



# PROTOTYPE OF “PRICEASY”

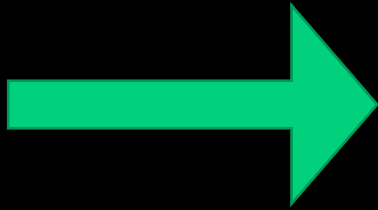
# **DATASET FEATURES FOR VARIOUS ITEMS :**

- ☐ Product Brand
- ☐ Manufacturing year ( for second hand items)
- ☐ Physical condition (scratches , torning etc)
- ☐ Material of the product
- ☐ Since how much time item is being used

# BASIC WORKFLOW :



Data Model



Data Pre-processing



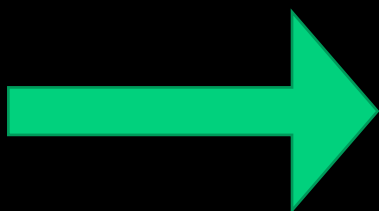
Train Test Split



Linear & Lasso Regression Model



New Data



Trained Linear & Lasso Regression Model




FINAL  
PRICE

Prediction

# ROADMAP OF IMPLEMENTATION :

- ❑ Getting ready made Data model containing standard key features and price in iterative phase
- ❑ Coding part with Jupyter's notebook with Python language
- ❑ Getting basic working idea and familiarity with libraries such as Pandas , Matplotlib , Seaborn
- ❑ Importing Linear Regression model
- ❑ Importing Lasso regression model from sklearn
- ❑ Importing metrics module
- ❑ Data collection and processing
- ❑ Loading .csv file to panda data frame
- ❑ Inspecting data row wise simultaneously

- ❑ Forming predictive logic by quantity of different dependant attributes
- ❑ Also we have to keep record of missing attribute values
- ❑ Checking the distribution of categorical data
- ❑ Encoding this categorical data in numerical values , so that Machine Learning model can understand it.. For example, let processors type for laptop is mainly INTEL or RYZON so it can be encoding like “0” for INTEL , “1” for RYZON .Further for it can be encoded like 101 stands for INTEL i3 , 110 stands for INTEL i5 , 111 stands for INTEL i7 and similar in this kind of other things according to frequency digits can be decided .In this way data will be converted in suitable numbers.
- ❑ Splitting the Data into Training data and Test Data
- ❑ Size and variability in the dataset effects on the accuracy. The larger the data set , the more it will be precise.
- ❑ Model training via Linear regression model .Simply using  $Y = mX + C$

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- ❑ Prediction on Training data
  - ❑ Then evaluating R square error defined as difference between the predicted value and original value. It depends on each on the magnitude of the value that we are using.
  - ❑ Visualization of the actual prices and predicted prices by plotting their values on graph.
  - ❑ Then the same way prediction on Testing data as well as R squared value and visualization.
  - ❑ Then all these will be done with Lasso Regression
  - ❑ By following this steps this project can be implemented at very initial level and then it can be expanded.



# DEMO INTERFACE :

## PRICEASY - THE PRICE CONSULTANT

Enter Specifications for car

Manufacturing Year:

Current Market Price:

Kilometers Driven:

Fuel Type:


- ☒ Petrol
- ☐ Deisel
- ☐ CNG

Transmission Type:

- ☒ Automatic
- ☐ Manual

DISPLAY PRICE





Here, is the price based on your entries.

4.33144922

# Validation in year :

## PRICEASY - THE PRICE CONSULTANT

Enter Specifications for car

Manufacturing Year:

Current Market Price:

Kilometers Driven:

Fuel Type:

- ☒ Petrol
- ☐ Deisel
- ☐ CNG

Transmission Type:

- ☒ Automatic
- ☐ Manual

DISPLAY PRICE

Here, is the price based on your entries.

ENTER year in between range:

# Validation in KMS DRIVEN :

## PRICEASY - THE PRICE CONSULTANT

Enter Specifications for car

Manufacturing Year:

Current Market Price:

Kilometers Driven:

Fuel Type:

- ☒ Petrol
- ☐ Deisel
- ☐ CNG

Transmission Type:

- ☒ Automatic
- ☐ Manual

DISPLAY PRICE

Here, is the price based on your entries.

This is overdriven car, don't go for it.