the UI elements in Data Entry page rendered properly		1.Enter URL and click go 2.Verify all the UI elements displayed or not 3.Press the Check Price button in the home page	-	All the UI elements rendered properly	Working as expected	Pass		N		Nabil Ahmed
DataEntryPage_TC_002	Functional	Data Entry Page	Verify user is able to enter all values		1.Enter URL and click go 2.Verify all the UI elements displayed or not 3.Press the Check Price button in the home page 4. Verify all the UI elements displayed or not 5. Verify if all values can be entered	2012 12 12 Manual Yes Golf Volvo n Petrol	User should be able to enter all values in data entry page	Working as expected	Pass		N		Avinash
DataEntryPage_TC_003	Functional	Data Entry Page	Verify the Output Display page can be reachable		1.Enter URL and click go 2.Verify all the UI elements displayed or not 3.Press the Check Price button in the home page 4. Verify all the UI elements displayed or not	-	User should navigate to Output Display Page	Working as expected	Pass		N		Gobhinath
OutputDisplayPage_TC_001	UI	Output Display Page	Verify all the UI elements in Output Display page rendered properly		1.Enter URL and click go 2.Verify all the UI elements displayed or not 3.Press the Check Price button in the home page 4. Verify all the UI elements displayed or not 5. Verify if all values can be entered	-	All the UI elements rendered properly	Working as expected	Pass		N		Nabil Ahmed
OutputDisplayPage_TC_002	Functional	Output Display Page	Verify user is able to get predicted result		1.Enter URL and click go 2.Verify all the UI elements displayed or not 3.Press the Check Price button in the home page 4. Verify all the UI elements displayed or not 5. Verify if all values can be entered 6. Press the submit Button 7. Verify all the UI elements displayed or not	-	Predicted Car Resale Value is displayed on the page	Working as expected	Pass		N		Avinash

8.3 UAT Report

8.3.1 Defect Analysis

This report shows the number of resolved or closed bugs at each severity level, and how they were resolved

Resolution	Severity 1	Severity 2	Severity 3	Severity 4	Subtotal
By Design	2	3	1	1	7
Duplicate	1	0	3	0	4
External	2	0	0	1	3
Fixed	2	2	1	2	7
Not Reproduced	0	0	1	0	1
Skipped	0	0	1	1	2
Won't Fix	0	2	2	1	5
Totals	7	7	9	6	29

8.3.2 Test Case Analysis

This report shows the number of test cases that have passed, failed, and untested

Section	Total Cases	Not Tested	Fail	Pass
Home Page	5	0	0	5
Data Entry Page	15	0	0	15
Output Page	4	0	0	4
Hyper Parameter Tuning	3	0	0	3
Final Model Building	2	0	0	2
Flask Application	10	0	0	10
Train Model on IBM	3	0	0	3
Final Report Output	4	0	0	4

9. RESULTS

9.1 Performance Metrics

S No.	Name	Description
1.	Metrics	Regression Model: LGBM Regressor MAE: 1327.55 MSE: 9492244.28 RMSE: 3080.95 RMSLE: 8.03 R2 Score: 0.8668 Adjusted R2 Score: 0.8668
2.	Tune the Model	Hyperparameter Tuning: 1) Learning Rate: [0.01, 0.03, 0.05, 0.07] 2) Boosting Type: ['gbdt', 'dart', 'goss', 'rf'] 3) Number of Estimators: [100, 200, 300] Validation Method: Grid Search Cross Validation Best Parameters: 1 . Learning Rate – 0.07 2 . Boosting Type – ‘gbdt’ 3 . Number of Estimators - 300

10. ADVANTAGES & DISADVANTAGES

Advantages

- a. Application is easy to use
- b. User Friendly
- c. No Cost
- d. No need to commission any agent to get car resale value estimate

Disadvantages

- a. User needs to fill every asked detail of the car
- b. Doesn't work for cars from different distributions
- c. Not always accurate

11. CONCLUSION

The increased prices of new cars and the financial incapability of the customers to buy them, used Car sales are on a global increase. Therefore, there is an urgent need for a Car Resale Value Prediction system which effectively determines the worthiness of the car in terms of cost. The proposed system is a web application that will help users to determine the accurate price of used cars.

12. FUTURE SCOPE

In future, large historical data of car price can be used to train the model, and which can help improve the estimation of the machine learning model. Moreover, we can build an application for mobile phone platforms like android, iOS for interacting with users. For better performance, we plan to judiciously design deep learning neural networks.

13. APPENDIX

Source Code

User Interface

car.html

```
<!DOCTYPE html>
<html lang="en" dir="ltr">
  <head>
    <meta charset="utf-8">
    <title>Car Resale Value Predicting Application</title>
    <link rel="icon" type="image/x-icon" href="../static/Images/favicon.ico">
    <link rel="stylesheet" href="../static/css/style.css">
    <link rel="stylesheet" href="https://cdnjs.cloudflare.com/ajax/libs/font-
awesome/4.7.0/css/font-awesome.min.css">
  </head>
  <body>
    <section class="header">
      <nav>
        <a href="/"></a>

      </nav>
      <div class="text-box">
        <h1>Car resale value Predictor</h1>
        <p>Best system to predict the amount of resale value based on the parameters provided by
the user .</p>
        <a href="/predict_page" class="visit-btn ">Check price</a>
      </div>
    </section>

  </body>
</html>
```

style.css

```
*{
  margin: 0;
  padding: 0;
}

.header{
  min-height: 100vh;
  width: 100%;
  background-image: linear-
gradient(rgba(25,30,30,0.7),rgba(25,30,30,0.7)),url(..Images/car1.png);
```

```

    background-position: center;
    background-size: cover;
    position: relative;
}
nav{
    display: flex;
    padding: 2% 6%;
    justify-content: space-between;
    align-items: center;
}
.nav-links{
    flex: 1;
    text-align: right;
}
.nav-links ul li{
    list-style: none;
    display: inline-block;
    padding: 8px 12px;
    position: relative;
}
.nav-links ul li a{
    color: white;
    text-decoration: none;
    font-size: 13px;
}
.text-box{
    text-align: center;
    position: relative;
    color: #FFE4C4;
    top: 50%;
}
.text-box h1{
    margin-top: 50px;
    font-size: 55px;
}
.text-box p{
    margin: 10px 0 40px;
    font-size: 15px;
}
.visit-btn{
    display: inline;
    border: 3px solid #fff;
    padding: 10px 14px;
    font-size: 15px;
    background: transparent;

