

DOAA CA1

# CARVALUIFY

**YOUR ONE-STOP CAR PRICE PREDICTOR WEBSITE**

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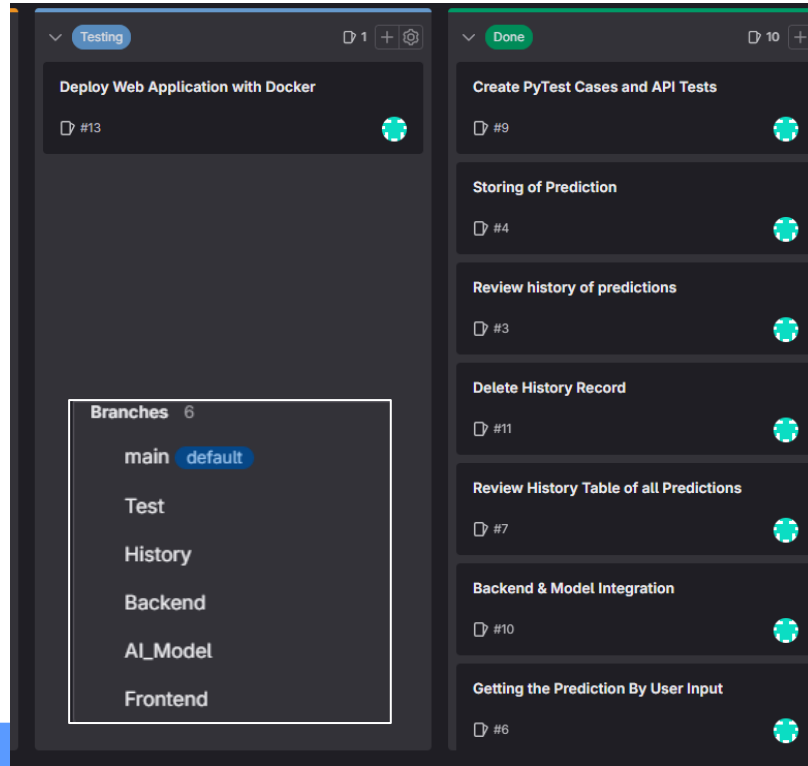


# PROJECT OBJECTIVE

Using **regression modelling** for car price prediction, **predict car prices**, given a set of input car feature data.

For our **'100,000 UK Cars'** dataset, it contained extensive data on car model, car registration year, mileage, tax and more.

# DEVOPS PROCESS



## SETTING UP GITLAB AND SCRUM BOARD

- Managed a **scrum board** to enable us to keep track of our web application development.
- The scrum board contained 4 key labels, 'To-Do', 'In Progress', 'Testing', and 'Done'.
- Created **6** branches for frontend, backend, model building and testing for us to make changes to the folder in a controlled environment.

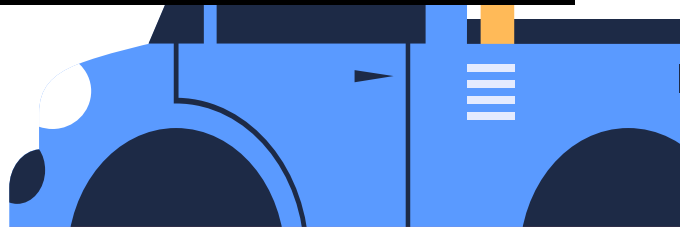
# PRESENTING: CARVALUIFY

## CARVALUIFY

CarValuify is a **simple-to-use web application** that enables you to predict your car prices & valuations through inputting your car details. Simply enter the necessary fields, and our AI algorithm will work out the rest!

**PREDICT YOUR CAR PRICE BELOW**

Enter Car Registration Year: e.g., 2015	Enter Mileage: e.g., 10000
Enter Transmission Type: Manual	Enter MPG: e.g., 30
Enter Age of Car: e.g., 5	Enter Tax: e.g., 200
Enter Fuel Type: Petrol	Enter Engine Size: e.g., 2.5



# GENERAL INFORMATION



## **Authentication required for non-public pages.**

Bypassing the authentication by typing in direct link **is addressed and is prevented.**

**Username : student**


**Password : student**



## **Responsive + Flexible Website Design using Bootstrap 4 and TailwindCSS.**

Webpage is viewable on multiple devices and is customizable to the screen size.

# CARVALUIFY | CAR PRICE PREDICTOR



**WELCOME BACK TO CARVALUIFY!**

Username

This field is required.

