

Title: **Predicting CO2 emission in Vehicles**

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Introduction

Overview

With the development of the world economy, the situation of carbon dioxide emissions control is becoming increasingly serious. CO₂ emissions are the primary driver of global climate change. It is widely recognized to avoid the worst impacts of climate change. A typical vehicle emits about 4.6 metric tons of carbon dioxide per year. This causes global warming and effects the environment. The number can vary based on a vehicle's fuel, fuel economy, and the number of miles driven per year. When gasoline in the vehicle burns, the carbon and hydrogen separate. The hydrogen combines with oxygen to form water (H₂O), and carbon combines with oxygen to form carbon dioxide (CO₂). This is how CO₂ is emitted from vehicles.

This project will be predicting the amount of CO₂ released by the vehicles by exploring the inputs of the vehicle such as engine-size, fuel-consumption etc.

Purpose

The goal of the project is to develop an emission predictor which predicts the amount of the CO₂ emitted by a vehicle with certain characteristics. The model will be trained and tested using Machine Learning Techniques. The predictor is built by using Flask applications. HTML and CSS are used for creating and designing the webpage.

Literature Survey

Existing problem

CO₂ emissions is highly renewable, qualitative and quantitative. Exposure to high CO₂ levels in vehicles results in unpleasant feeling, fatigue, drowsiness or lethargy among the drivers and passengers. CO₂ emitted from vehicles can be reduced little by air filtration. The CO₂ which is emitted from vehicles causes climate change by trapping heat, and also leads to respiratory diseases in the

human beings. Extreme weather, food supply disruptions, and increased wildfires are few effects of climate change caused by CO₂ emissions.

Carbon emissions include a number of different chemicals and particulates that are produced when fuel is burned in an engine. Some of the major substances found in a car's exhaust include carbon dioxide, ozone, and carbon monoxide. Other chemicals often found in exhaust gasses include benzene and nitrogen oxides. Many of these chemicals serve an important purpose in different parts of the atmosphere, but they can have bad consequences when human beings inhale them directly. Mitigation of Carbon Dioxide emission is the challenge of the future in order to stabilize global warming. The factors affecting CO₂ emissions of vehicles include the GDP, population, urbanization rate, transportation development level, transportation energy intensity, energy consumption structure, and industrial structure. The level of traffic development is a comprehensive indicator, and there exist some differences in how to measure it quantitatively. The classical regression analysis requires the independent variables to be linearly independent.

The carbon emissions are calculated according to the total amount of smoke. The amount of CO₂ emission from the transport sector (including cars) accounts for about 20% of total CO₂ emissions. Accordingly, from the viewpoint of preventing global warming, reducing that proportion is a key issue. In regard to CO₂ emissions from cars, fuel economy standards are getting tougher all over the world, so improving the fuel economy of cars is strongly desired. It is considered that the fuel economy of engines will be further improved by boosting engine efficiency and by hybridization or electrification of cars. Improving fuel economy by improving "driving operation" (i.e. the operation in which a car is driven) and by smoothing traffic flows will come into the picture in the near future.

Vehicle emissions have a number of harmful effects on human health. Exposure to car emissions increases the risk of getting certain cancers. Carbon traps heat in the atmosphere, preventing it from escaping from the earth. This has led to a warmer earth over the past century, increasing the odds of severe weather patterns, droughts, and other problems.

This problem can be rectified by using Machine Learning Techniques by predicting the amount of CO₂ emitted from a vehicle by giving few inputs or features of the vehicle. By doing this prediction, the amount of CO₂ from the vehicles can be known and reduced by doing few measures.

Proposed solution

Significantly reducing CO2 emissions from cars will not be easy, but the available data can be used to extract the features, know the behaviour of cars, and try to reduce the emissions. Machine Learning techniques can be used in this regard. The solution to the above discussed problems is to build a predictor which predicts the amount of CO2 released by a vehicle. The vehicle CO2 emission model is derived based on the theory of vehicle dynamics. When the CO2 emission predictor predicts the amount of CO2 released, then the car might be modified or replaced with parts which cause less amount of CO2 to be released.

In this project, we take a dataset which contains features of the vehicles like

- Type of car
- Car class
- Engine size
- Cylinder size
- Transmission of the car
- Fuel Type
- Fuel Consumption
- Combine Fuel Consumption
- Fuel Consumption highway

Now using machine learning techniques like Linear Regression, we prepare the model. The following steps are to be performed to the dataset:

- Download the dataset: Machine Learning depends heavily on data, without data, it is impossible for a machine to learn. It is the most crucial aspect that makes algorithm training possible.
- Pre-process the data:
 - i. contains handling the null values
 - ii. handling the categorical values if the dataset contains
 - iii. normalizing the data wherever required
 - iv. identifying the independent and dependent variables in the dataset
 - v. splitting the dataset into Train and Test sets. These tests are used while calculating accuracy for the dataset and predicting the output i.e. the CO2 emitted value from the vehicle.
- Analyze the pre-processed data: involves the dropping few columns we