```



```

color: white;
text-decoration:none;
}

```

value.html

```

<!DOCTYPE html>
<html lang="en" dir="ltr">
<head>
<link rel="stylesheet" href="../static/css/value.css">
<title>Car Resale Value Predicting Application</title>
<link rel="icon" type="image/x-icon" href="../static/Images/favicon.ico">
<script src="https://kit.fontawesome.com/b9b6bac803.js" crossorigin="anonymous"></script>
<link rel="stylesheet" href="https://cdnjs.cloudflare.com/ajax/libs/font-awesome/4.7.0/css/font-
awesome.min.css">
<style>
    table, th, td {
        padding: 10px;
    }
</style>
</head>
<body>
<div class="container">
    <div class="header">
        <h1>Get the Accurate Resale Value of Your Car</h1>
    </div>
    <form action="http://localhost:5000/predict" class="form">
    <div class="form-control">
        <label for="year" padding:10px>Registration year : </label>
        <input id="year" maxlength="50" name="regyear" type="text" autocomplete="off"/>
        <i class="fas fa-check-circle"></i>
        <i class="fas fa-exclamation-circle"></i>
        <span></span>
    </div>
    <div class="form-control">
        <label for="month">Registration Month : </label>
        <input id="month" maxlength="50" name="regmonth" type="text" autocomplete="off"/>
        <i class="fas fa-check-circle"></i>
        <i class="fas fa-exclamation-circle"></i>
        <span></span>
    </div>
    <div class="form-control">
        <label for="power">Power of car in PS: </label>
        <input id="power" maxlength="50" name="powerps" type="text" autocomplete="off"/>
        <i class="fas fa-check-circle"></i>
        <i class="fas fa-exclamation-circle"></i>

```

```

<span></span>
</div>
<div class="form-control">
<label for="kilometer">Kilometers that car have driven : </label>
<input id="kilometer" maxlength="50" name="kms" type="text" autocomplete="off"/>
<i class="fas fa-check-circle"></i>
<i class="fas fa-exclamation-circle"></i>
<span></span>
</div>
<div class="form-control">
    <h3>Gear Type</h3>
    <table style="width:50%">
        <tr>
            <th></th>
            <th></th>

        </tr>
        <tr>
            <td>Manual</td>
            <td><input type="radio" name="geartype" value="manual" id="manual"
/></td>
        </tr>
        <tr>
            <td>Automatic</td>
            <td><input type="radio" name="geartype" value="automatic"
id="automatic" /></td>
        </tr>
        <tr>
            <td>Not mentioned</td>
            <td><input type="radio" name="geartype" value="not-declared" id="not"
/></td>
        </tr>
    </table>
    <i class="fas fa-check-circle"></i>
    <i class="fas fa-exclamation-circle"></i>
    <span></span>
</div>
<div class="form-control">
    <h3>Your car is repaired or damaged :</h3>
    <table style="width:50%">
        <tr>
            <th></th>
            <th></th>

        </tr>

```

```

        <tr>
            <td>Yes</td>
            <td><input type="radio" name="damage" value="yes"
id="yes"/></td>
        </tr>
        <tr>
            <td>No</td>
            <td><input type="radio" name="damage" value="no"
id="no"/></td>
        </tr>
        <tr>
            <td>Not Declared</td>
            <td><input type="radio" name="damage" value="not-declared"
id="notdec"/></td>
        </tr>
    </table>
    <i class="fas fa-check-circle"></i>
    <i class="fas fa-exclamation-circle"></i>
    <span></span>
</div>
<div class="form-control">
    <label for="model">Model Type : </label>
    <select name="model" id="model">
        <option value="" disabled selected hidden>Choose Model Name...</option>
    <option value="golf">Golf </option>
    <option value="grand">Grand </option>
    <option value="fabia">Fabia </option>
    <option value="3er">3er </option>
    <option value="2_reihe">2 Reihe </option>
    <option value="andere">Andere </option>
    <option value="c_max">C Max </option>
    <option value="3_reihe">3 Reihe </option>
    <option value="passat">Passat </option>
    <option value="navara">Navara </option>
    <option value="ka">Ka </option>
    <option value="polo">Polo </option>
    <option value="twingo">Twingo </option>
    <option value="a_klasse">A klasse </option>
    <option value="scirocco">Scirocco </option>
    <option value="5er">5er </option>
    <option value="meriva">Meriva </option>
    <option value="arosa">Arosa </option>
    <option value="c4">C4 </option>
    <option value="civic">Civic </option>
    <option value="transporter">Transporter </option>

```

<option value="punto">Punto </option>
<option value="e_klasse">E Klasse </option>
<option value="clio">Clio </option>
<option value="kadett">Kadett </option>
<option value="kangoo">Kangoo </option>
<option value="corsa">Corsa </option>
<option value="one">One </option>
<option value="fortwo">Fortwo </option>
<option value="1er">1er </option>
<option value="b_klasse">B Klasse </option>
<option value="signum">Signum </option>
<option value="astra">Astra </option>
<option value="a8">A8 </option>
<option value="jetta">Jetta </option>
<option value="fiesta">Fiesta </option>
<option value="c_klasse">C Klasse </option>
<option value="micra">Micra </option>
<option value="vito">Vito </option>
<option value="sprinter">Sprinter </option>
<option value="156">156 </option>
<option value="escort">Escort </option>
<option value="forester">Forester </option>
<option value="xc_reihe">Xc Reihe </option>
<option value="scenic">Scenic </option>
<option value="a4">A4 </option>
<option value="a1">A1 </option>
<option value="insignia">Insignia </option>
<option value="combo">Combo </option>
<option value="focus">Focus </option>
<option value="tt">Tt </option>
<option value="a6">A6 </option>
<option value="jazz">Jazz </option>
<option value="omega">Omega </option>
<option value="slk">Slk </option>
<option value="7er">7er </option>
<option value="80">80 </option>
<option value="147">147 </option>
<option value="glk">Glk </option>
<option value="100">100 </option>
<option value="z_reihe">Z Reihe </option>
<option value="sportage">Sportage </option>
<option value="sorento">Sorento </option>
<option value="v40">V40 </option>
<option value="5er">5er </option>
<option value="ibiza">Ibiza </option>

