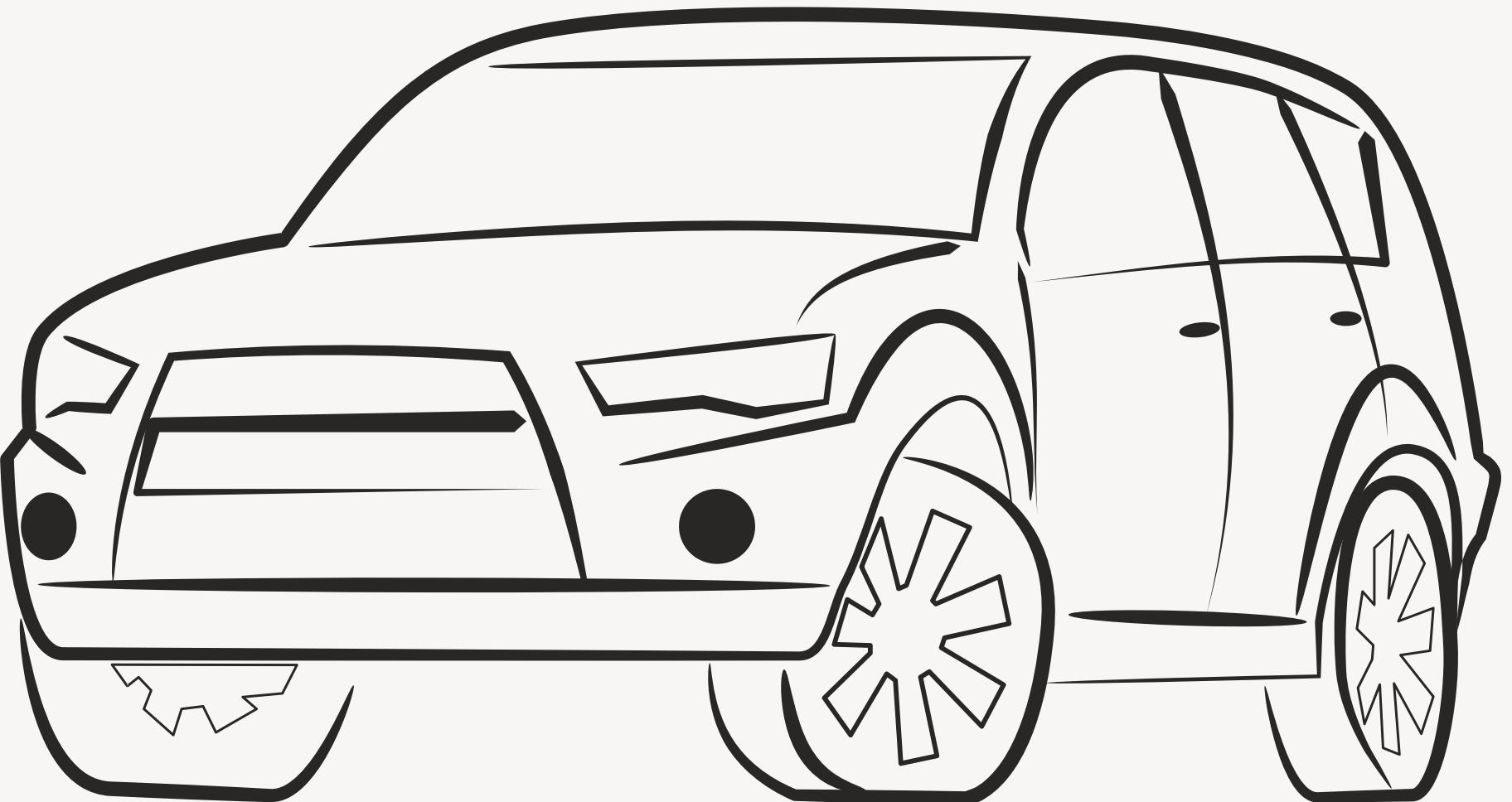


# ML-based Car Price Prediction

Dismantling Data | Building Model





# Agenda

- Project Overview
- Demo - How it will look like?
- Development Process
- Quick Data Insights
- Model Choice & Outcomes
- How model works
- Next Steps
- Q&A



# Our Objective

"To build a model to recommend unit price(\$000) for cars by using past contracts data in year 2160"

