

MACHINE LEARNING-BASED VEHICLE PERFORMANCE ANALYZER

NALAIYA THIRAN PROJECT BASED LEARNING

On

PROFESSIONAL READINESS FOR INNOVATION, EMPLOYABILITY, AND ENTREPRENEURSHIP

A PROJECT REPORT

TEAM ID: PNT2022TMID46974

Team members:

ISWARYA V - 822119104016

ABINAYA M - 822119104003

DIVYA M - 822119104012

ABIRAMI M - 822119104004

BACHELOR OF ENGINEERING

IN

COMPUTER SCIENCE ENGINEERING

UNIVERSITY COLLEGE OF ENGINEERING - PATTUKKOTTAI

(A Constituent College of Anna University, Chennai)

PATTUKKOTTAI-614701

NOVEMBER 2022

TABLE OF CONTENTS

1 INTRODUCTION

2	LITERATURE SURVEY
3	IDEATION & PROPOSED SOLUTION
4	REQUIREMENT ANALYSIS
5	PROJECT DESIGN
6	PROJECT PLANNING & SCHEDULING
7	CODING & SOLUTIONING
8	TESTING
9	RESULT
10	ADVANTAGES & DISADVANTAGES
11	CONCLUSION

12	FUTURE SCOPE
12	APPENDIX

1.INTRODUCTION

1.1 Project Overview

Forest fires are a major environmental issue, creating economic and ecological damage while endangering human lives. There are typically about 100,000 wildfires in the United States every year. Over 9 million acres of land have been destroyed due to treacherous wildfires. It is difficult to predict and detect Forest Fire in a sparsely populated forest area and it is more difficult if the prediction is done using ground-based methods like Camera or Video-Based approach. Satellites can be an important source of data prior to and also during the Fire due to its reliability and efficiency. The various real-time forest fire detection and prediction approaches, with the goal of informing the local fire authorities.

Automotive Technologies are providing improvised services to the driver's safety and vehicle security under the umbrella of Intelligent Transportation System (ITS). In the development of ITS, advanced Automotive Technologies shall play a crucial role in determining the overall experience of users by making it much at ease in terms of reducing the risk of road accidents, risk of cybercrime in the vehicle, buying a used car etc. It is often noted that judging the driver's driving skill is subjective and is difficult to set a standard for driver's skills. The modern approach to transportation system is focusing on rapidly evolving with the intelligent vehicles. High rise in recorded traffic density, road accidents and crisis faced in regulating the effective management of traffic control in urban and rural 4 areas have concerned us to develop a smart solution in context to ITS.

The automotive industry has great expectations from these futuristic solutions to improve the safety of people and security of vehicles. It is observed that the users are shifting from individualistic approach to the data-centric approach based on OBD-II scanner to avail the augmented driving experience. In spite of the modern command, control, communication, computers and intelligent systems, we are still facing numerous calamities in which thousands of precious humans lives are lost in accidents. Therefore, it should be an immediate need to tackle the small scale yet serious issues using the state-of-the-art techniques. We are mainly focusing on analyzing the data which is collected from the vehicle using the OBD-II scanner and eventually providing the driver's safety solutions. We aim to obtain the solutions by

observing the blin techniques from sup	d-spots accurately pervised learning.	and	efficiently	using	pattern	recognition

2. LITERATURE SURVEY

- ♣ Singh D, Singh M., "Internet of Vehicles for Smart and Safe Driving", International Conference on Connected Vehicles and Expo (ICCVE), Shenzhen, 19 -23 Oct.,2015. (This paper has discussed about smart transportation services in cloud (Cloud-STS) for safety and convenience. STS provide driver centric board services in the cloud networks.
- ♣ STS composed of Vehicle to Wi Fi networks (V to Wi Fi), Vehicle to Cloud Network (V to CN), Vehicle to Vehicle (V to V), and Cloud Network to service provider (CN to SP). The idea is to utilize the (Wi Fi enabled) Smart Highways and 3Dcameraenabled dash board 5 navigation device to enhance accident prevention /monitoring and control.)
- ♣ Zhang, Y., Lin, W., and Chin, Y., "Data -Driven Driving Skill Characterization: Algorithm Comparison and Decision Fusion," SAE Technical Paper2009 -01 -1286, 2009, https://doi.org/10.4271/2009 -01 1286.Azevedo, C. L Cardoso. (By adapting vehicle control systems to the skill level of the driver, the overall vehicle active safety provided to the driver can be further enhanced for the existing active vehicle controls, such as ABS, Traction Control, Vehicle Stability Enhancement Systems.
- As a follow-up to the feasibility study in, this paper provides some recent results on data-driven driving skill characterization. In particular, the paper presents an enhancement of discriminant features, the comparison of three different learning algorithms for recognizer design, and the performance enhancement with decision fusion. The paper concludes with the discussions of the experimental results and some of the future work.)
- → J. E. Meseguer, C. T. Calafate, J. C. Cano and P. Manzoni, "Driving Styles: A smartphone application to assess driver behavior," 2013 IEEE Symposium on Computers and Communications (ISCC), Split,2013, pp.000535 -000540. oi:10.1109/ISCC.2013.6755001. (The Driving Styles architecture integrates both data mining techniques and neural networks to generate a classification of driving styles by analyzing the driver behavior along each route.