<option value="3er">3er </option>
<option value="mustang">Mustang </option>
<option value="eos">Eos </option>
<option value="touran">Touran </option>
<option value="getz">Getz </option>
<option value="a3">A3 </option>
<option value="almera">Almera </option>
<option value="megane">Megane </option>
<option value="7er">7er </option>
<option value="1er">1er </option>
<option value="lupo">Lupo </option>
<option value="r19">R19 </option>
<option value="zafira">Zafira </option>
<option value="caddy">Caddy </option>
<option value="2_reihe">2 Reihe </option>
<option value="mondeo">Mondeo </option>
<option value="cordoba">Cordoba </option>
<option value="colt">Colt </option>
<option value="impreza">Impreza </option>
<option value="vectra">Vectra </option>
<option value="berlingo">Berlingo </option>
<option value="80">80 </option>
<option value="m_klasse">M Klasse </option>
<option value="tiguan">Tiguan </option>
<option value="i_reihe">I Reihe </option>
<option value="espace">Espace </option>
<option value="sharan">Sharan </option>
<option value="6_reihe">6 Reihe </option>
<option value="panda">Panda </option>
<option value="up">Up </option>
<option value="seicento">Seicento </option>
<option value="ceed">Ceed </option>
<option value="5_reihe">5 Reihe </option>
<option value="yeti">Yeti </option>
<option value="octavia">Octavia </option>
<option value="mii">Mii </option>
<option value="rx_reihe">Rx Reihe </option>
<option value="6er">6er </option>
<option value="modus">Modus </option>
<option value="fox">Fox </option>
<option value="matiz">Matiz </option>
<option value="beetle">Beetle </option>
<option value="c1">C1 </option>
<option value="rio">Rio </option>
<option value="touareg">Touareg </option>

<option value="logan">Logan </option>
<option value="spider">Spider </option>
<option value="cuore">Cuore </option>
<option value="s_max">S Max </option>
<option value="a2">A2 </option>
<option value="x_reihe">X Reihe </option>
<option value="a5">A5 </option>
<option value="galaxy">Galaxy </option>
<option value="c3">C3 </option>
<option value="viano">Viano </option>
<option value="s_klasse">S Klasse </option>
<option value="1_reihe">1 Reihe </option>
<option value="avensis">Avensis </option>
<option value="sl">Sl </option>
<option value="roomster">Roomster </option>
<option value="q5">Q5 </option>
<option value="kaefer">Kaefer </option>
<option value="santa">Santa </option>
<option value="cooper">Cooper </option>
<option value="leon">Leon </option>
<option value="4_reihe">4 Reihe </option>
<option value="500">500 </option>
<option value="laguna">Laguna </option>
<option value="ptcruiser">Ptcruiser </option>
<option value="clk">Clk </option>
<option value="primera">Primera </option>
<option value="exeo">Exeo </option>
<option value="159">159 </option>
<option value="transit">Transit </option>
<option value="juke">Juke </option>
<option value="qashqai">Qashqai </option>
<option value="carisma">Carisma </option>
<option value="accord">Accord </option>
<option value="corolla">Corolla </option>
<option value="lanos">Lanos </option>
<option value="phaeton">Phaeton </option>
<option value="boxster">Boxster </option>
<option value="verso">Verso </option>
<option value="swift">Swift </option>
<option value="rav">Rav </option>
<option value="kuga">Kuga </option>
<option value="picanto">Picanto </option>
<option value="kalos">Kalos </option>
<option value="superb">Superb </option>
<option value="stilo">Stilo </option>

<option value="alhabbra">Alhabbra </option>
<option value="911">911 </option>
<option value="mx_reihe">Mx Reihe </option>
<option value="m_reihe">M Reihe </option>
<option value="roadster">Roadster </option>
<option value="ypsilon">Ypsilon </option>
<option value="cayenne">Cayenne </option>
<option value="galant">Galant </option>
<option value="justy">Justy </option>
<option value="90">90 </option>
<option value="sirion">Sirion </option>
<option value="crossfire">Crossfire </option>
<option value="6_reihe">6 Reihe </option>
<option value="agila">Agila </option>
<option value="duster">Duster </option>
<option value="cr_reihe">Cr Reihe </option>
<option value="v50">V50 </option>
<option value="discovery">Discovery </option>
<option value="c_reihe">C Reihe </option>
<option value="v_klasse">V Klasse </option>
<option value="yaris">Yaris </option>
<option value="c5">C5 </option>
<option value="aygo">Aygo </option>
<option value="cc">Cc </option>
<option value="carnival">Carnival </option>
<option value="fusion">Fusion </option>
<option value="bora">Bora </option>
<option value="forfour">Forfour </option>
<option value="100">100 </option>
<option value="cl">Cl </option>
<option value="tigra">Tigra </option>
<option value="156">156 </option>
<option value="300c">300c </option>
<option value="100">100 </option>
<option value="147">147 </option>
<option value="q3">Q3 </option>
<option value="spark">Spark </option>
<option value="v70">V70 </option>
<option value="x_type">X Type </option>
<option value="5_reihe">5 Reihe </option>
<option value="ducato">Ducato </option>
<option value="s_type">S Type </option>
<option value="x_trail">X Trail </option>
<option value="toledo">Toledo </option>
<option value="altea">Altea </option>