- don't require for our prediction, understanding the data type and summary of the features, observing numerical and categorical values and converting into numerical values if any required.
- Train and test the machine with pre-processed data: Data is to be split into dependent and independent variables. The dependent variable is nothing but output in the dataset and the independent variable is all inputs in the dataset. The data can be divided into train and test sets by considering 80% of the data for training and 20% for testing. The split can be done by passing an argument "random_state".
- Save the model and its dependencies: Pickle is used for serializing and de-serializing Python object structures called as flattening. Serialization refers to the process of converting an object in memory to a byte stream that can be stored on disk or sent over a network. Later on, this character stream can then be retrieved and de-serialized back to a Python object.
- Build a Web application using flask that integrates with the model built.

In order to build a web application using flask we need to create folders named

- Static: this folder contains the images and the CSS styles for the web page we designed.
- Templates: this folder contains the .html files we created for building our web page.
- Training: this folder contains the .ipynb file we created and trained and tested the dataset and performed data visualization techniques.
- The main folder contains the .pkl file i.e. the pickle file.
- After training and testing the model, the model should be dumped and saved. This file should be placed in the main folder.

Some of the functions used in Flask are:

- app – our flask application name
- model – it will contain the model which we build
- @app.route() – it is a decorator that can redirect to different functions.
 - i. one for routing to the home page ('/')
 - ii. route to the prediction page ('/Prediction')
 - iii. route to the same home page itself ('/home1')
 - iv. routing to the result page ('/predict')
- render_template () – it is used for render our html page from the templates folder
- predict () – is taking the values form the prediction page and storing it into a variable and then we are creating a Data Frame along with the

values and 9 independent features and finally we are predicting the values using or loaded model which we build and storing the output in a variable and returning it to the result page.

- `app.run(debug=False)` – for running our app

The code for the Flask application is provided in the appendix.

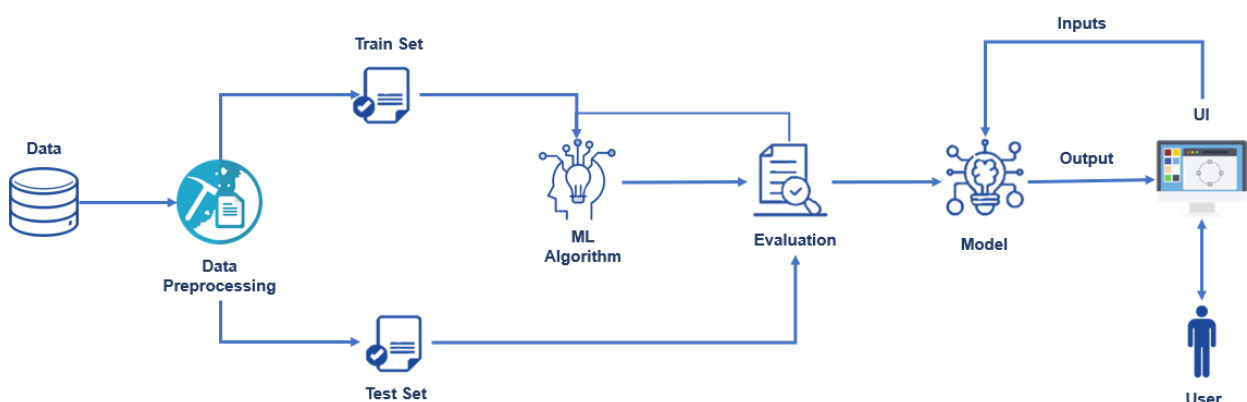
Using the web applications, we build three html pages

1. The home page: contains the introduction of the project. This page navigates to the prediction page where we give inputs and predict our model.
2. The prediction page: contains the inputs of the vehicles i.e. the features of the cars. This page navigates to the final page where the CO2 emitted value is shown.
3. The final page: contains the value of the predicted CO2 emitted by a vehicle.

By doing all these applications we can predict the amount of CO2 emitted by a vehicle and decrease the adverse effects caused by CO2 emission. This is the solution for the above discussed problem.