- ♣ In particular, based on parameters such as speed, acceleration, and revolutions per minute of the engine (rpm), we have implemented a neural network-based algorithm that is able to characterize the type of road on which the vehicle is moving, as well as the degree of aggressiveness of each 6 driver.
- ♣ The final goal is to assist drivers at correcting the bad habits in their driving behavior, while offering helpful tips to improve fuel economy. In this work we take advantage of two key-points: the evolution of mobile terminals and the availability of a standard interface to access car data.)
- ♣ Kenneth L. Clarkson. 1985. Algorithms for Closest -Point Problems (Computational Geometry). Ph.D. Dissertation. Stanford University, Palo Alto, CA. UMI Order Number: AAT 8506171. (This dissertation reports a variety of new algorithms for solving closest point problems. The input to these algorithms is a set or sets of points in d-dimensional space, with an associated L (, p) metric.
- → The problems considered are:(1) The all nearest neighbors problem. For point set A, find the nearest neighbors in A of each point in A.(2) The nearest foreign neighbor problem. For point sets A and B, find the closest point in B to each point in A. The geometric minimum spanning tree problem.) Goszczynska H., Kowalczyk L., Kuraszkiewicz B. (2014) Correlation Matrices as a ool to Analyze the Variability of EEG Maps.
- ♣ In: Piętka E., Kawa J., Wieclawek W. (eds) Information Technologies in Biomedicine, Volume 4. Advances in Intelligent Systems and Computing, vol 284. Springer. (The aim of this paper is to present the selected examples of possible applications of image of correlation coefficients matrix of EEG map series in the analysis of variation of the topography of the iso potential areas in EEG maps, and thus in the assessment of stationarity, spatiotemporal variability and trends of changes of bioelectric activity of the brain.

2.2 References

- ➤ Singh D, Singh M., "Internet of Vehicles for Smart and Safe Driving", International Conference on Connected Vehicles and Expo (ICCVE), Shenzhen, 19-23 Oct., 2015.
- ➤ Zhang, Y., Lin, W., and Chin, Y., "Data-Driven Driving Skill Characterization: Algorithm Comparison and Decision Fusion," SAE Technical Paper 2009 -01-1286, 2009, https://doi.org/10.4271/2009-01-1286.Azevedo, C. L Cardoso.
- ➤ J.E.Meseguer, C.T.Calafate, J.C.Canoand P.Manzoni, "Driving Styles: A smartphone application to assess driver behavior, "2013 IEEE Symposium on Computers and Communications (ISCC), Split, 2013, pp.000535- 000540.doi: 10.1109/ISCC.2013.6755001.
- ➤ Schneider, A,Hommel,G.,& Blettner,M.(2010). Linear Regression Analysis: Part 14 of a Series on Evaluation of Scientific Publications. Deutsches Ärzteblatt International, 107(44), pp. 776–782.

2.3 Problem Statement Definition



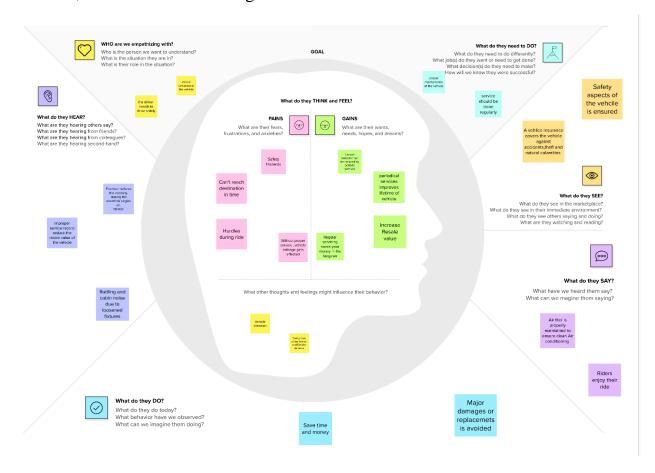
mire

Problem Statement (PS)	I am (Customer)	I'm trying to	But	Because	Which makes me feel
PS-1	Owner of a vehicle.	Collect the properties of the vehicle to predict the performance.	I don't know which properties to select.	I don't know which are the properties to be considered.	Confused.
PS-2	Owner of a vehicle.	Make predictions.	I don't know how the variables are correlated to each other.	There are many properties and what should I consider.	Little worried.

3. IDEATION & PROPOSED SOLUTION

3.1 Empathy Map Canvas

An empathy map is a collaborative tool teams can use to gain a deeper insight into their customers. Much like a user persona, an empathy map can represent a group of users, such as a customer segment.



3.2 Ideation & Brainstorming

Brainstorming provides a free and open environment that encourages everyone within a team to participate in the creative thinking process that leads to problem-solving.

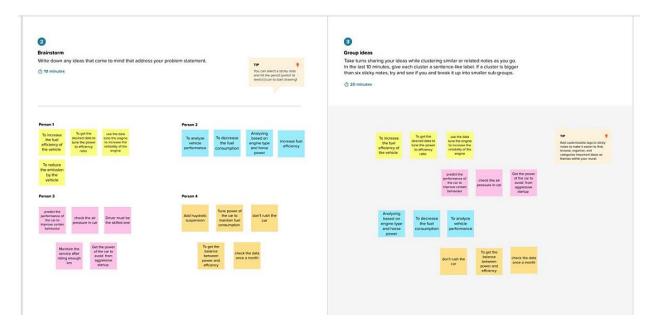
Prioritizing volume over value, out-of-the-box ideas are welcome and built upon, and all participants are encouraged to collaborate, helping each other develop a rich amount of creative solutions.

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

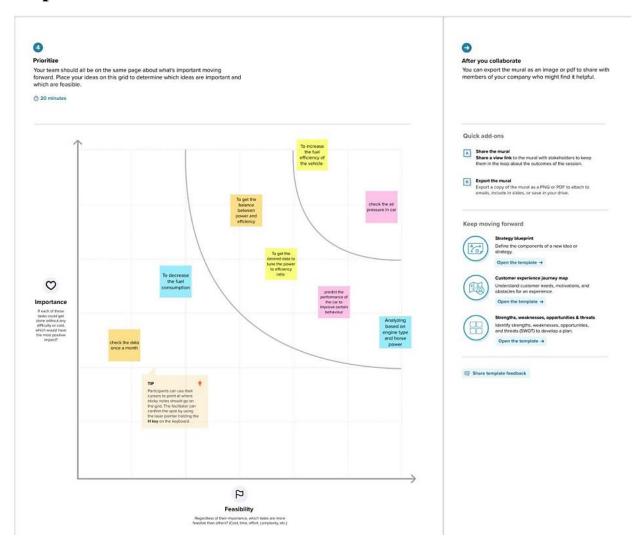
Step-1: Team Gathering, Collaboration and Select the Problem Statement:



Step-2: Brainstorm, Idea Listing and Grouping:



Step-3: Idea Prioritization:



3.3 Proposed Solution:

S.No.	Parameter	Description
1.	Problem Statement (Problem to be solved)	The objective of this project is to predict the price of used cars using the various Machine Learning models by using User Interface (UI).
2.	Idea / Solution description	To train the system with the dataset using a regression model and it will be integrated to the web-based application where the user is notified with the status.
3.	Novelty / Uniqueness	By using the optimal regression model to predict the value in a less amount to time and predict its value.
4.	Social Impact / Customer Satisfaction	The customer can get an idea about the resale value of their vehicle to predict the performance. By knowing the vehicle brand, fuel type, kilometeres driven.
5.	Business Model (Revenue Model)	The web-based application has a friendly UI for the customer to enter their vehicles detail and the system predicts the value within few seconds.
6.	Scalability of the Solution	Machine learning approaches, this project proposed a scalable framework for predicting values for different type of used cars. The solution given by the trained system is efficient and is nearly accurate value of the vehicle.

3.4 Problem Solution Fit:

CS

J&P

TR

ΕM

1. CUSTOMER SEGMENT(S)

The customer is one who wants to predict the performance of the vehicle.

6. CUSTOMER CONSTRAINTS

What constraints prevent your customers from taking action or limit their choices of solutions? i.e. spending power, budget, no cash, network connection, available devices.

- To determine the worthiness of the car by their own within few minutes
- A loss function is to be optimized by spending money for dealers, brokers to buy or sell a car.

5. AVAILABLE SOLUTIONS

CC

RC

problem or need to get the job done? What have they tried in the past? What pros & cons do these solutions have? i.e. pen and paper is an alternative to digita

- In the past User cannot find the value of used car buy their own without prior knowledge about cars.
- •A person who don't know much about the car can also make predictions for used cars easily.

2. JOBS-TO-BE-DONE / PROBLEMS

To build a supervised machine learning model using regression algorithms for forecasting the value of a vehicle based on multiple attributes such as Condition of Engine, Year of Registration, Kilometers, Number of Owner

9. PROBLEM ROOT CAUSE

- The price predicted by the dealers or brokers for used car is not trustful
- · Users can predict the correct valuation of the car remotely without human intervention like car dealers.
- · User can eliminate the valuation predicted by the dealer.

7. BEHAVIOUR

The History of Your Car's condition and documents produced by them will be Suspicious. The model is to be built would give the nearest value of the vehicle by eliminating anonymous value predicted by using humans.

What triggers customers to act? i.e. seeing their neighbour installing solar panels, reading about a more efficient solution in the news.

Users can predict the correct valuation of the car by their own like Olxcars, Cars24 and other car resale value prediction websites by using model,year,owner,etc

If you are working on an existing business, write down your current solution first, fill in the canvas, and check how much it tilts reality. If you are working on a new business proposition, then keep it blank until you fill in the canvas and come up with a solution that tilts within customer limitations, solves a problem and matches customer behaviour.

• The main aim of this project is to predict the price of used cars using the Machine Learning(ML) algorithms and collection data's about different cars.

8. CHANNELS of BEHAVIOUR

8.1 ONLINE What kind of actions do customers take online? Extract online channels from #7

8.2 OFFLINE
What kind of actions do customers take offline? Extract offline channels from #7 and use them for customer development.

· Customer should predict the worth of the car by using different parameters

- given by the owner. · User Should confirm the details provided about the vehicle in
- RTOonline.

4. EMOTIONS: BEFORE / AFTER

astomers feel when they face a problem or a job and afterwards? secure > confident, in control - use it in your communication strategy & design

 User will be in fear about the biased values predicted by the humans based on the condition of the car.

After:

 User can determine the worthiness of the car by their own without human intervention.

• The project should take parameters related to used car as inputs and enable the customers to make decisions by their own.

- · User can decide by seeing the exterior and interior condition of the car.
- · User can test the performance of the car and to buy it up in a affordable price based on its condition.

4.REQUIREMENT ANALYSIS

4.1 Functional requirement:

Following are the functional requirements of the proposed solution:

FR No.	Functional Requirement (Epic)
ED 1	Extends Issued
FR-1	Enter the Inputs
FR-2	User Essential
FR-3	Data Prepossessing
FR-4	User input Evaluation
FR-5	Prediction

4.2 Non-Functional requirements:

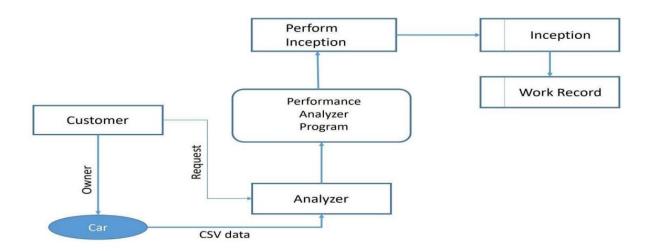
Following are the non-functional requirements of the proposed solution.

FR No.	Non-Functional Requirement	Description
NFR-1	Get Inputs through a form	The analyzer allows the user to improve performance based on the results provided. It is easy to use with just the data required.
NFR-2	Predict the performance of the vehicle	The security is improved by using vehicle alarm, wheel lock, vehicle lock and also GPS tracker
NFR-3	Sample Dataset for training purpose	The reliability rating is good due to best performance, less frequency of problem occurrence and cost for repairing is low
NFR-4	Evaluating the given user values	The vehicle is upgraded in their quality and infrastructure to provide better performance like good mileage, smooth travel.
NFR-5	Fuel consumption and efficiency of the vehicle	The data required is collected by research persons and this data can be used to provide better results.