<option value="7er">7er </option>
 <option value="voyager">Voyager </option>
 <option value="calibra">Calibra </option>
 <option value="bravo">Bravo </option>
 <option value="range_rover">Range Rover </option>
 <option value="antara">Antara </option>
 <option value="tucson">Tucson </option>
 <option value="q7">Q7 </option>
 <option value="citigo">Citigo </option>
 <option value="jimny">Jimny </option>
 <option value="cx_reihe">Cx Reihe </option>
 <option value="wrangler">Wrangler </option>
 <option value="lybra">Lybra </option>
 <option value="range_rover_sport">Range Rover Sport </option>
 <option value="lancer">Lancer </option>
 <option value="159">159 </option>
 <option value="freelander">Freelander </option>
 <option value="captiva">Captiva </option>
 <option value="c2">C2 </option>
 <option value="500">500 </option>
 <option value="range_rover_evoque">Range Rover Evoque </option>
 <option value="sander0">Sander0 </option>
 <option value="note">Note </option>
 <option value="900">900 </option>
 <option value="147">147 </option>
 <option value="defender">Defender </option>
 <option value="cherokee">Cherokee </option>
 <option value="clubman">Clubman </option>
 <option value="samara">Samara </option>
 <option value="2_reihe">2 Reihe </option>
 <option value="1er">1er </option>
 <option value="3er">3er </option>
 <option value="601">601 </option>
 <option value="3_reihe">3 Reihe </option>
 <option value="4_reihe">4 Reihe </option>
 <option value="5er">5er </option>
 <option value="6_reihe">6 Reihe </option>
 <option value="legacy">Legacy </option>
 <option value="pajero">Pajero </option>
 <option value="auris">Auris </option>
 <option value="niva">Niva </option>
 <option value="5_reihe">5 Reihe </option>
 <option value="s60">S60 </option>
 <option value="nubira">Nubira </option>
 <option value="vivaro">Vivaro </option>


```

<option value="g_klasse">G Klasse </option>
<option value="lodgy">Lodgy </option>
<option value="850">850 </option>
<option value="serie_2">Serie 2 </option>
<option value="6er">6er </option>
<option value="charade">Charade </option>
<option value="croma">Croma </option>
<option value="outlander">Outlander </option>
<option value="gl">Gl </option>
<option value="doblo">Doblo </option>
<option value="musa">Musa </option>
<option value="amarok">Amarok </option>
<option value="156">156 </option>
<option value="move">Move </option>
<option value="9000">9000 </option>
<option value="v60">V60 </option>
<option value="145">145 </option>
<option value="aveo">Aveo </option>
<option value="200">200 </option>
<option value="300c">300c </option>
<option value="b_max">B Max </option>
<option value="delta">Delta </option>
<option value="terios">Terios </option>
<option value="rangerover">RangeRover </option>
<option value="90">90 </option>
<option value="materia">Materia </option>
<option value="kalina">Kalina </option>
<option value="elefantino">Elefantino </option>
<option value="i3">I3 </option>
<option value="kappa">Kappa </option>
<option value="serie_3">Serie 3 </option>
<option value="48429">48429 </option>
<option value="serie_1">Serie 1 </option>
<option value="discovery_sport">Discovery Sport </option>
</select>
<i class="fas fa-check-circle"></i>
<i class="fas fa-exclamation-circle"></i>
<span></span>
</div>
<div class="form-control">
<label for="brand">Brand :</label>
<select name="brand" id="brand">
<option value="" disabled selected hidden>Choose Brand Name...</option>
<option value="volkswagen">Volkswagen </option>
<option value="audi">Audi </option>

```

```

<option value="jeep">Jeep </option>
<option value="skoda">Skoda </option>
<option value="bmw">Bmw </option>
<option value="peugeot">Peugeot </option>
<option value="ford">Ford </option>
<option value="mazda">Mazda </option>
<option value="nissan">Nissan </option>
<option value="renault">Renault </option>
<option value="mercedes_benz">Mercedes Benz </option>
<option value="opel">Opel </option>
<option value="seat">Seat </option>
<option value="citroen">Citroen </option>
<option value="honda">Honda </option>
<option value="fiat">Fiat </option>
<option value="mini">Mini </option>
<option value="smart">Smart </option>
<option value="hyundai">Hyundai </option>
<option value="sonstige_autos">Sonstige Autos </option>
<option value="alfa_romeo">Alfa Romeo </option>
<option value="subaru">Subaru </option>
<option value="volvo">Volvo </option>
<option value="mitsubishi">Mitsubishi </option>
<option value="kia">Kia </option>
<option value="suzuki">Suzuki </option>
<option value="lancia">Lancia </option>
<option value="porsche">Porsche </option>
<option value="toyota">Toyota </option>
<option value="chevrolet">Chevrolet </option>
<option value="dacia">Dacia </option>
<option value="daihatsu">Daihatsu </option>
<option value="trabant">Trabant </option>
<option value="saab">Saab </option>
<option value="chrysler">Chrysler </option>
<option value="jaguar">Jaguar </option>
<option value="daewoo">Daewoo </option>
<option value="rover">Rover </option>
<option value="land_rover">Land Rover </option>
<option value="lada">Lada </option>
</select>
<i class="fas fa-check-circle"></i>
<i class="fas fa-exclamation-circle"></i>
<span></span>
</div>
<div class="form-control">
<label for="fuelType">Fuel Type :</label>