## Constraints

- Predict retail price based on retail data as input to a model which includes past contracts.
- None of the current industry information is relevant as we are dealing in 2160 year

# Data Set

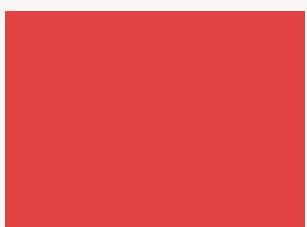
colour	booking_to_delivery_date
plant_nr	unit_price_inK
plant_city	type_of_sale
plant_country_name	date_of_sale
most_expensive_part_l1	mode_of_transport
most_expensive_part_l2	no_of_units
shipto_city	cust_group_name
shipto_nr	cust_segment_name
soldto_city	Customer_Id
soldto_nr	delivery_month_from
car_nr	delivery_month_to
shipto_country	type
soldto_country	item_class_l1
.	item_class_l2
	item_class_l3

# Information used for model

These are the features/data used for models.

colour  
plant\_nr  
plant\_city  
plant\_country\_name  
most\_expensive\_part\_l1  
most\_expensive\_part\_l2  
shipto\_city  
shipto\_nr  
soldto\_city  
soldto\_nr  
car\_nr  
shipto\_country  
soldto\_country

booking\_to\_delivery\_date  
unit\_price\_inK  
type\_of\_sale  
date\_of\_sale  
mode\_of\_transport  
no\_of\_units  
cust\_group\_name  
cust\_segment\_name  
Customer\_Id  
delivery\_month\_from  
delivery\_month\_to  
type  
item\_class\_11  
item\_class\_12  
item class 13