5. PROJECT DESIGN

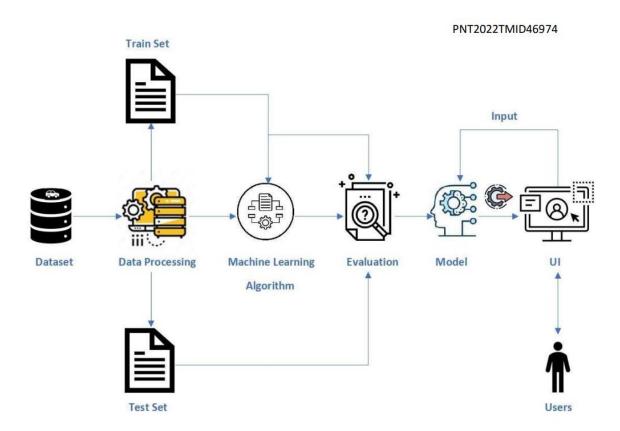
5.1 Data Flow Diagrams

A Data Flow Diagram (DFD) is a traditional visual representation of the information flows within a system. A neat and clear DFD can depict the right amount of the system requirement graphically. It shows how data enters and leaves the system, what changes the information, and where data is stored.



5.2 Solution & Technical Architecture

A solution architecture (SA) is an architectural description of a specific solution. Solution Architectures combine guidance from different enterprise architecture viewpoints (business, information and technical), as well as from the enterprise solution architecture (ESA).



5.3 User Stories

User Type	Functional Requirement (Epic)	User Story Number	User Story / Task	Acceptance criteria	Priority	Release
Customer	Access the webpage	USN-1	As a user, anyone can access the webpage to check the specifications of the vehicle.	can access my webpage online at any time.	High	Sprint-1
Customer	Performance of the vehicle	USN-2	As per the usage of the user, the performance of the vehicle should be predicted easily.	Prediction can be done in an easy way.	High	Sprint-2
Customer	Accuracy to check the performance and health of the car	USN-3	By using our prediction, it helps to check the health of the car.	efficiency of	High	Sprint-1

6.PROJECT PLANNING & SCHEDULING

6.1 Sprint Planning & Estimation

Product Backlog, Sprint Schedule, and Estimation

Use the below template to create product backlog and sprint schedule

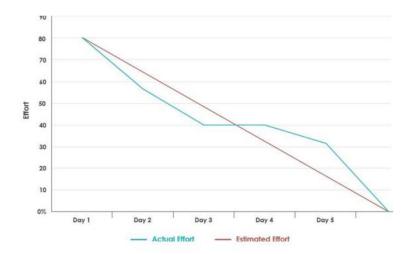
Sprint	Functional Requirement (Epic)	User Story Number	User Story / Task	Story Points	Priority	Team Members
Sprint-1	Data Preparation	USN-1	Collecting Car dataset and pre- processing it	10		
Sprint-1	Data Modeling	USN-2	create an ml model to predict the car Performance	5		
Sprint-1	Model Evaluation	USN-3	Calculate the performance, error rate, and complexity of ML model	5		
Sprint-2	Model Deployment	USN-4	Using flask and deploying model finally in IBM cloud using IBM storage and Watson Studio	20		
Sprint-3	Model prediction	USN-5	Predict the type of model. Here we have used random forest regression	10		
Sprint-3	Local deployment	USN-5	Local deployment will be done	5		

			while running the python fie in spyder		
Sprint-3	Application building	USN-6	As a user, I can give the necessary inputs	5	
Sprint-4	dashboard	USN-7	As a user, I can use the application by entering Car data	20	

Velocity:

Average Velocity = 80 / 20 = 4 Story Points per Day

BurnDown Chart:



7.CODING & SOLUTIONING

7.1 Feature 1

```
In [4]: import numpy as np
         import pandas as pd
In [5]: #Reading The Dataset
         datas = pd.read_csv(r"C:\Users\sunda\Desktop\IBM\car performance.csv")
         datas.head()
Out[5]:
            mpg cylinders displacement horsepower weight acceleration model year origin
                             307.0
                                            130 3504
          0 18.0
                      8
                                                           12.0
                                                                          70 1 chevrolet chevelle malibu
          1 15.0
                        8
                                 350.0
                                             165 3693
                                                               11.5
                                                                           70
                                                            11.0
                                           150 3438
          2 18.0 8
                                318.0
                                                                          70
                                                                                          plymouth satellite
                        8
                                 304.0
                                             150 3433
                                                               12.0
                                                                          70
          3 16.0
                                                                                             amc rebel sst
          4 17.0 8 302.0 140 3449
                                                                          70 1
                                                                                               ford torino
In [6]: # HandLing Missing Values
         datas.isnull().any()
Out[6]: mpg
         cylinders
         displacement
                          False
         horsepower
                          False
         weight
                          False
         acceleration
                          False
         model year
                          False
         origin
                          False
         car name
                          False
         dtype: bool
In [7]: # Splitting The Dataset Into Dependent And Independent Variable.
         x = datas.iloc[:,1:8].values
In [8]: y = datas.iloc[:,0].values
In [9]: # Split The Dataset Into Train Set And Test Set
         from \ sklearn.model\_selection \ import \ train\_test\_split \\ x\_train\_x\_test\_y\_train\_y\_test = train\_test\_split(x\_y\_, test\_size=0.2\_, random\_state=0)
In [10]: # Normalizing
         from sklearn.preprocessing import StandardScaler
         sd = StandardScaler()
x_train = sd.fit_transform(x_train)
         x_{test} = sd.fit_{transform(x_{test})}
```

```
In [11]: x_train
Out[11]: array([[ 1.49526939, 1.22961301, 1.24359144, ..., -0.79520768,
                          -1.13752513, -0.73301171],
                        [-0.85285735, -0.92367663, -1.16092059, ..., 1.24411524, -1.41177304, 0.5068698 ],
                        [-0.85285735, -0.92367663, -0.68001818, ..., 0.05760009,
                           1.05645814, 0.5068698 ],
                       [-0.85285735, -1.206235 , -1.45480539, ..., 1.42950823, -0.86327722, 0.5068698 ], [0.32120602, 0.56706235, -0.09224857, ..., -0.2390287 ,
                        -1.41177304, -0.73301171],
[-0.85285735, -0.99188037, -0.86703579, ..., 0.31715028,
                          -0.31478141, 0.5068698 ]])
In [12]: # Build The Model With The Random Forest Regressor
              from sklearn.ensemble import RandomForestRegressor
              d = RandomForestRegressor (n_estimators=30, random_state = 0)
             d.fit(x_train,y_train)
Out[12]: RandomForestRegressor(n_estimators=30, random_state=0)
In [13]: # prediction
             y_pred = d.predict(x_test)
             y_pred
Out[13]: array([14.38333333, 24.25666667, 14.21666667, 20.56666667, 18.47333333,
                                                           3, 21.15 , 16.30333333, 25.76 , , 19.53666667, 27.32333333, 16.54333333,
                        30.21666667, 34.63333333, 21.15
                        36.60333333, 36.27
                         32.99333333, 28.32333333, 27.49666667, 17.03 , 35.82
                        16.47333333, 23.54
                                                          , 23.16666667, 20.7
                                                                                                  , 33.69666667,
                                      , 33.79666667, 30.37333333, 31.93666667, 16.57333333,
67, 32.99 , 19.79666667, 34.08333333, 20.85666667,
                         20.26666667, 32.99
                       20.26666667, 20.54333333, 17.14 , 34.78333333, 12.76666667, 13.73333333, 15.2 , 28.32 , 32.76666667, 28.74333333, 22.68666667, 20.54333333, 16.50666667, 23.38 , 29.88333333, 34.31666667, 26.5 , 17.63 , 27.78333333, 15.9666667, 12.9666667, 18.86666667, 26.91666667, 31.95666667, 15.68 , 20.81 , 25.97 , 19.84666667, 21.6 , 13.46666667, 15.33333333, 14.2 , 18.90333333, 24.72666667, 14.21666667,
                       , 15.84666667, 21.6 , 13.46666667, 21.6 , 13.46666667, 24.72666667, 14.21666667, 24.87666667, 13.25 , 22.96666667, 18.77666667, 23.83333333, 32.166666667, 28.176666667, 31.236666667, 31.94 , 14.35
In [14]: # Model Evaluation
              from sklearn.metrics import r2_score
              accuracy = r2_score(y_test,y_pred)
             accuracy
Out[14]: 0.8914224071232417
```