Theoretical Analysis

Block Diagram

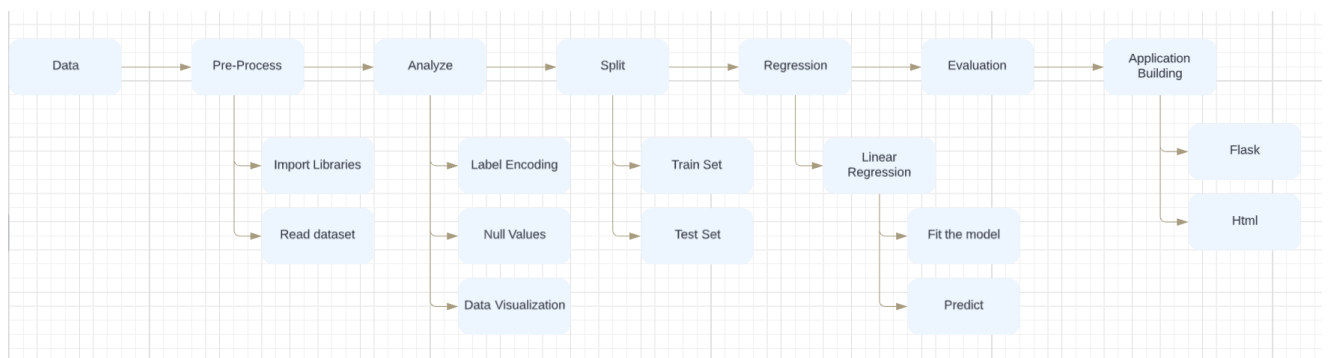


The above diagram is a representation of how the project works. The data is pre-processed. The data is divided into train and test sets. Using ML algorithm, the

data is evaluated. The ML algorithms include regression and classification and, in this project, linear regression is used as the output is in continuous data. The model is then used for prediction. When the user gives the inputs to the model, the model predicts the amount of CO2 released by the vehicle and gives the output to the user. The model uses Linear Regression and fits the input train and output train into the regression.

For the output to be displayed, web applications are built using Flask. The first page will be the introduction of the project and this page navigates the user to the input pages where the user can give the features of the vehicle and predict the CO2 emission. This page navigates the user by one click on the predict button to the final page where the user will be able to see the predicted value depicted by the model which is obtained by the Machine Learning Algorithms.

Software Designing



The above picture is a design of how the project works.

- The data is loaded and pre-processed.
- While pre-processing the data, the required libraries are imported, the dataset is read.
- The next step is to analyze the dataset. Analysing the dataset includes:
 - Handling the null values i.e. checking if the dataset contains any null values or not, if contained replacing the null values with a specific value in the dataset.
 - Performing data visualization techniques like plotting graphs between the input or the output to understand the dataset. Few data visualization techniques involve bar-graph, bar-plot etc.
 - Finding the correlation between the independent variables. Correlation is a statistical relationship between two variables and it

could be positive or negative. The correlation can be found by using heat map.

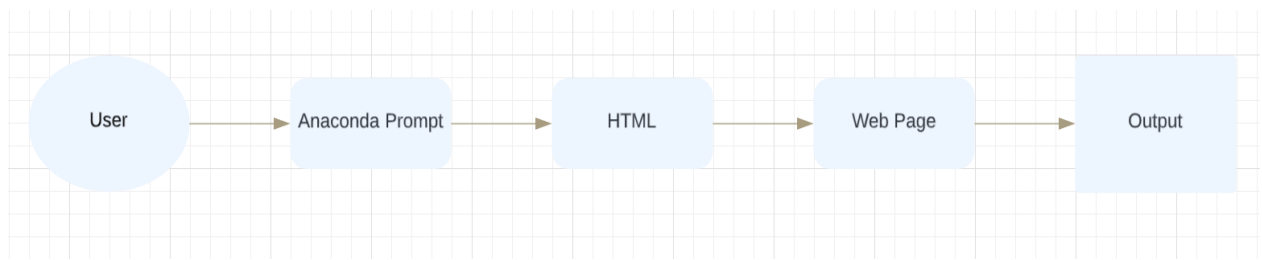
- i. Positive: both variables move in the same direction
 - ii. Negative: both variables are inversely proportional i.e. when the value of one variable increases, the values of the other one decreases.
- Label Encoding: this converts the categorical values to numerical values.
 - The dataset is to be split into train and test sets. 80% of the data is trained and 20% of the data is tested.
 - Next to the dataset the Machine learning algorithms are to be performed. Since the dataset is continuous dataset Regression is used. Using Linear Regression, the model is fitted.
 - Syntax: `variable_name = LinearRegression ()`

`variable_name=variable_name.fit (x_train, y_train)`

- Once the model is trained, the model is ready to predict. We used “predict method” on the model and pass the parameter: `x_test`. The predicted value i.e. the output will be stored in “`y_pred`”.
- The next step in the project is to evaluate the model. This can be done by calculating:
 - i. Mean Absolute Error: measure of errors between paired observations expressing the same phenomenon
 - ii. Mean Squared Error: average of the squared difference between the target value and the value predicted by the regression model.
 - iii. Root Mean Square Error: is the square root of the averaged squared difference between the target value and the value predicted by the model.

After evaluating the model, the model is saved and dumped using the pickle file.

- The last step in the design is to run the application. This is done using the Flask application. More detailed information is given in the following picture and description.



The above diagram is the representation of Flask application.

- The user accesses anaconda prompt and navigates to the folder where the files are located.
- Next is to navigate to the localhost where one can view the web page.
- It runs on the localhost:5000
- Localhost navigates to the output page.
- Three pages of the web page are:
 - i. Home page: introduction of the web page.
 - ii. Predict page: inputs of the vehicles are given.
 - iii. Prediction page: the predicted value of the CO2 emission is displayed.