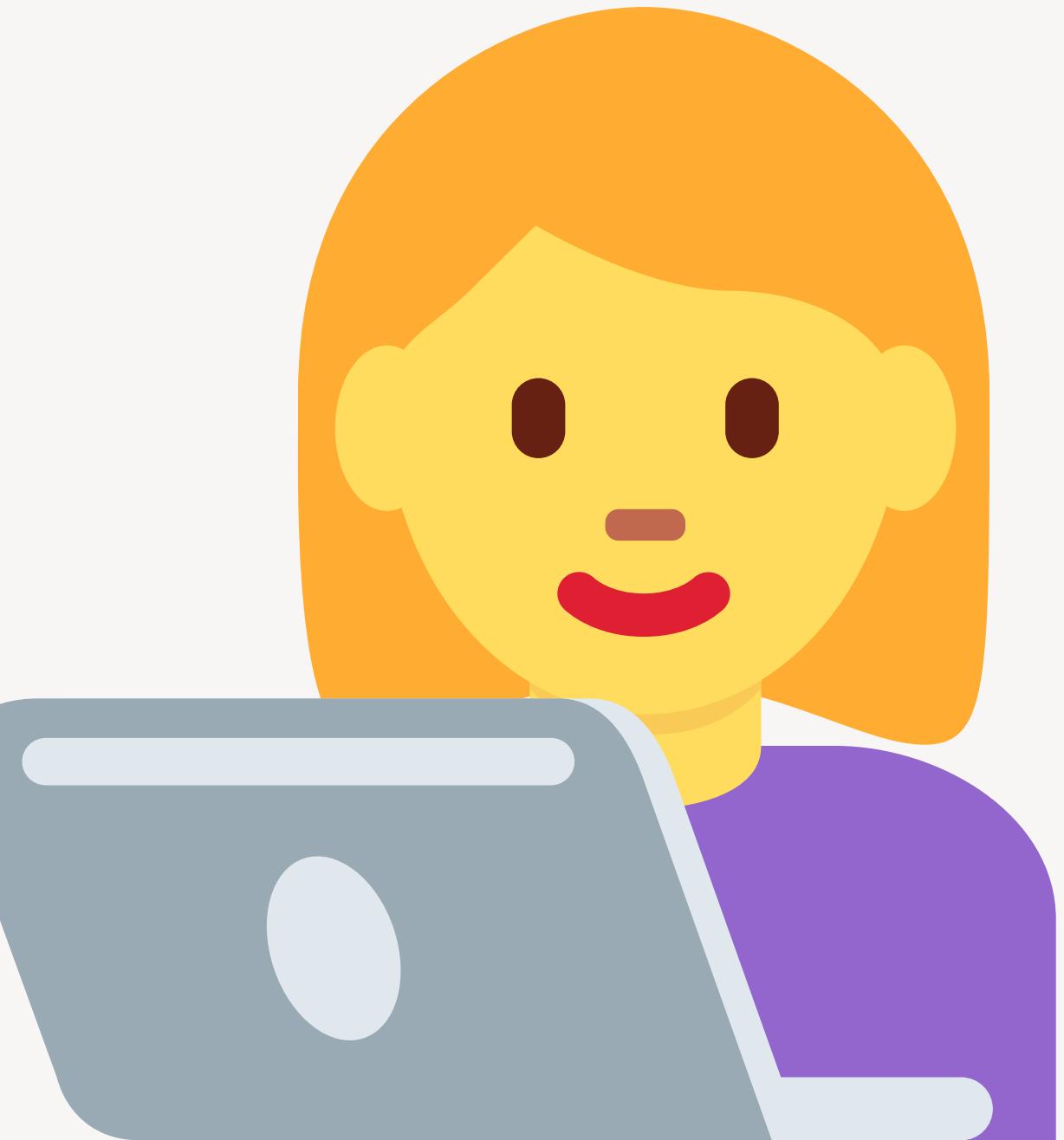




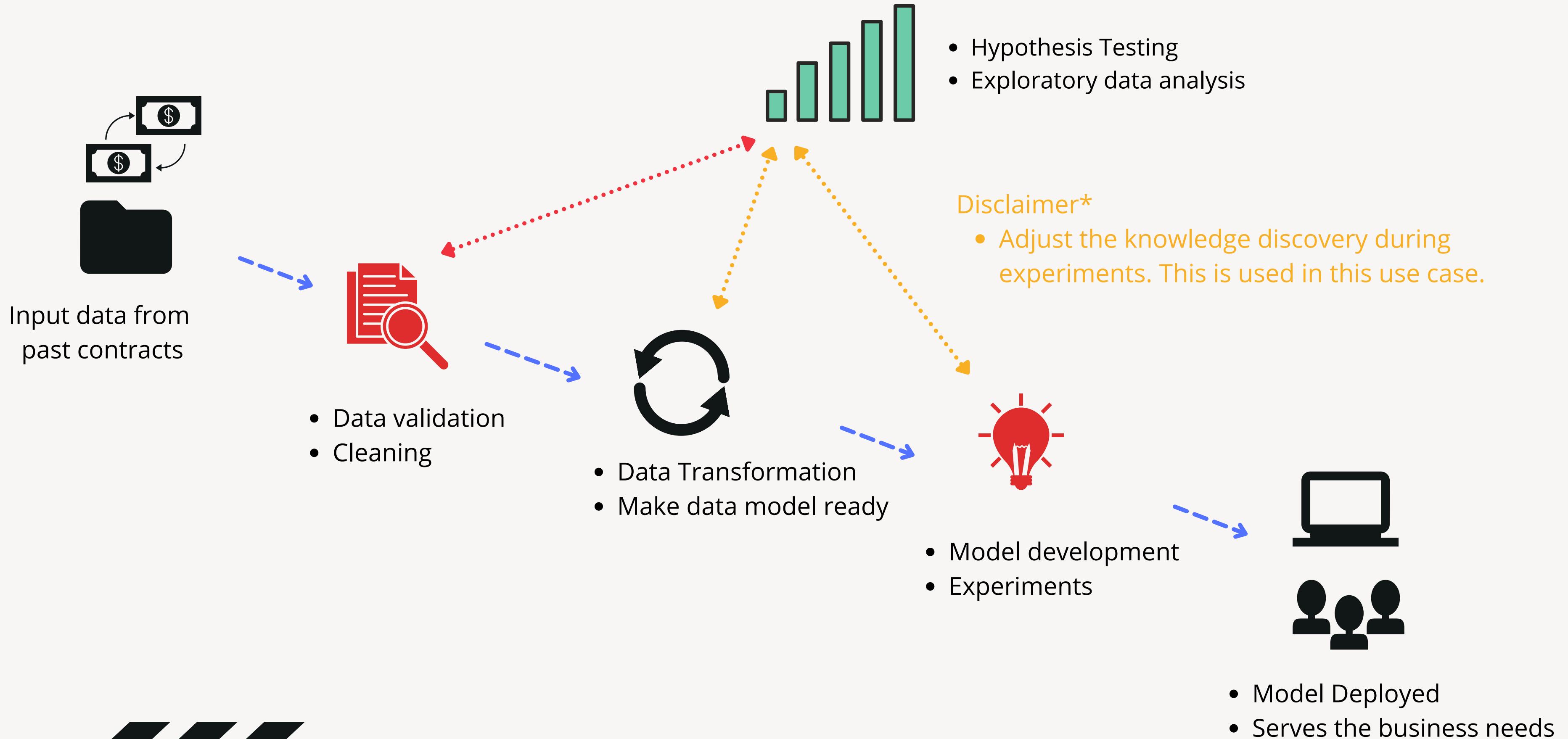
# Demo

\*Disclaimer

This demo depicts how the system works. It needs lot of work to have good look and feel,