Password

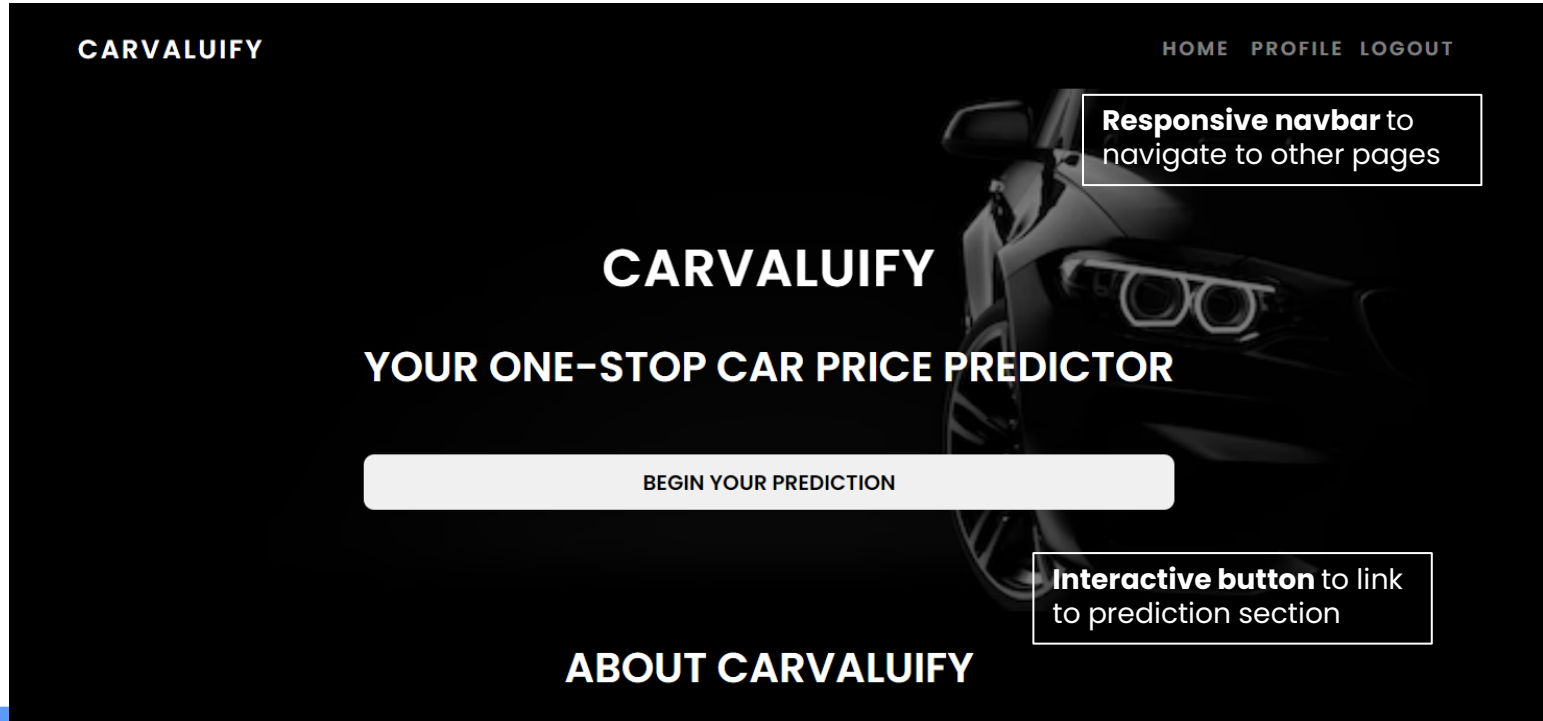
This field is required.

LOGIN

**Login Page :** Prompt users to enter the username and password before proceeding

Errors & validation checks are displayed using Flask's **flash()** function.

# CARVALUIFY | CAR PRICE PREDICTOR



CARVALUIFY

HOME PROFILE LOGOUT

**Responsive navbar** to  
navigate to other pages

CARVALUIFY  
YOUR ONE-STOP CAR PRICE PREDICTOR









BEGIN YOUR PREDICTION

**Interactive button** to link  
to prediction section

ABOUT CARVALUIFY

# CARVALUIFY | CAR PRICE PREDICTOR

## PREDICT YOUR CAR PRICE BELOW

 Enter Car Registration Year: e.g., 2015	 Enter Mileage: e.g., 10000
 Enter Transmission Type: Manual	 Enter MPG: e.g., 30
 Enter Age of Car: e.g., 5	 Enter Tax: e.g., 200
 Enter Fuel Type: Petrol	 Enter Engine Size: e.g., 2.5