Experimental Investigations

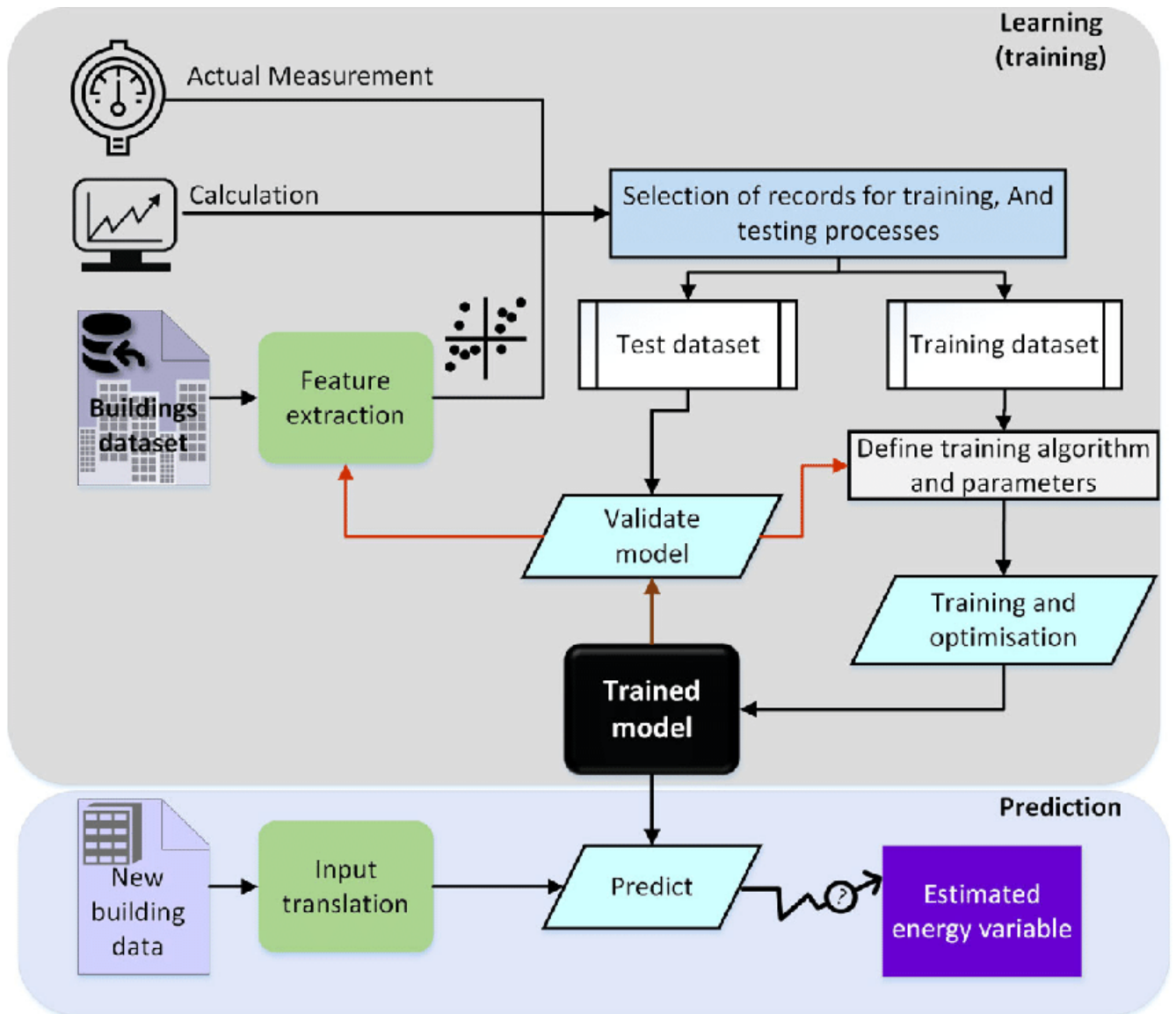
Various test cases were developed for each test scenario to check the correctness of the predicted value. Manual testing was also performed for each of the test case where the inputs are given and executed without any tool. The flask application used in the project also gives the value which is similar to the manual tested value. The given table below are few features from the dataset and CO2 emission value is predicted and calculated by manual.

Test Case	Input									Expected Result	Actual Result	Status
	Type	Class	Engine Size	Cylinder Size	Transmission	Fuel Type	Fuel Consumption City	Fuel Consumption Highway	Combine Fuel			
1	Audi	Full Size	2	4	AV6	Diesel	9.9	6.7	33	173.66	173.62	Pass
2	BMW	Mini van	3.7	6	M7	Natural Gas	6	5.8	25	216.21	216.21	Pass
3	Fiat	Mid size	1.8	12	A6	Ethanol	11.2	7.5	28	265.68	265.68	Pass
4	Honda	SUV small	5.9	8	AM5	Regular Gasoline	13.4	12.6	27	299.54	299.55	Pass
5	Jeep	VAN cargo	4.7	4	AS8	Natural Gas	17.4	11.3	29	260.28	260.23	Pass
6	Kia	Mini van	5.9	10	M7	Diesel	9.9	7.4	18	283.91	283.91	Pass
7	Nissan	Full size	2	8	AS10	Ethanol	11.5	8.1	19	266.411	266.4	Pass
8	Scion	SUV small	2.4	6	AV7	Regular Gasoline	12.8	7.5	30	249.44	249.44	Pass

The above table depicts the experimental investigations of the project.

