Vehicle Insurance Prediction

Problem Statement:

The Dataset contains information on policyholders having the attributes like policy tenure, age of the car, age of the car owner, the population density of the city, make and model of the car, power, engine type, etc. You need to develop a machine learning model for identifying the potential policyholders. Potential policyholders are those who may opt for insurance services in future.

Variable	Description
policy_id	Unique identifier of the policyholder
policy_tenure	Time period of the policy
age_of_car	Normalized age of the car in years
age_of_policyholder	Normalized age of policyholder in years
area_cluster	Area cluster of the policyholder
population density	Population density of the city (Policyholder City)
make	Encoded Manufacturer/company of the car
segment	Segment of the car (A/ B1/ B2/ C1/ C2)
model	Encoded name of the car
fuel_type	Type of fuel used by the car
max_torque	Maximum Torque generated by the car (Nm@rpm)
max_power	Maximum Power generated by the car (bhp@rpm)
engine_type	Type of engine used in the car
airbags	Number of airbags installed in the car
is_esc	Boolean flag indicating whether Electronic Stability Control (ESC) is present in the car or not.
is_adjustable_steering	Boolean flag indicating whether the steering wheel of the car is adjustable or not.
is_tpms	Boolean flag indicating whether Tyre Pressure Monitoring System (TPMS) is present in the car or not.
is_parking_sensors	Boolean flag indicating whether parking sensors are present in the car or not.
is_parking_camera	Boolean flag indicating whether the parking camera is present in the car or not.
rear_brakes_type	Type of brakes used in the rear of the car
displacement	Engine displacement of the car (cc)
cylinder	Number of cylinders present in the engine of the car
transmission_type	Transmission type of the car
gear_box	Number of gears in the car

steering_type	Type of the power steering present in the car
turning_radius	The space a vehicle needs to make a certain turn (Meters)
length	Length of the car (Millimetre)
width	Width of the car (Millimetre)
height	Height of the car (Millimetre)
gross_weight	The maximum allowable weight of the fully-loaded car, including passengers, cargo and equipment (Kg)
is_front_fog_lights	Boolean flag indicating whether front fog lights are available in the car or not.
is_rear_window_wiper	Boolean flag indicating whether the rear window wiper is available in the car or not.
is_rear_window_washer	Boolean flag indicating whether the rear window washer is available in the car or not.
is_rear_window_defogger	Boolean flag indicating whether rear window defogger is available in the car or not.
is_brake_assist	Boolean flag indicating whether the brake assistance feature is available in the car or not.
is_power_door_lock	Boolean flag indicating whether a power door lock is available in the car or not.
is_central_locking	Boolean flag indicating whether the central locking feature is available in the car or not.
is_power_steering	Boolean flag indicating whether power steering is available in the car or not.
is_driver_seat_height_adjustable	Boolean flag indicating whether the height of the driver seat is adjustable or not.
is_day_night_rear_view_mirror	Boolean flag indicating whether day & night rearview mirror is present in the car or not.
is_ecw	Boolean flag indicating whether Engine Check Warning (ECW) is available in the car or not.
is_speed_alert	Boolean flag indicating whether the speed alert system is available in the car or not.
ncap_rating	Safety rating given by NCAP (out of 5)
is_claim	Outcome: Boolean flag indicating whether the policyholder file a claim in the next 6 months or not.

Part1-FeatureEngineeringandModelbuilding

1. Data Cleaning:

- oThe dataset is large and it may contain some data error.
- oln order to get clean, error free data some pre-processing requires.
- oRefer dataset.csv.
- oPerform necessary data cleansing, convert data types in appropriate data types, remove duplicates, remove inapplicable columns, change date formats.
- oVisualize missing data using bar plot and heat map. Handle missing values (if and as required).
- oSave the cleaned data.

2. EDA:

- Observe the variables and their distribution with EDA.
- oCreate new columns if required.
- oSave the output data.
- 3. Data Modeling:
 - oHandle the categorical column data.
 - •Prepare categorical features for correlation matrix.
 - oUse label encoding technique i.e. use label encoder to convert categorical values to the numerical values.
- 4. Model Building:
 - oSplit the data in training and testing dataset.
 - •Build following machine learning models.
 - Logistic Regression model
 - ■KNN Model Use
 - n neighbours = 100
 - SVM Model
- 5. Evaluate performance of all these modules.
 - Identify the best performing model for finding potential customers.
 - Save that model
 - Total Expected columns for Model building must be above 85
 - Model Accuracy Must Be below 90%

Part2-PredictionusingFlaskApplication

- Implement a flask -based web application for prediction.
- Load the trained model
- Create an route with("/",method="GET"). This page inputs required information for prediction.
- Create another route ("/Predict"). It should process inputs from user, do the prediction using ML model and display result.

Technologies/Platforms Used

•Application Type: ML Model building and Web application•

Datasets: Dataset.csv

•Technologies:

Python Programming

oMachine learning & data visualization related packages

oFlask for UI

Evaluation

- Students are expected to code in pairs. e.g. For a sub-group size of 10 students, we will evaluate 5 applications.
- Each application will get 1 mark for the full completion and 0 mark if not completed. At the end, we will calculate the percentage score of each group by the formula: group score = 100 * Number of complete programs / expected number or pairs in the group.
- The points will be rewarded to the top 3 groups.

* First group: 10 points

* Second group: 5 points

* Third group: 2 points

• If your application is ready, you should test each functionality thoroughly and then inform the lab mentors. Evaluations will be done in last 1 hour.