PREDICT CAR VALUATION

Added  
**attractive  
icons** for user  
engagement

**Dropdown box**  
options for  
easy selection  
for our user



# CARVALUIFY | CAR PRICE PREDICTOR

**Attractive graphic** for viewing of the predicted price



**YOUR CAR'S PRICE IS : \$ 11389.1**

RETURN HOME

**Simple 'Return Home' button** to easily return to the home page

# CARVALUIFY | CAR PRICE PREDICTOR

**Profile Page** for users  
to view prediction  
histories

**WELCOME BACK, STUDENT.**

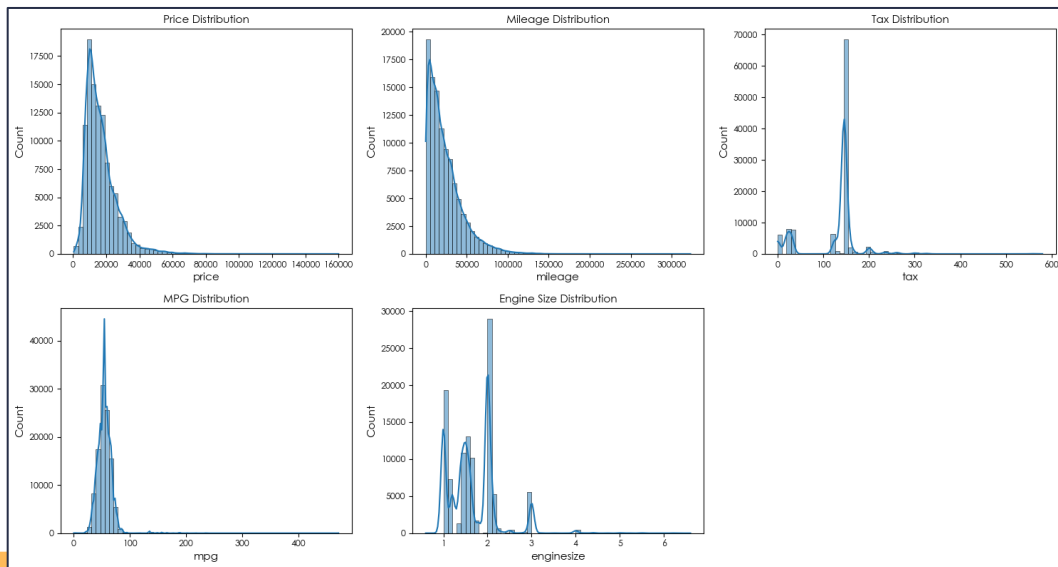
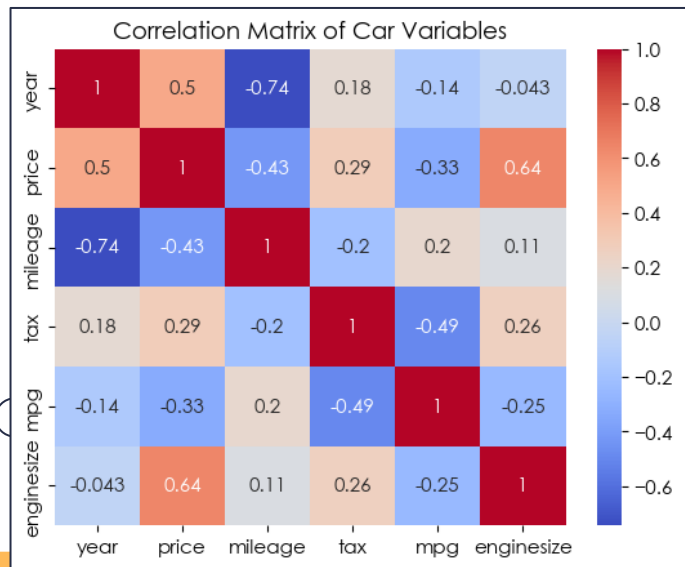
Check out your latest car valuation histories below.

Displays all the  
relevant columns and  
an 'X' for delete.

Year of Registration	User ID	Mileage	MPG	Car Age	Tax	Fuel Type	Transmission Type	Engine Size	Price (\$)	Date	Delete
2013	1	1000	20.0	50	20.0	petrol	manual	1.4	11389.1	04 Dec 23 08:00	✗
2014	1	1000	30.0	3	200.0	petrol	manual	1.2	11389.1	04 Dec 23 07:27	✗
2014	1	1000	20.0	6	200.0	petrol	manual	2.1	10960.9	04 Dec 23 06:47	✗
2017	1	5000	35.0	3	150.0	electric	automatic	2.0	20000.0	04 Dec 23 02:18	✗
2015	1	10000	30.0	5	200.0	petrol	manual	2.5	15000.0	04 Dec 23 02:18	✗

# MACHINE LEARNING REGRESSOR

- First, conducted extensive EDA on the dataset to discover insights and patterns in data.
- Found **moderate to high correlation** between most of the car variables in dataset.
- Most variables like Price and Mileage are **slightly skewed** (imbalanced distribution).