- The table contains input of the features and the expected and actual value of the CO2 emitted by the project. The last column judges if the actual and expected result are similar and gives a pass/fail status.
- The model predicts the CO2 value based on the inputs given.
- Pass status indicates that the expected and actual result is similar to each other.
- There are many other features of the vehicles but for instance a few are taken and the CO2 is calculated.

Flowchart



- The above diagram is the flowchart of predicting the value of CO₂ emitted from the vehicles.
- The dataset is built and the features of the dataset are extracted. Then the dataset is divided into records: train dataset and test dataset.
- From the test dataset the model is validated.
- From the train dataset, the training algorithms and the parameters are defined.
- The validated model is compared with the algorithms. From the training algorithm, the model is optimized and sent to the trained model.

- The process repeats. The trained model is sent to the prediction to predict the CO2 emitted.

Result

The results of the project showed that the CO2 emitted by a vehicle was predicted by the model. The model was able to predict the value of CO2 emitted when various features of the vehicle are given to the model as input. The prediction provides guidance for the realization of the on-line monitoring. In case, if the input values are given as null the model gives an error and asking the user to provide all the details of the vehicle. The model is able to predict the correct amount of CO2 emitted and this can be found in the experimental investigations we have calculated. This project helps the researchers to predict the amount of CO2 released by a vehicle having specific properties. The model can still be developed by adding more inputs.

Advantages & Disadvantages

Advantages

There are many advantages of predicting CO2 emission from vehicles. The first main advantage of the prediction is it can be used to reduce the CO2 emission from the featured vehicle. By reducing the CO2 emission, risk to human health is decreased and global warming is reduced and there will no variant in the climate change. The reduction of CO2 also reduces Air pollution caused by the heavy traffic and the emissions from all the vehicles. This leads to a clean and healthy environment. Features of the vehicle can be changed according to the generated CO2 value and therefore the CO2 can be reduced. Another advantage of this prediction is it is cost effective and can be done by any individual.

Disadvantages

There are few disadvantages of the prediction too. Sometimes the prediction value might vary with the actual value and this can result in false predictions. Another disadvantage is that when the user tries to give the input and cannot find their respective feature of the vehicle, the model becomes ineffective. The prediction requires more memory since it has to store all the dataset values, the html pages and the CSS format which are used to build the application. Few software's have also to be installed to perform all the predictions and

calculations to obtain the final output. Many applications are to be run to generate the predicted value and observe the value. This takes more space in the system. To build an application, one must also be aware of the html web applications to construct the final web page. Many features can be added to the model still but it requires lot of work and time. The proposed model performs better for large training set. But it takes more time to train model.

Applications

The prediction can be done for both classification and regression models. Being a CO₂ emission predictor, it can be helpful to traffic police. This can be done when the police realize a vehicle is emitting more amount of CO₂, they can predict the value and make the vehicle to stop to avoid CO₂ into the environment. The prediction can be integrated with the company who is working to reduce the CO₂ emission from the vehicles and also can be of great use to the pollution control.

Conclusion

Based on the analysis of vehicle features, and by the proximate analysis data to predict fuel characteristic factor, the calculation method of predicting CO₂ emission of vehicles is established. Successfully we were able to predict the CO₂ Emissions for the dataset with 86% accuracy by calculating the r^2_score of the dataset. Prediction with high accuracy can give information concerning about CO₂ emissions. The available data can be used to extract the features, know the behaviour of cars, and try to reduce the emissions. The model is able to extract the CO₂ emitted value by all the features and also is able to generate and error when the user tries to give a null value as an input.

Future Scope

The scope of the CO₂ emission predictor can be extended by adding new features to the model and saving space and memory used to build the model. The application could also include text facility for the users to give details about the vehicle. New Machine Learning techniques can we used to predict the model easily and calculate the predict value. It may also allow software

applications to become accurate in predicting outcomes. It enables machines to make data-driven decisions, which is more efficient than explicitly programming to carry out certain task.