```

```

        <select name="fuelType" id="fuel">
        <option value="" disabled selected hidden>Choose Fuel Type...</option>
        <option value="petrol"> Petrol </option>
        <option value="diesel"> Diesel </option>
        <option value="not-declared"> Not Declared </option>
        <option value="lpg">LPG </option>
        <option value="cng">CNG </option>
        <option value="hybrid">Hybrid </option>
        <option value="others">Others </option>
        <option value="electric">Electric </option>
        </select>
        <i class="fas fa-check-circle"></i>
        <i class="fas fa-exclamation-circle"></i>
        <span></span>
    </div>
    <div class="form-control">
    <label for="vehicletype">Vehicle type: </label>
    <select name="vehicletype" id="vehicle" >
    <option value="" disabled selected hidden>Choose Vehicle Type...</option>
    <option value="coupe">Coupe </option>
    <option value="suv">SUV </option>
    <option value="kleinwagen">Kleinwagen </option>
    <option value="limousine">Limousine </option>
    <option value="cabrio">Cabrio </option>
    <option value="bus">Bus </option>
    <option value="kombi">Kombi </option>
    <option value="andere">Andere </option>
    <option value="volkswagen">Volkswagen </option>
    </select>
    <i class="fas fa-check-circle"></i>
    <i class="fas fa-exclamation-circle"></i>
    <span></span>
    </div>
    <input type="submit" id="submit"></input>
</form>
</div>
</body>
</html>

```

```

value.css
*{
padding:0px;
margin:0;
box-sizing:border-box;
font-family: cursive;

```

```
font-weight: bold;
color: #E74C3C;
}

body{
  background-image: linear-
gradient(rgba(25,30,30,0.7),rgba(25,30,30,0.7)),url(../Images/car2.png);
  min-height:100vh;
  display:flex;
  justify-content:center;
  align-items:center;
}

.header{
  color:Black;
  text-align:center;
  padding:10px 0px 10px 100px;
}

#model{
  width:500px;
  color: black;
}

#brand{
  width:500px;
  color: black;
}

#fuel{
  width:500px;
  color: black;
}

#vehicle{
  width:500px;
  color: black;
}

.form{
  padding:30px 40px;
}

.form-control{
  margin-bottom:10px;
```

```

padding-bottom: 20px;
position: relative;
margin-left: 100px;
}

.form-control label{
display: block;
margin-bottom: 5px;
}

.form-control input{
border: 2px solid #f0f0f0;
width: 80%;
font-size: .8rem;
padding: 5px;
display: inline-table;
}

.form-control i{
position: absolute;
right: 20px;
top: 35px;
visibility: hidden;
}

.form-control span{
position: absolute;
left: 0;
bottom: 0;
visibility: hidden;
font-weight: bolder;
font-style: italic;
font-size: 1rem;
}

.form-control.success input{
border-color: #2ecc71;
}

.form-control.error input{
border-color: #e74c3c;
}

.form-control.error span{
color: red;
}

```

```

        visibility:visible;
    }

    .form-control.success i.fa-check-circle {
        border-color:#2ecc71;
        visibility:visible;
    }

    .form-control.error i.fa-exclamation-circle {
        border-color:#e73c3c;
        visibility:visible;
    }

```

```

.form #submit{
    background-color:#E74C3C;
    border:none;
    outline:none;
    color:white;
    width:500px;
    border-radius:4px;
    padding:10px;
    cursor:pointer;
    transition:all .5s;
    font-size:1rem;
    margin-left: 100px;
}

```

```

.form #submit:hover{
    background-color:#6441a5;
}

```

```

.form-control #manual{
    padding-bottom: 20%;
}

```

```

predict.html
<!DOCTYPE html>
<html lang="en">
<head>
    <meta charset="UTF-8">
    <meta http-equiv="X-UA-Compatible" content="IE=edge">
    <meta name="viewport" content="width=device-width, initial-scale=1.0">
    <link rel="stylesheet" href="../static/css/predict.css">

```

```

<title>Car Resale Value Predicting Application</title>
<link rel="icon" type="image/x-icon" href="../static/Images/favicon.ico">
</head>
<body>
    <section class="header">
        <nav>
            <a href="/"></a>
        </nav>
        <div class="text-box">
            <h1>The Predicted Car Resale Value is </h1>
                <h1>{{predict}}</h1>
            </div>
        </section>

</body>
</html>

```

```

predict.css
.header{
    min-height: 100vh;
    width: 100%;
    background-image: linear-
gradient(rgba(25,30,30,0.7),rgba(25,30,30,0.7)),url(..Images/car3.jpg);
    background-position: center;
    background-size: cover;
    position: relative;
}

```

```

.text-box{
    text-align: center;
    position: relative;
    color: #FFE4C4;
    top:50%;
}
.text-box h1{
    margin-top: 50px;
    font-size: 55px;
}