7.1 Feature 2

```
In [54]: import numpy as np
           import pandas as pd
In [55]: #Reading The Dataset
           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.
           "You might work to remove those creatives before you shall a cos_client = ibm_boto3.client(service_name='s3',
   ibm_api_key_id='BqUsBKYfxzrLPRBCAnlsYi1-20YTS38c8Ks157IQ1BS7',
   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 = 'vehicleperformanceanalyzer-donotdelete-pr-ccgjs2talld8s5'
           object_key = 'car performance.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 )
           datas = pd.read_csv(body)
           datas.head()
Out[55]: mpg cylinders displacement horsepower weight acceleration model year origin
            0 18.0
                      8 307.0 130 3504 12.0
                                                                                70 1 chevrolet chevelle malibu
                                      350.0
                                                     165 3693
                                                                                       70
                                                                       11.0
                          8
                                      318.0
            2 18.0
                                                  150 3438
                                                                                     70
            3 16.0
                            8
                                      304.0
                                                    150 3433
                                                                         12.0
                                                                                       70
            4 17.0 8 302.0 140 3449 10.5 70 1
In [56]: # Handling Missing Values
           datas.isnull().any()
Out[56]: mpg
           cylinders
                               False
           displacement
                               False
           horsepower
                               False
            weight
                               False
           acceleration
                               False
           model year
                               False
           origin
car name
dtype: bool
                               False
```

```
In [63]: # prediction
             v pred = d.predict(x test)
y_pred
                       26.81 , 25.97 , 19.84eebee/, 21.6
15.33333333 , 14.2 , 18.99333333 , 24.726
34.87666667 , 13.25 , 22.9666667 , 18.776
32.16666667 , 28.17666667 , 31.23666667 , 31.94
                                                                                          , 14.35
 In [64]: !pip install ibm_watson_machine_learning
              Requirement already satisfied: ibm_watson_machine_learning in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (1.0.257)
Requirement already satisfied: importlib-metadata in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm_watson_ma
              chine_learning) (4.8.2)
Requirement already satisfied: requests in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm_watson_machine_lear
              ning) (2.26.0)

Requirement already satisfied: pandas<1.5.0,>=0.24.2 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm_watson
               machine learning) (1.3.4)
              Requirement already satisfied: tabulate in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm_watson_machine_lear
              ning) (0.8.9)
              Requirement already satisfied: urllib3 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm watson machine learn
              Requirement already satisfied: ibm-cos-sdk==2.11.* in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm watson m
              achine_learning) (2.11.0)

Requirement already satisfied: certifi in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm_watson_machine_learn
              ing) (2022.9.24)
              Requirement already satisfied: packaging in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm_watson_machine_lea
              rning) (21.3)
Requirement a
                          ent already satisfied: lomond in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm_watson_machine_learni
              ng) (0.3.3)
             Requirement already satisfied: jmespath<1.0.0,>=0.7.1 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-cos-s dk==2.11.*->ibm_watson_machine_learning) (0.10.0)
Requirement already satisfied: ibm-cos-sdk-e2.11.0 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from ibm-cos-sdk=2.11.*->ibm_watson_machine_learning) (2.11.0)
              Requirement already satisfied: ibm-cos-sdk-s3transfer==2.11.0 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from i
             hequirement already satisfied: Limitor-sak-sak marker ==2.1.4 in /opt/conda/envs/python-3.9/lib/python3.9/site-packages (from 1 bm-cos-sak=2.11.*-)ibm_watson_machine_learning) (2.11.0)

Requirement already satisfied: python-dateutil<a.0.0,>=2.1 in /opt/conda/envs/python-3.9/lib/python3.9/site-packages (from 1bm-cos-sak-core=2.11.0.*-)ibm_watson_machine_learning) (2.8.2)

Requirement already satisfied: pytz>=2017.3 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from pandas<1.5.0,>=0.2
             4.2->ibm_watson_machine_learning) (2021.3)
Requirement already satisfied: numpy>=1.17.3 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from pandas<1.5.0,>=0.2
4.2->ibm_watson_machine_learning) (1.20.3)
             Requirement already satisfied: six>=1.5 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from python-dateutil<3.0.0,> =2.1->ibm-cos-sdk-core==2.11.0->ibm-cos-sdk=2.11.*->ibm_watson_machine_learning) (1.15.0)
Requirement already satisfied: idna<4,>=2.5 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from requests->ibm_watson_machine_learning)
              n_machine_learning) (3.3)
               Requirement already satisfied: zipp>=0.5 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from importlib-metadata->ib
               m watson machine learning) (3.6.0)
               Requirement already satisfied: pyparsing!=3.0.5,>=2.0.2 in /opt/conda/envs/Python-3.9/lib/python3.9/site-packages (from packagi
               ng->ibm_watson_machine_learning) (3.0.4)
 In [65]: from ibm watson machine learning import APIClient
               client = APIClient(wml_credentials)
 In [66]: def guid_from_space_name(client, space_name):
    space = client.spaces.get_details()
                    #print(space)
                    return(next(item for item in space['resources'] if item['entity']["name"] == space_name )['metadata']['id'])
  In [67]: space_uid = guid_from_space_name(client, 'models')
              print("Space UID = '
                                          + space uid)
               Space UTD = Regreg32-c021-4332-a82c-9994699d398a
  In [68]: client.set.default_space(space_uid)
 Out[68]: 'SUCCESS'
```

```
In [69]: client.software specifications.list()
             NAME
                                                    ASSET ID
                                                                                                    TYPE
             default_py3.6
                                                    0062b8c9-8b7d-44a0-a9b9-46c416adcbd9
                                                                                                   base
             kernel-spark3.2-scala2.12
pytorch-onnx_1.3-py3.7-edt
                                                    020d69ce-7ac1-5e68-ac1a-31189867356a
                                                                                                   hase
                                                    069ea134-3346-5748-b513-49120e15d288
09c5a1d0-9c1e-4473-a344-eb7b665ff687
             scikit-learn 0.20-py3.6
                                                                                                   base
             spark-mllib_3.0-scala_2.12
pytorch-onnx_rt22.1-py3.9
                                                    09f4cff0-90a7-5899-b9ed-1ef348aebdee
0b848dd4-e681-5599-be41-b5f6fccc6471
                                                                                                   base
                                                    Ocdb0f1e-5376-4f4d-92dd-da3b69aa9bda
Oe6e79df-875e-4f24-8ae9-62dcc2148306
             ai-function_0.1-py3.6
             shiny-r3.6
             tensorflow_2.4-py3.7-horovod
                                                    1092590a-307d-563d-9b62-4eb7d64b3f22
                                                                                                   base
            pytorch_1.1-py3.6
tensorflow_1.15-py3.6-ddl
autoai-kb_rt22.2-py3.10
                                                    10ac12d6-6b30-4ccd-8392-3e922c096a92
                                                    111e41b3-de2d-5422-a4d6-bf776828c4b7
                                                                                                   base
                                                    125b6d9a-5b1f-5e8d-972a-b251688ccf40
                                                    12b83a17-24d8-5082-900f-0ab31fbfd3cb
             runtime-22.1-py3.9
             scikit-learn_0.22-py3.6
                                                    154010fa-5h3h-4ac1-82af-4d5ee5abbc85
                                                                                                   hase
                                                    1b70aec3-ab34-4b87-8aa0-a4a3c8296a36
1bc6029a-cc97-56da-b8e0-39c3880dbbe7
             default_r3.6
                                                                                                    base
             pytorch-onnx_1.3-py3.6
                                                                                                   base
             kernel-spark3.3-r3.6
pytorch-onnx_rt22.1-py3.9-edt
                                                   1c9e5454-f216-59dd-a20e-474a5cdf5988
1d362186-7ad5-5b59-8b6c-9d0880bde37f
                                                                                                   base
             tensorflow_2.1-py3.6
spark-mllib_3.2
                                                    1eb25b84-d6ed-5dde-b6a5-3fbdf1665666
20047f72-0a98-58c7-9ff5-a77b012eb8f5
             tensorflow 2.4-pv3.8-horovod
                                                    217c16f6-178f-56bf-824a-b19f20564c49
                                                                                                   base
             runtime-22.1-py3.9-cuda
                                                    26215f05-08c3-5a41-a1b0-da66306ce658
             do py3.8
                                                    295addb5-9ef9-547e-9bf4-92ae3563e720
                                                                                                   base
             autoai-ts 3.8-pv3.8
                                                    2aa0c932-798f-5ae9-abd6-15e0c2402fb5
                                                                                                   base
             tensorflow_1.15-py3.6
                                                    2b73a275-7cbf-420b-a912-eae7f436e0bc
                                                                                                    base
             kernel-spark3.3-pv3.9
                                                    2h7961e2-e3h1-5a8c-a491-482c8368839a
                                                                                                   hase
                                                    2c8ef57d-2687-4b7d-acce-01f94976dac1
2e51f700-bca0-4b0d-88dc-5c6791338875
             pytorch_1.2-py3.6
spark-mllib_2.3
                                                                                                   base
             pytorch-onnx_1.1-py3.6-edt
spark-mllib_3.0-py37
                                                    32983cea-3f32-4400-8965-dde874a8d67e
36507ebe-8770-55ba-ab2a-eafe787600e9
                                                                                                   base
             spark-mllib_2.4
autoai-ts_rt22.2-py3.10
                                                    390d21f8-e58b-4fac-9c55-d7ceda621326
396b2e83-0953-5b86-9a55-7ce1628a406f
                                                                                                   base
             xgboost 0.82-py3.6
                                                    39e31acd-5f30-41dc-ae44-60233c80306e
                                                                                                   base
                                                    40589d0e-7019-4e28-8daa-fb03b6f4fe12
40e73f55-783a-5535-b3fa-0c8b94291431
             pytorch-onnx_1.2-py3.6-edt
             pytorch-onnx_rt22.2-py3.10
                                                                                                   base
             default r36pv38
                                                    41c247d3-45f8-5a71-b065-8580229facf0
                                                    4269d26e-07ba-5d40-8f66-2d495b0c71f7
             autoai-ts_rt22.1-py3.9
                                                                                                   base
             autoai-obm 3.0
                                                    42b92e18-d9ab-567f-988a-4240ba1ed5f7
                                                                                                   base
             pmml-3.0_4.3
spark-mllib_2.4-r_3.6
                                                    493bcb95-16f1-5bc5-bee8-81b8af80e9c7
49403dff-92e9-4c87-a3d7-a42d0021c095
                                                                                                   base
             xgboost_0.90-py3.6
pytorch-onnx_1.1-py3.6
                                                   4ff8d6c2-1343-4c18-85e1-689c965304d3
50f95b2a-bc16-43bb-bc94-b0bed208c60b
                                                                                                   base
                                                    52c57136-80fa-572e-8728-a5e7cbb42cde
55a70f99-7320-4be5-9fb9-9edb5a443af5
5c1b0ca2-4977-5c2e-9439-ffd44ea8ffe9
             autoai-ts_3.9-py3.8
spark-mllib_2.4-scala_2.11
             spark-mllib 3.0
                                                                                                   base
             autoai-obm_2.0
                                                    5c2e37fa-80b8-5e77-840f-d912469614ee
                                                    5c3cad7e-507f-4b2a-a9a3-ab53a21dee8b
             spss-modeler_18.1
                                                                                                   base
             cuda-py3.8
runtime-22.2-py3.10-xc
                                                    5d3232bf-c86b-5df4-a2cd-7bb870a1cd4e
                                                                                                   base
                                                    5e8cddff-db4a-5a6a-b8aa-2d4af9864dab
             autoai-kb_3.1-py3.7
                                                   632d4b22-10aa-5180-88f0-f52dfb6444d7
                                                                                                   base
             Note: Only first 50 records were displayed. To display more use 'limit' parameter.
In [73]: software spec uid = client.software specifications.get uid by name("runtime-22.1-py3.9")
            software_spec_uid
Out[73]: '12b83a17-24d8-5082-900f-0ab31fbfd3cb'
In [76]: model_details = client.repository.store_model(model=d,meta_props={
                 client.repository.ModelMetaNames.NAME:"churn deployment",
client.repository.ModelMetaNames.TYPE:"scikit-learn_1.0",
            client.repository.ModelMetaNames.SOFTWARE_SPEC_UID:software_spec_uid})
model_id = client.repository.get_model_id(model_details)
In [77]: model_id
Out[77]: '1ba6b7e9-4cef-4995-81ed-392f4395ce63'
In [78]: x train[0]
Out[78]: array([ 1.49526939, 1.22961301, 1.24359144, 1.38832919, -0.79520768,
                      -1.13752513, -0.73301171])
Out[79]: array([14.56666667])
```