Bibliography

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- [4] Machine Learning based Prediction System for Detecting Air Pollution (ijert.org)
- [5] Machine learning approaches for the prediction of materials properties: APL Materials: Vol 8, No 8 (scitation.org)
- [6] Prediction of CO2 Emissions Based on Multiple Linear Regression Analysis (sciencedirectassets.com)

Appendix

Source Code

Flask Code

app.py

```
from flask import Flask, request, render_template
import numpy as np
import pandas as pd
import pickle
import joblib
app=Flask(__name__)
model=pickle.load(open("CO2.pkl", 'rb'))
LE=joblib.load('column2')
@app.route('/')
def home():
    return render_template('home1.html')
@app.route('/Prediction', methods=['POST', 'GET'])
```

```

def prediction():
    return render_template('indexnew.html')
@app.route('/Home',methods=['POST','GET'])
def my_home():
    return render_template('home1.html')

@app.route('/predict',methods=["POST","GET"])
def predict():
    Make = float(request.form["Make"])
    Vehicle_class = float(request.form["Vehicle_class"])
    Engine_Size = float(request.form["Engine_Size"])
    Cylinders = float(request.form["Cylinders"])
    Transmission = float(request.form["Transmission"])
    Fuel_Type = float(request.form["Fuel_Type"])
    Fuel_Consumption_City = float(request.form["Fuel_Consumption_City"])
    Fuel_Consumption_Hwy = float(request.form["Fuel_Consumption_Hwy"])
    Fuel_Consumption_Comb = float(request.form["Fuel_Consumption_Comb"])

    data=[[Make,Vehicle_class,Engine_Size,Cylinders,Transmission,Fuel_Type,Fuel
_Consumption_City,Fuel_Consumption_Hwy,Fuel_Consumption_Comb]]
    x=model.predict(data)
    return render_template("resultnew1.html",y=x[0][0])
if __name__=="__main__":
    app.run(debug=True)

```

HTML codes

home1.html

```

<!DOCTYPE html>
<html>
<head>
<style>
    body {
        background-size: cover;
        background-repeat: no-repeat;
        background-imageurl({{url_for('static',filename='builtin/new.jpg')}})
    }
</style>
<p align="center" style="font-size:48px">
    <b>CO2 EMISSION FROM VEHICLES</b>
</p>
<p style="font-size:24px">

```

The amount of CO₂ emission from the transport sector accounts for about 20% of total CO₂ emissions.</br>

Regarding CO₂ emissions from cars, the fuel economy standards are getting tougher all over the world and this can be controlled by improving the fuel economy of cars.</br>

The fuel economy of engines will be further improved by boosting engine efficiency and by hybridization of cars.

A typical passenger vehicle emits about 4.6 metric tons of carbon dioxide per year.</br>

Reduction of CO₂ emission from cars can be done by making vehicles more efficient or by changing the fuel used.

</p>

```
<form method="post" action="/Prediction">
<br>
<br>
<br><table align="center">
  <tr><td>
<input type="submit" name="PREDICT" value="PREDICT"
onclick="validate();">
</td></tr>
<style>
  input[type = submit] {
    background-color: maroon;
    border: none;
    text-decoration: none;
    color: white;
    font-size: 36px;
    font-family: "Times New Roman";
    padding: 10px 10px;
    margin: 10px 10px;
    cursor: pointer;
  }
</style>
</head>
</html>
```

indexnew.html

```
<!DOCTYPE html>
<html>
<head>
<style>
  body {
    background-size: cover;
    background-repeat: no-repeat;
    background-image:url({ {url_for('static',filename='builtin/second.jpg')}})
  }
```



```

</style>

<table align="center" border=3 bgcolor="lightgrey">
<tr>
<h1 align="center" style>CO2 Cars Emission Predictor</h1>
<form method="post" action="/predict">
<tr><td>
<h3 align="center">Select the type of CAR</h3>
<td><select id="Make" name="Make">
  <option value="0">ACURA</option>
  <option value="1 ">ALFA ROMEO</option>
  <option value="2">ASTON MARTIN</option>
  <option value="3">AUDI</option>
  <option value="4">BENTLEY</option>
  <option value="5">BMW</option>
  <option value="7 ">BUICK</option>
  <option value="8">CADILLAC</option>
  <option value="9 ">CHEVROLET</option>
  <option value="10 ">CHRYSLER</option>
  <option value="11 ">DODGE</option>
  <option value="12 ">FIAT</option>
  <option value="13 ">FORD</option>
  <option value="15 ">GMC</option>
  <option value="16 ">HONDA</option>
  <option value="17 ">HYUNDAI</option>
  <option value="18 ">INFINITI</option>
  <option value="19 ">JAGUAR</option>
  <option value="20 ">JEEP</option>
  <option value="21 ">KIA</option>
  <option value="22 ">LAMBORGHINI</option>
  <option value="23 ">LAND ROVER</option>
  <option value="24 ">LEXUS</option>
  <option value="25 ">LINCOLN</option>
  <option value="26 ">MASERATI</option>
  <option value="27 ">MAZDA</option>
  <option value="28 ">MERCEDES-BENZ</option>
  <option value="29 ">MINI</option>
  <option value="30 ">MITSUBISHI</option>
  <option value="31 ">NISSAN</option>
  <option value="32 ">PORSCHER</option>
  <option value="33 ">RAM</option>
  <option value="34 ">ROLLS-ROYCE</option>
  <option value="35 ">SCION</option>
  <option value="36 ">SMART</option>