```

```

.text-box p{
    margin: 10px 0 40px;
    font-size: 15px;
}

```

```
body{
    margin: 0;
}
```

```
nav{
    display: flex;
    padding: 2% 6%;
    justify-content: space-between;
    align-items: center;
}
```

```
app.py
# Import Libraries
import pandas as pd
import numpy as np
from flask import Flask, render_template, Response, request
import pickle
from sklearn.preprocessing import LabelEncoder
import requests
```

```
# NOTE: you must manually set API_KEY below using information retrieved from your IBM
Cloud account.
```

```
API_KEY = "04ZW6LlLwAfofEU2VHPt69RKCWVc9U1o5LXkAU_66qA"
token_response = requests.post('https://iam.cloud.ibm.com/identity/token',
data={"apikey":API_KEY, "grant_type": 'urn:ibm:params:oauth:grant-type:apikey'})
mltoken = token_response.json()["access_token"]
header = {'Content-Type': 'application/json', 'Authorization': 'Bearer ' + mltoken}
```

```
app = Flask(__name__)#initiate flask app
```

```
def load_model(file='../Result/resale_model.sav'):#load the saved model
    return pickle.load(open(file, 'rb'))
```

```
@app.route('/')
def index():#main page
    return render_template('car.html')
```

```
@app.route('/predict_page')
def predict_page():#predicting page
    return render_template('value.html')
```

```
@app.route('/predict', methods=['GET','POST'])
def predict():
    reg_year = int(request.args.get('regyear'))
```



```

powerps = float(request.args.get('powerps'))
kms= float(request.args.get('kms'))
reg_month = int(request.args.get('regmonth'))

gearbox = request.args.get('geartype')
damage = request.args.get('damage')
model = request.args.get('model')
brand = request.args.get('brand')
fuel_type = request.args.get('fuelType')
veh_type = request.args.get('vehicletype')

new_row = {'yearOfReg':reg_year, 'powerPS':powerps, 'kilometer':kms,
           'monthOfRegistration':reg_month, 'gearbox':gearbox,
           'notRepairedDamage':damage,
           'model':model, 'brand':brand, 'fuelType':fuel_type,
           'vehicletype':veh_type}

print(new_row)

new_df = pd.DataFrame(columns=['vehicletype','yearOfReg','gearbox',
                              'powerPS','model','kilometer','monthOfRegistration','fuelType',
                              'brand','notRepairedDamage'])
new_df = new_df.append(new_row, ignore_index=True)
labels = ['gearbox','notRepairedDamage','model','brand','fuelType','vehicletype']
mapper = {}

for i in labels:
    mapper[i] = LabelEncoder()
    mapper[i].classes = np.load('../Result/'+str('classes'+i+'.npy'), allow_pickle=True)
    transform = mapper[i].fit_transform(new_df[i])
    new_df.loc[:,i+'_labels'] = pd.Series(transform, index=new_df.index)

labeled = new_df[['yearOfReg','powerPS','kilometer','monthOfRegistration'] +
[x+'_labels' for x in labels]]

X = labeled.values.tolist()
print("\n\n", X)
#predict = reg_model.predict(X)

# NOTE: manually define and pass the array(s) of values to be scored in the next line
payload_scoring = {"input_data": [{"fields": [['yearOfReg', 'powerPS', 'kilometer',
'monthOfRegistration','gearbox_labels', 'notRepairedDamage_labels',
'model_labels','brand_labels', 'fuelType_labels', 'vehicletype_labels']], "values": X}]}

response_scoring = requests.post('https://us-
```

```

south.ml.cloud.ibm.com/ml/v4/deployments/c0f74260-1f5f-43ad-8d71-
eb12ef099507/predictions?version=2022-11-13', json=payload_scoring,
headers={'Authorization': 'Bearer ' + mltoken})
    predictions = response_scoring.json()
    print(response_scoring.json())
    predict = predictions['predictions'][0]['values'][0][0]
    print("Final prediction :",predict)

    return render_template('predict.html',predict=predict)

if __name__ == '__main__':
    reg_model = load_model()#load the saved model
    app.run(host='localhost', debug=True, threaded=False)

car_resale_value_prediction_modelling.py
import pandas as pd
import numpy as np
from sklearn.preprocessing import LabelEncoder
from sklearn.model_selection import train_test_split, GridSearchCV
from sklearn.metrics import mean_absolute_error, mean_squared_error, r2_score
import pickle
import wandb

#regression models
from sklearn.ensemble import BaggingRegressor, RandomForestRegressor,
HistGradientBoostingRegressor, ExtraTreesRegressor
from xgboost.sklearn import XGBRegressor
from lightgbm import LGBMRegressor

wandb.login(key='b75e0564aba32dce859c60044418df71ce7389a8')

data = pd.read_csv('../input/naalaiya-thiran/Preprocessed/autos_preprocessed.csv', header=0,
sep=',', encoding='Latin1')

labels = ['gearbox', 'notRepairedDamage', 'model', 'brand', 'fuelType', 'vehicleType']

mapper = {}
for i in labels:
    mapper[i] = LabelEncoder()
    mapper[i].fit(data[i])
    tr = mapper[i].transform(data[i])
    np.save(str('classes'+i+'.npy'), mapper[i].classes_)
    data.loc[:, i+'_labels'] = pd.Series(tr, index=data.index)

labeled = data[['price', 'yearOfRegistration', 'powerPS', 'kilometer', 'monthOfRegistration']]

```

```

        +[x+"_labels" for x in labels]]

print(labeled.columns)

def find_scores(Y_actual, Y_pred, X_train):
    mae = mean_absolute_error(Y_actual, Y_pred)
    mse = mean_squared_error(Y_actual, Y_pred)
    rmse = np.sqrt(mse)
    rmsle = np.log(rmse)
    r2 = r2_score(Y_actual, Y_pred)
    n, k = X_train.shape
    adj_r2_score = 1 - ((1-r2)*(n-1)/(n-k-1))

    wandb.log({"mae": mae, "mse": mse, 'rmse':rmse, 'rmsle':rmsle, 'r2':r2, 'adj_r2':adj_r2_score})

def bagging_regressor():
    config_defaults = {
        'n_estimators':100,
        'max_samples':0.4,
        'bootstrap':True,
        'random_state':42
    }
    wandb.init(config=config_defaults)
    config = wandb.config

    X = labeled.iloc[:,1:].values
    Y = labeled.iloc[:,0].values.reshape(-1,1)

    X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.4, random_state=42)

    model = BaggingRegressor(
        n_estimators=config.n_estimators,
        bootstrap=config.bootstrap,
        max_samples=config.max_samples,
        random_state = config.random_state)

    model.fit(X_train, Y_train)

    Y_pred = model.predict(X_test)

    find_scores(Y_test, Y_pred, X_train)

bagging_regressor_configs = {
    "name":'BaggingRegressor',
    "method": "grid",

```

```

"metric": {
    "name": "adj_r2",
    "goal": "maximize"
},
"parameters": {
    "n_estimators": {
        "values": [100, 200, 300]
    },
    "max_samples": {
        "values": [0.4, 0.5, 0.6]
    }
}
}

```

```

sweep_id = wandb.sweep(sweep=bagging_regressor_configs, project="car_resale_value")
wandb.agent(sweep_id=sweep_id, function=bagging_regressor)