8. Testing

8.1 Test Cases

GIF Gesellschaft für Industry for schung is an automotive company based out of Germany, covering the vehicle transmissions and the design, testing, and development of powertrain system. GSA system, which was developed by GIF, has proven in many years of testing to be a valuable tool, mainly by complex transmission and gear development projects. GSA system is not only applied for objectively assessing the change of gears in a vehicle, but also permits measuring and analyzing the influencing factors of gear shifting quality.

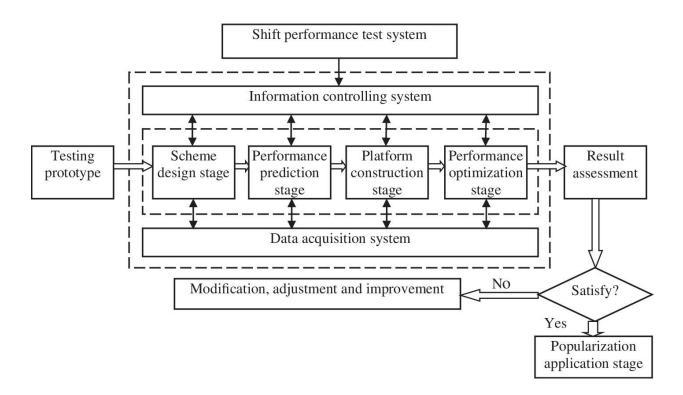
Therefore, during the new development and integration of transmissions, transmission component testing, and quality assurance, GSA system is indispensable as the right tool. Based on GSA shifting performance test evaluation system, this project completes the control strategy research of improving the shifting quality of automobile transmission. The technical route to be adopted. The development process of transmission shifting performance is mainly divided into scheme design stage, performance prediction stage, platform construction stage, performance optimization stage, and popularization application stage.

First, the research idea is clear, the research scheme is determined, and the theoretical research framework of shifting test and evaluation is constructed. Second, relevant materials are collected, the performance parameters of transmission and shifting control mechanism are summarized, and the shifting performance of the whole system is predicted. Third, the shifting control performance bench test and GSA test system are built to carry out the subjective and objective test and research of shifting performance, and the real-time acquisition of shifting performance is carried out.

The performance indicators in the process are analyzed and evaluated to provide data support for improving the shifting performance of the system. Then the optimization improvement measures are specified and the shifting performance is re-evaluated to obtain the best matching relationship of the shifting performance and verify the effectiveness of the 25 control strategy research scheme. Finally, the shifting optimization test and analysis path based on GSA test technology is formed.