# MACHINE LEARNING REGRESSOR

- Dropped unnecessary columns not useful for the analysis.

```
# Dropping unnecessary columns from the dataset
car_df.drop(['tax(£)'], axis=1, inplace=True)
car_df
```

- Imputed missing values with **MEDIAN** to remove NULLS and prevent loss of data rows.

```
# Columns with missing values: 'tax' and 'mpg'
# Since 'tax' and 'mpg' are numerical columns, we'll fill missing values with their median
for col in ['tax', 'mpg']:
    median = car_df[col].median()
    car_df[col].fillna(median, inplace=True)
```

- Dropped **DUPLICATE** rows of data to prevent biasness to certain groups of values.

```
# Count number of duplicate rows of data and print them
duplicate = car_df[car_df.duplicated(keep=False)]
print("Number of duplicate rows before removal:", duplicate.count().sum())
```

# MACHINE LEARNING REGRESSOR

## Feature Engineering Performed

- Robust Scaling of Data
- Added new column 'Car Age' (Now – Registration Year)
- One-hot Encoded Labels
- Train-Test Split X and Y

## Cross-Validation Performance

- Best-performing Model : RANDOM FOREST REGRESSOR
- MAPE : 0.1156
- RMSE : 2962.17

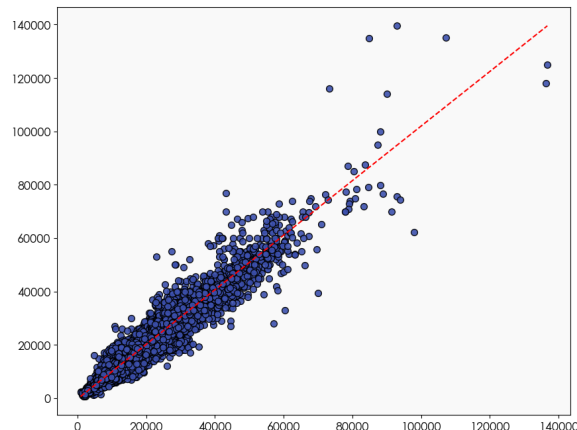
## Hyperparameter Tuning – Randomized Search

- **R2 : 0.92**
- **MAPE : 10.93%**

### Regression Model Evaluation Metrics:

```
-----  
Mean Squared Error (MSE): 7247929.99  
Root Mean Squared Error (RMSE): 2692.20  
Mean Absolute Percentage Error (MAPE): 10.93%  
Mean Absolute Error (MAE): 1713.48  
R-squared Coefficient: 0.92  
Explained Variance Score: 0.92
```

Predicted vs. Actual Scatterplot



# WEB APIs DEVELOPED

## POST API/ADD

API used to add a new prediction record to the database table.

Login is needed.

## GET API/GET

API used to get an existing prediction record from the database table.

Login is needed.

## DELETE API/DELETE

API used to delete an existing prediction record from the database table.

Login is needed.

## POST API/PREDICT

API used to perform predictions and gain price insights on the car.

Login is needed.

# TESTING WITH PYTEST

**54 TESTS**

## ○ VALIDITY TESTING

Tested on valid data to determine if ordinary data can be used. Ensured data entered was acceptable.

## RANGE TESTING

Extreme values for columns such as Year, MPG, Car Age and Mileage etc. Tested both lower and upper bounds.

## CONSISTENCY TESTING ○

Given identical inputs, checked if prediction and output prices are the same.

## EXPECTED FAILURE TESTING

○ Tested XFAIL on:

- Invalid Year of Registration
- Invalid Mileage ( < 0)
- Invalid MPG
- Invalid Car Age
- Invalid tax value

## UNEXPECTED FAILURE TESTING ○

Tested invalid inputs on:

- Year of Registration (Non-integer)
- MPG (Less than 0)
- Mileage (Negative value)

# TESTING WITH PYTEST

**52 TESTS DONE**

```
-- Docs: https://docs.pytest.org/en/stable/how-to/capture-warnings.html
===== short test summary info =====
FAILED tests/test_application.py::test_CarEntryClass[car_data4] - ValueError: Year must be an integer.
FAILED tests/test_application.py::test_CarEntryClass[car_data5] - ValueError: mpg must be greater than 0
FAILED tests/test_application.py::test_CarEntryClass[car_data7] - ValueError: mileage must be a positive number.
FAILED tests/test_application.py::test_CarEntryClass[car_data21] - ValueError: Year must be an integer.
FAILED tests/test_application.py::test_getAPI[predictionList1] - assert 2014 == 2019
===== 5 failed, 40 passed, 5 xfailed, 2 xpassed, 4 warnings in 5.22s =====
```





**THANK YOU!**