```

```

def random_forest_regressor():
    config_defaults = {
        'n_estimators': 100,
        'max_samples': 0.4,
        'criterion': 'squared_error',
        'bootstrap': True,
        'random_state': 42
    }
    wandb.init(config=config_defaults)
    config = wandb.config

    X = labeled.iloc[:, 1:].values
    Y = labeled.iloc[:, 0].values.reshape(-1, 1)

```

```

X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.4, random_state=42)

```

```

model = RandomForestRegressor(
    n_estimators=config.n_estimators,
    criterion = config.criterion,
    bootstrap=config.bootstrap,
    max_samples=config.max_samples,
    random_state = config.random_state)

```

```

model.fit(X_train, Y_train)

```

```

Y_pred = model.predict(X_test)

```

```

find_scores(Y_test, Y_pred, X_train)

```

```

random_forest_configs = {
    "name": 'RandomForestRegressor',
    "method": "grid",
    "metric": {
        "name": "adj_r2",
        "goal": "maximize"
    },
    "parameters": {
        "n_estimators": {
            "values": [100, 200, 300]
        },
        "max_samples": {
            "values": [0.4, 0.5, 0.6]
        }
    }
}

```

```

sweep_id = wandb.sweep(sweep=random_forest_configs, project="car_resale_value")
wandb.agent(sweep_id=sweep_id, function=random_forest_regressor)

```

```

def hist_gradient_boost_regressor():

```

```

    config_defaults = {
        'loss': 'squared_error',
        'learning_rate': 0.1,
        'max_iter': 100,
        'random_state': 42
    }

```

```

    wandb.init(config=config_defaults)
    config = wandb.config

```

```

    X = labeled.iloc[:, 1:].values
    Y = labeled.iloc[:, 0].values.reshape(-1, 1)

```

```

    X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.4, random_state=42)

```

```

    model = HistGradientBoostingRegressor(
        loss=config.loss,
        learning_rate=config.learning_rate,
        max_iter=config.max_iter,
        random_state=config.random_state)

```

```

    model.fit(X_train, Y_train)

```

```

    Y_pred = model.predict(X_test)

```

```

find_scores(Y_test, Y_pred, X_train)

hist_gradient_boost_configs = {
    "name": 'HistGradientBoostingRegressor',
    "method": "grid",
    "metric": {
        "name": "adj_r2",
        "goal": "maximize"
    },
    "parameters": {
        "loss": {
            "values": ['squared_error', 'absolute_error']
        },
        "learning_rate": {
            "values": [0.01, 0.03, 0.05, 0.07]
        },
        "max_iter": {
            "values": [100, 200, 300]
        },
        "random_state": {
            "values": [42]
        }
    }
}

sweep_id = wandb.sweep(sweep=hist_gradient_boost_configs, project="car_resale_value")
wandb.agent(sweep_id=sweep_id, function=hist_gradient_boost_regressor)

def extra_tree_regressor():
    config_defaults = {
        'criterion': 'squared_error',
        'max_samples': 0.4,
        'bootstrap': True,
        'random_state': 42
    }
    wandb.init(config=config_defaults)
    config = wandb.config

    X = labeled.iloc[:, 1:].values
    Y = labeled.iloc[:, 0].values.reshape(-1, 1)

    X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.4, random_state=42)

    model = ExtraTreesRegressor(

```

```

        criterion=config.criterion,
        bootstrap = config.bootstrap,
        max_samples=config.max_samples,
        random_state = config.random_state)

model.fit(X_train, Y_train)

Y_pred = model.predict(X_test)

find_scores(Y_test, Y_pred, X_train)

extra_tree_configs = {
    "name": 'ExtraTreesRegressor',
    "method": "grid",
    "metric": {
        "name": "adj_r2",
        "goal": "maximize"
    },
    "parameters": {
        "criterion": {
            "values": ['squared_error', 'absolute_error']
        },
        "max_samples": {
            "values": [0.4, 0.5, 0.6]
        }
    }
}

sweep_id = wandb.sweep(sweep=extra_tree_configs, project="car_resale_value")
wandb.agent(sweep_id=sweep_id, function=extra_tree_regressor)

def XGB_regressor():
    config_defaults = {
        'learning_rate': 0.1,
        'n_estimators': 500,
        'booster': 'gbtree',
        'eta': 0.01,
        'random_state': 42
    }
    wandb.init(config=config_defaults)
    config = wandb.config

    X = labeled.iloc[:, 1:].values
    Y = labeled.iloc[:, 0].values.reshape(-1, 1)

```

```

X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.4, random_state=42)

model = XGBRegressor(
    learning_rate=config.learning_rate,
    n_estimators = config.n_estimators,
    random_state = config.random_state)

model.fit(X_train, Y_train)

Y_pred = model.predict(X_test)

find_scores(Y_test, Y_pred, X_train)

extra_tree_configs = {
    "name": 'XGBRegressor',
    "method": "grid",
    "metric": {
        "name": "adj_r2",
        "goal": "maximize"
    },
    "parameters": {
        "learning_rate": {
            "values": [0.01, 0.03, 0.05, 0.07]
        },
        "n_estimators": {
            "values": [100, 200, 300]
        },
        "booster": {
            "values": ['gbtree', 'gblinear']
        },
        "eta": {
            "values": [0.01, 0.03, 0.05, 0.07]
        }
    }
}

sweep_id = wandb.sweep(sweep=extra_tree_configs, project="car_resale_value")
wandb.agent(sweep_id=sweep_id, function=XGB_regressor)

def LGBM_regressor():
    config_defaults = {
        'objective': 'root_mean_squared_error',
        'reg_sqrt': True,
        'metric': 'rmse',
        'random_state': 42
    }