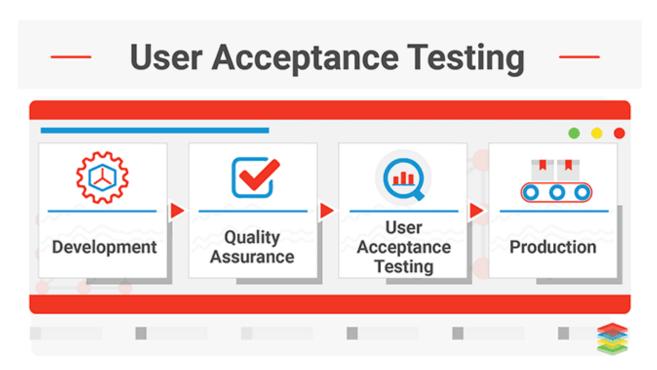
GSA test analysis means that the whole test system includes hardware acquisition equipment and software processing analysis tools. Through real-time measurement and analysis of the force, travel, acceleration, and other important parameters of the

shifting lever knob, the performance of the whole vehicle is systematically evaluated from the aspects of the shifting and selection force (travel), system stiffness, free play, and dynamic impact.



8.2 User Acceptance Testing (UAT)

It is a process to check that system accepts requirements of a user or not. It's performed at a time when the system used by actual users. User acceptance testing comes after \rightarrow Unit Testing \rightarrow Integration Testing \rightarrow System Testing \rightarrow Acceptance Testing in the process of testing.



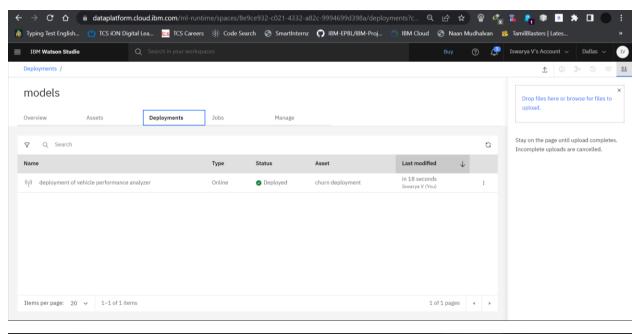
Alpha Testing

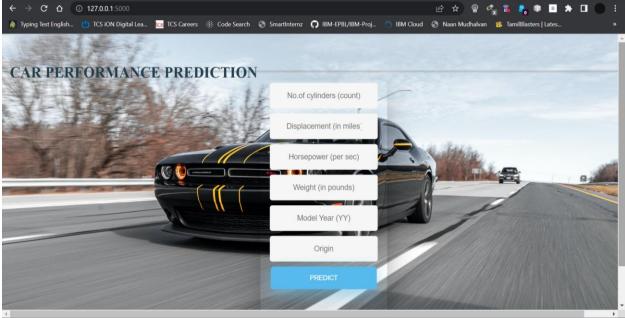
It is the type of testing, executed to identify all possible issues/bugs before releasing a product to every data users or public. It simulates real users by using Black Box and White Box testing techniques. The primary function is to carry out tasks executed by a typical user. Alpha testing carried out in a developed environment or in a lab environment where a product developed and the user of a product are internal employees of an organization.

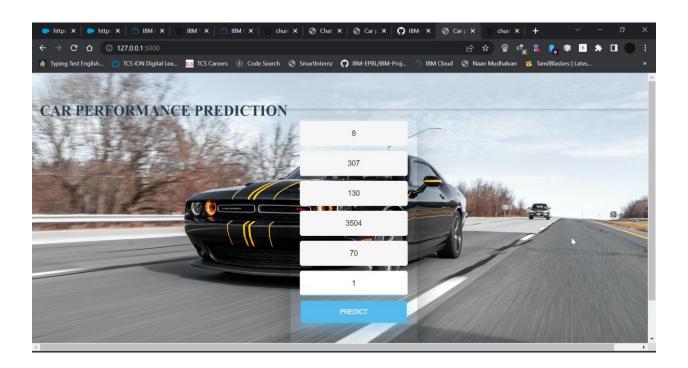
Beta Testing

It is the type of testing in which users of software or application are real users. In this application is tested in a real environment and considered as a form of an external User Acceptance testing. Beta version of a software released to a limited number of a user when tested in a Real-Time environment with the help of real users, to obtain feedback on product quality. Beta testing reduces failures, risks and provides increased quality of a product through customer validation. It is a final test before shipping a product to the customers. In this type of testing, getting direct feedback from users is a significant benefit. It is required to test a product finally in a Real-Time environment.

9.RESULTS







10. ADVANTAGES & DISADVANTAGES

ADVANTAGES

- Easy Implementation
- Low cost
- Can know and maintain the driver and car performance

DISADVANTAGE

• It is very difficult to find the place for placing temperature sensor and pressure sensor.

11.CONCLUSION

In this paper we have obtained some newer insights about the car data analysis such as economic driving index (ECN_DRVG_INDX) and safety driving index (SFTY_DRVG_INDX.) The results have proven to be approximately 80% fitting the given features and are very helpful to be used in different use cases such as a parameter in finding the driver's driving performance in a driving school, as a good estimate for finding an optimal price for a used car that can be based on several factors which we have analyzed in this paper etc. We also found that the model used to train the data can be improved further by finding better hyper parameter values for the features. It is also possible that different features can be considered for improving the hypothesis.

12. FUTURE SCOPE

There is more scope in future for research and analysis of fuel efficiency by including other factors like the road condition and real-time traffic with the help of google maps, this would help in analyzing much deeper. The knowledge discovered from the research and future work can be used by the car manufacturing companies to improve the fuel economy by considering the characteristics that substantially influence the fuel efficiency.

	13. APPENDIX							
(GitHub link: <u>https://gith</u>	nub.com/IBM-EPBL/IB	M-Project-5893-1658819345					