```

```

<option value="37 ">SRT</option>
<option value="38 ">SUBARU</option>
<option value="39 ">TOYOTA</option>
<option value="40 ">VOLKSWAGEN</option>
<option value="41 ">VOLVO</option>
</select>
</td></tr>
<tr><td>
<h3 align="center">Select the Car Class </h3>
<td><select id="Vehicle_class" name="Vehicle_class">
  <option value="0 ">COMPACT</option>
  <option value="1 ">FULL-SIZE</option>
  <option value="2 ">MID-SIZE</option>
  <option value="3 ">MINICOMPACT</option>
  <option value="4 ">MINIVAN</option>
  <option value="5 ">PICKUP TRUCK - SMALL</option>
  <option value="6 ">PICKUP TRUCK - STANDARD</option>
  <option value="7 ">SPECIAL PURPOSE VEHICLE</option>
  <option value="8 ">STATION WAGON - MID-SIZE</option>
  <option value="9 ">STATION WAGON - SMALL</option>
  <option value="10 ">SUBCOMPACT</option>
  <option value="11 ">SUV-SMALL</option>
  <option value="12 ">SUV - STANDARD</option>
  <option value="13 ">TWO-SEATER</option>
  <option value="14 ">VAN - CARGO</option>
  <option value="15 ">VAN - PASSENGER</option>
</tr>
<td><h3 align="center">Enter the engine size </h3>
<td><input type="text" name="Engine_Size"></td>
</tr>
<tr>
<td><h3 align="center">Enter the Cylinder size </h3>
<td><input type="text" name="Cylinders"></td>
</tr>
<tr><td>
<h3 align="center">Select transmission of the car </h3>
<td><select id="Transmission" name="Transmission">
  <option value="1 ">A6</option>
  <option value="2 ">A7</option>
  <option value="3 ">A8</option>
  <option value="4 ">A9</option>
  <option value="0 ">A10</option>
  <option value="4 ">AM5</option>
  <option value="6 ">AM6</option>

```

```

        <option value="7">AM7</option>
        <option value="8">AM8</option>
        <option value="3">AM9</option>
        <option value="10 ">AS10</option>
        <option value="11">AS6</option>
        <option value="12">AS7</option>
        <option value="13 ">AS8</option>
        <option value="14 ">AS9</option>
        <option value="15">AV10</option>
        <option value="16 ">AV6</option>
        <option value="17 ">AV7</option>
        <option value="18">AV8</option>
        <option value="22 ">M7</option>

<tr><td>
<h3 align="center">Select the Fuel TypeRegular Gasoline </h3>
<td><select id="Fuel_Type" name="Fuel_Type">
        <option value="3">Premium Gasoline</option>
        <option value="0">Diesel</option>
        <option value="4">Regular Gasoline</option>
        <option value="1">Ethanol</option>
        <option value="2">Natural Gas</option>
</tr></td>
<tr>
<td><h3 align="center" >Enter the Fuel Consumption City(L/100km)
</h3>
<td><input type="text" name="Fuel_Consumption_City"></td>
</tr>
<tr>
<td><h3 align="center">Enter the Fuel Consumption Highway(L/100km)
</h3>
<td><input type="text" name="Fuel_Consumption_Hwy"></td>
</tr>
<tr>
<td><h3 align="center">Enter the Combine Fuel Consumption(mpg) </h3>
<td><input type="text" name="Fuel_Consumption_Comb"></td>
</tr>
</select>
<table align="center">
<tr><td>
<input type="submit" value="PREDICT" onclick="validate();">
<style>
        input[type = submit] {
            background-color: maroon;

```

```

        border: none;
        text-decoration: none;
        color: white;
        font-size: 22px;
        font-family: "Times New Roman";
        padding: 10px 10px;
        margin: 10px 10px;
        cursor: pointer;
    }
</style>
</td></tr>
</table>
</td>
</form>
</tr>
</table>
</td>
</tr>
</td>
</tr>
</select>
</form>
</td>
</tr>
</table>
</table>
</head>
</html>

```

resultnew1.html

```

<!DOCTYPE html>
<html>
<head>
<style>
    body {
        background-size: cover;
        background-repeat: no-repeat;
        background-image: url({url_for('static',filename='builtin/picture.jpg')})
    }
</style>
<br><br><table align="center" border=3>
    <tr><td>

```

```

<h1 align="center">
    <br>
    CO2 CARS EMISSION PREDICTOR
</h1></td></tr>
    <td><h2 align="center" style="background-color: white">CO2 Emission
by this car is {{y}} </h2>
    </td>
</table>
</head>
</html>

```

Output

Open the anaconda prompt and navigate to the folders

```

Anaconda Prompt (anaconda3) - python app.py

(base) C:\Users\veera>cd OneDrive
(base) C:\Users\veera\OneDrive>cd desktop
(base) C:\Users\veera\OneDrive\Desktop>cd flask
(base) C:\Users\veera\OneDrive\Desktop\Flask>python app.py
* Serving Flask app "app" (lazy loading)
* Environment: production
  WARNING: This is a development server. Do not use it in a production deployment.
  Use a production WSGI server instead.
* Debug mode: on
* Restarting with windowsapi reloader
* Debugger is active!
* Debugger PIN: 328-712-589
* Running on http://127.0.0.1:5000/ (Press CTRL+C to quit)

```

In the browser type “localhost:5000”



The output:

Introduction Page:



- This is the first page called the introduction page.
- By clicking on the predict button, it navigates to the prediction page as follows.