```



```

wandb.init(config=config_defaults)
config = wandb.config

X = labeled.iloc[:,1:].values
Y = labeled.iloc[:,0].values.reshape(-1,1)

X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.4, random_state=42)

model = LGBMRegressor(
    learning_rate=config.learning_rate,
    n_estimators = config.n_estimators,
    random_state = config.random_state)

model.fit(X_train, Y_train)

Y_pred = model.predict(X_test)

find_scores(Y_test, Y_pred, X_train)

lgbm_configs = {
    "name": 'LGBMRegressor',
    "method": "grid",
    "metric": {
        "name": "adj_r2",
        "goal": "maximize"
    },
    "parameters": {
        "learning_rate": {
            "values": [0.01, 0.03, 0.05, 0.07]
        },
        "objective": {
            "values": ['root_mean_squared_error']
        },
        "boosting_type": {
            "values": ['gbdt','dart','goss','rf']
        },
        "reg_sqrt": {
            "values": [True]
        },
        "metric": {
            "values": ['rmse']
        },
        "n_estimators": {
            "values": [100,200,300]
        },
    },

```

```

        "random_state": {
            "values": [42]
        }
    }
}

sweep_id = wandb.sweep(sweep=lgbm_configs, project="car_resale_value")
wandb.agent(sweep_id=sweep_id, function=LGBM_regressor)

car_resale_value_prediction_LGBM.py
import pandas as pd
import numpy as np
from sklearn.preprocessing import LabelEncoder
from sklearn.model_selection import train_test_split
from sklearn.metrics import mean_absolute_error, mean_squared_error, r2_score
import pickle

#regression model
from lightgbm import LGBMRegressor

import os, types
import pandas as pd
from botocore.client import Config
import ibm_boto3

def __iter__(self): return 0

# @hidden_cell
# The following code accesses a file in your IBM Cloud Object Storage. It includes your
# credentials.
# You might want to remove those credentials before you share the notebook.
cos_client = ibm_boto3.client(service_name='s3',
    ibm_api_key_id='8DImq73hywb09uzAo_TsAZI_ocZgFLuhQdwmfUJZTX',
    ibm_auth_endpoint="https://iam.cloud.ibm.com/oidc/token",
    config=Config(signature_version='oauth'),
    endpoint_url='https://s3.private.us.cloud-object-storage.appdomain.cloud')

bucket = 'carresalevalueprediction-donotdelete-pr-whcxr42j79mqcv'
object_key = 'autos_preprocessed.csv'

body = cos_client.get_object(Bucket=bucket,Key=object_key)['Body']
# add missing __iter__ method, so pandas accepts body as file-like object
if not hasattr(body, "__iter__"): body.__iter__ = types.MethodType( __iter__, body )

```

```

data = pd.read_csv(body)
data.head()

labels = ['gearbox', 'notRepairedDamage', 'model', 'brand', 'fuelType', 'vehicleType']

mapper = {}
for i in labels:
    mapper[i] = LabelEncoder()
    mapper[i].fit(data[i])
    tr = mapper[i].transform(data[i])
    np.save(str('classes'+i+'.npy'), mapper[i].classes_)
    data.loc[:, i+'_labels'] = pd.Series(tr, index=data.index)

labeled = data[['price', 'yearOfRegistration', 'powerPS', 'kilometer', 'monthOfRegistration']
               +[x+"_labels" for x in labels]]

print(labeled.columns)

def find_scores(Y_actual, Y_pred, X_train):
    scores = dict()
    mae = mean_absolute_error(Y_actual, Y_pred)
    mse = mean_squared_error(Y_actual, Y_pred)
    rmse = np.sqrt(mse)
    rmsle = np.log(rmse)
    r2 = r2_score(Y_actual, Y_pred)
    n, k = X_train.shape
    adj_r2_score = 1 - ((1-r2)*(n-1)/(n-k-1))

    scores['mae']=mae
    scores['mse']=mse
    scores['rmse']=rmse
    scores['rmsle']=rmsle
    scores['r2']=r2
    scores['adj_r2_score']=adj_r2_score

    return scores

X = labeled.iloc[:,1:].values
Y = labeled.iloc[:,0].values.reshape(-1,1)

X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.4, random_state=42)

model =
LGBMRegressor(boosting_type="gbdt", learning_rate=0.07, metric="rmse", n_estimators=300, objective="root_mean_squared_error", random_state=42, reg_sqrt=True)

```

```

model.fit(X_train, Y_train)

Y_pred = model.predict(X_test)

find_scores(Y_test, Y_pred, X_train)

pickle.dump(model, open('resale_model.sav', 'wb'))

get_ipython().system('pip install -U ibm-watson-machine-learning')

from ibm_watson_machine_learning import APIClient
import json

wml_credentials = {
    "apikey": "Qo9j8ni7qMJ8j1C8VFDRFHbuGRAhYWcTlkVqnYg1AGkE",
    "url": "https://us-south.ml.cloud.ibm.com"
}

wml_client = APIClient(wml_credentials)
wml_client.spaces.list()

SPACE_ID= "bf7bc386-40bf-4d85-91e6-eedd2c53f245"

wml_client.set.default_space(SPACE_ID)

wml_client.software_specifications.list(100)

import sklearn
sklearn.__version__

MODEL_NAME = 'CRVP'
DEPLOYMENT_NAME = 'CRVP'
DEMO_MODEL = model

software_spec_uid = wml_client.software_specifications.get_id_by_name('runtime-22.1-py3.9')
model_props = {
    wml_client.repository.ModelMetaNames.NAME: MODEL_NAME,
    wml_client.repository.ModelMetaNames.TYPE: 'scikit-learn_1.0',
    wml_client.repository.ModelMetaNames.SOFTWARE_SPEC_UID: software_spec_uid
}

model_details = wml_client.repository.store_model(
    model=DEMO_MODEL,
    meta_props=model_props,

```

```
    training_data=X_train,
    training_target=Y_train
)
model_details
model_id = wml_client.repository.get_model_id(model_details)
model_id
deployment_props = {
    wml_client.deployments.ConfigurationMetaNames.NAME:DEPLOYMENT_NAME,
    wml_client.deployments.ConfigurationMetaNames.ONLINE: {}
}
deployment = wml_client.deployments.create(
    artifact_uid=model_id,
    meta_props=deployment_props
)
```

GitHub & Project Demo Link:

1. [Click here to redirect Git hub repository](#)
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