Prediction Page:

Before entering the inputs:

CO2 Cars Emission Predictor

Select the type of CAR	ACURA ▼
Select the Car Class	COMPACT ▼
Enter the engine size	<input type="text"/>
Enter the Cylinder size	<input type="text"/>
Select transmission of the car	A6 ▼
Select the Fuel TypeRegular Gasoline	Premium Gasoline ▼
Enter the Fuel Consumption City(L/100km)	<input type="text"/>
Enter the Fuel Consumption Highway(L/100km)	<input type="text"/>
Enter the Combine Fuel Consumption(mpg)	<input type="text"/>

PREDICT

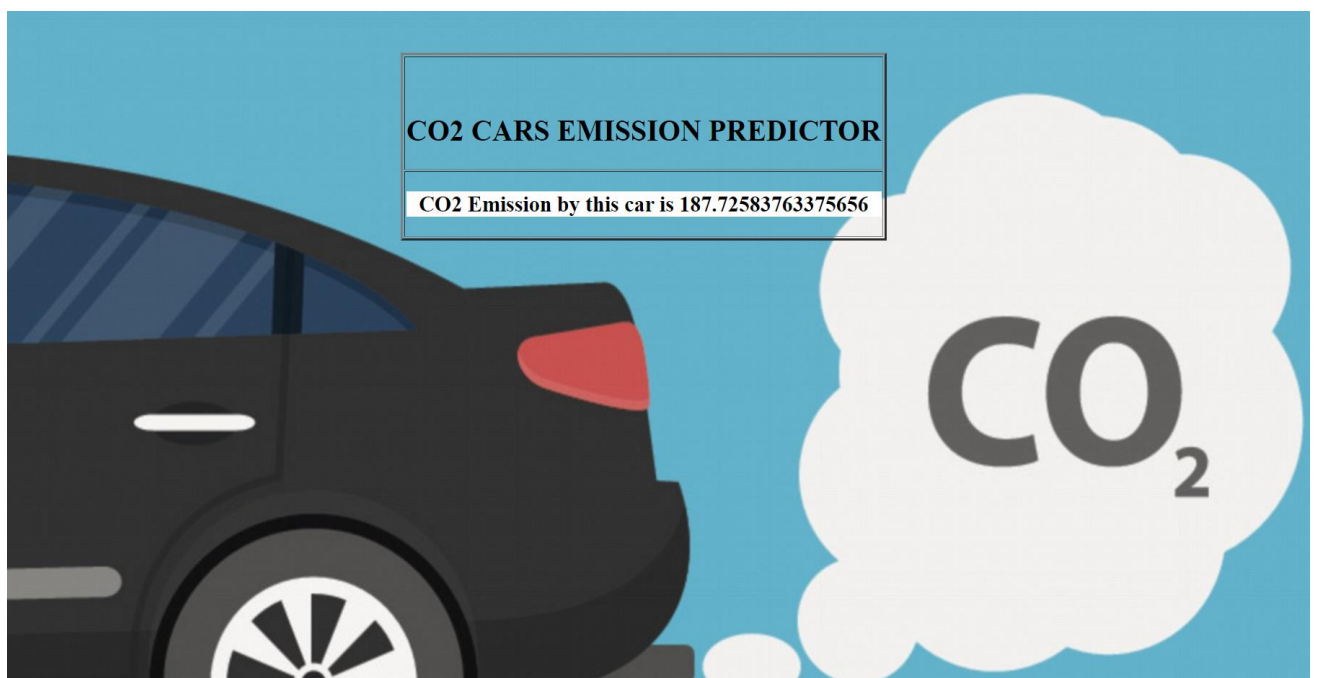
After entering the inputs:

CO2 Cars Emission Predictor

Select the type of CAR	BMW ▾
Select the Car Class	MINIVAN ▾
Enter the engine size	2.4
Enter the Cylinder size	4
Select transmission of the car	AS10 ▾
Select the Fuel TypeRegular Gasoline	Diesel ▾
Enter the Fuel Consumption City(L/100km)	9.9
Enter the Fuel Consumption Highway(L/100km)	7.5
Enter the Combine Fuel Consumption(mpg)	30

PREDICT

Clicking on the predict button, navigates to the predicted value of the CO2 emitted by the entered inputs.



The CO2 emission for the given inputs is 187.72