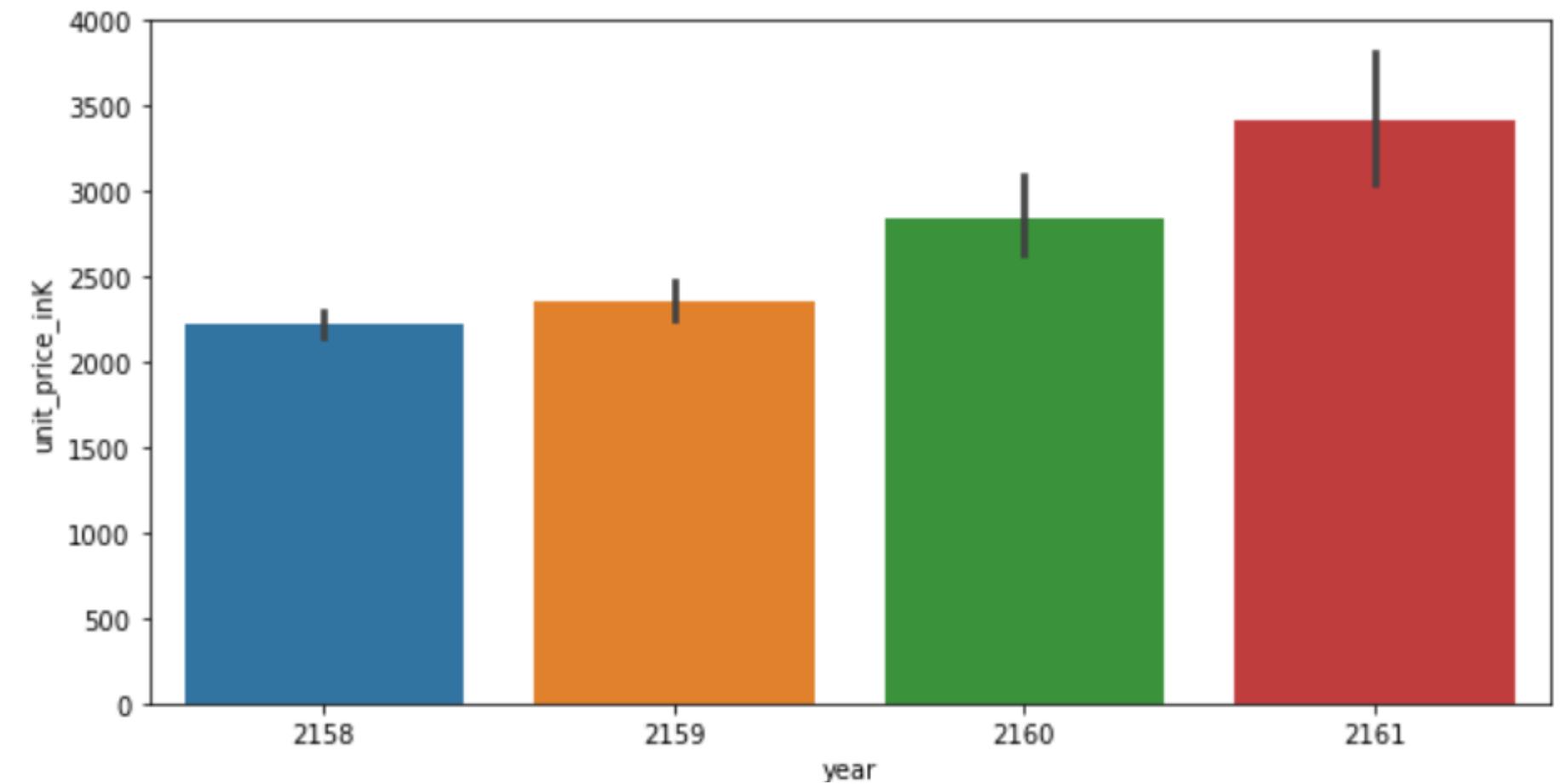
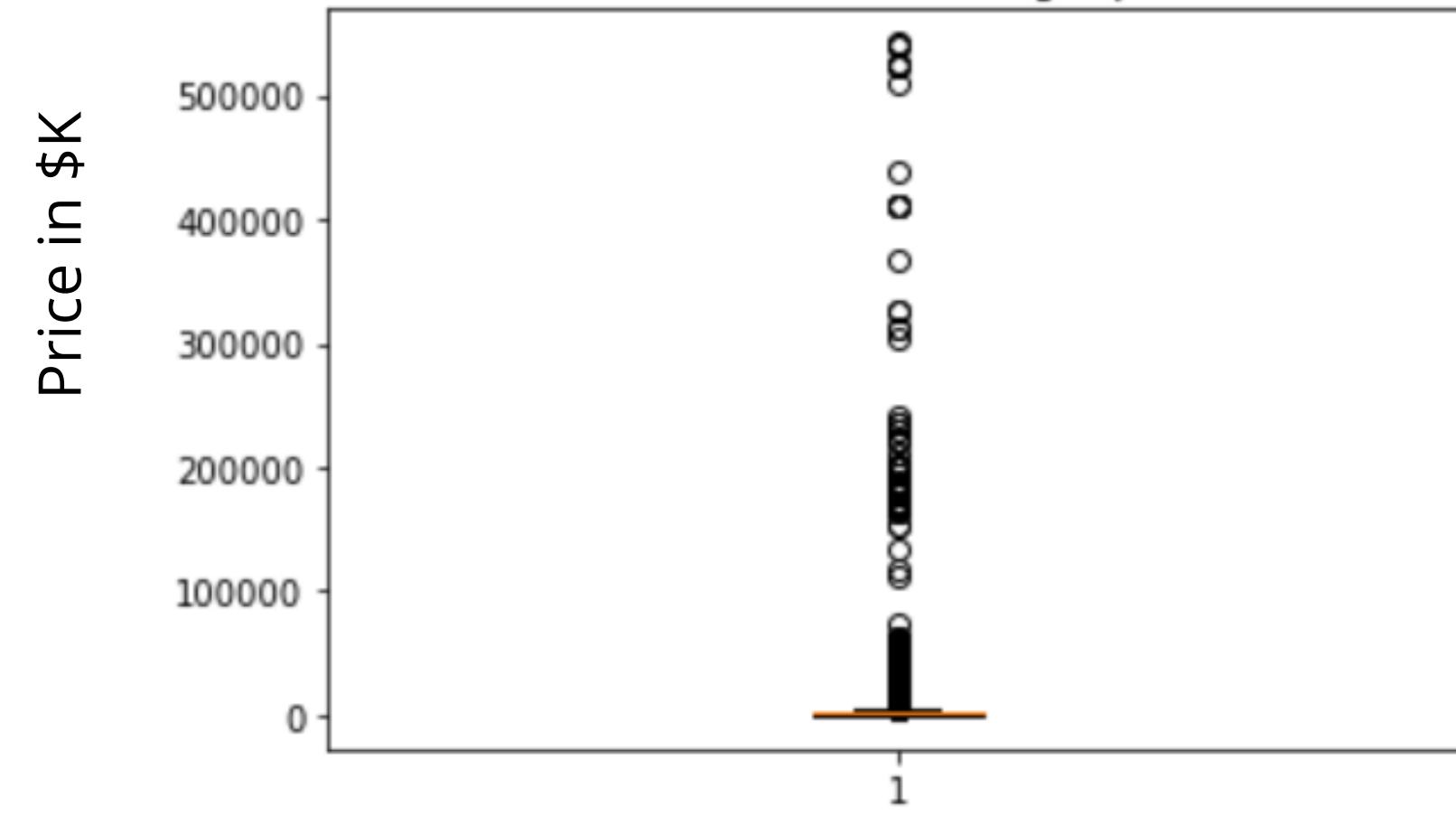


# Model Development Process



# Data Insights 1

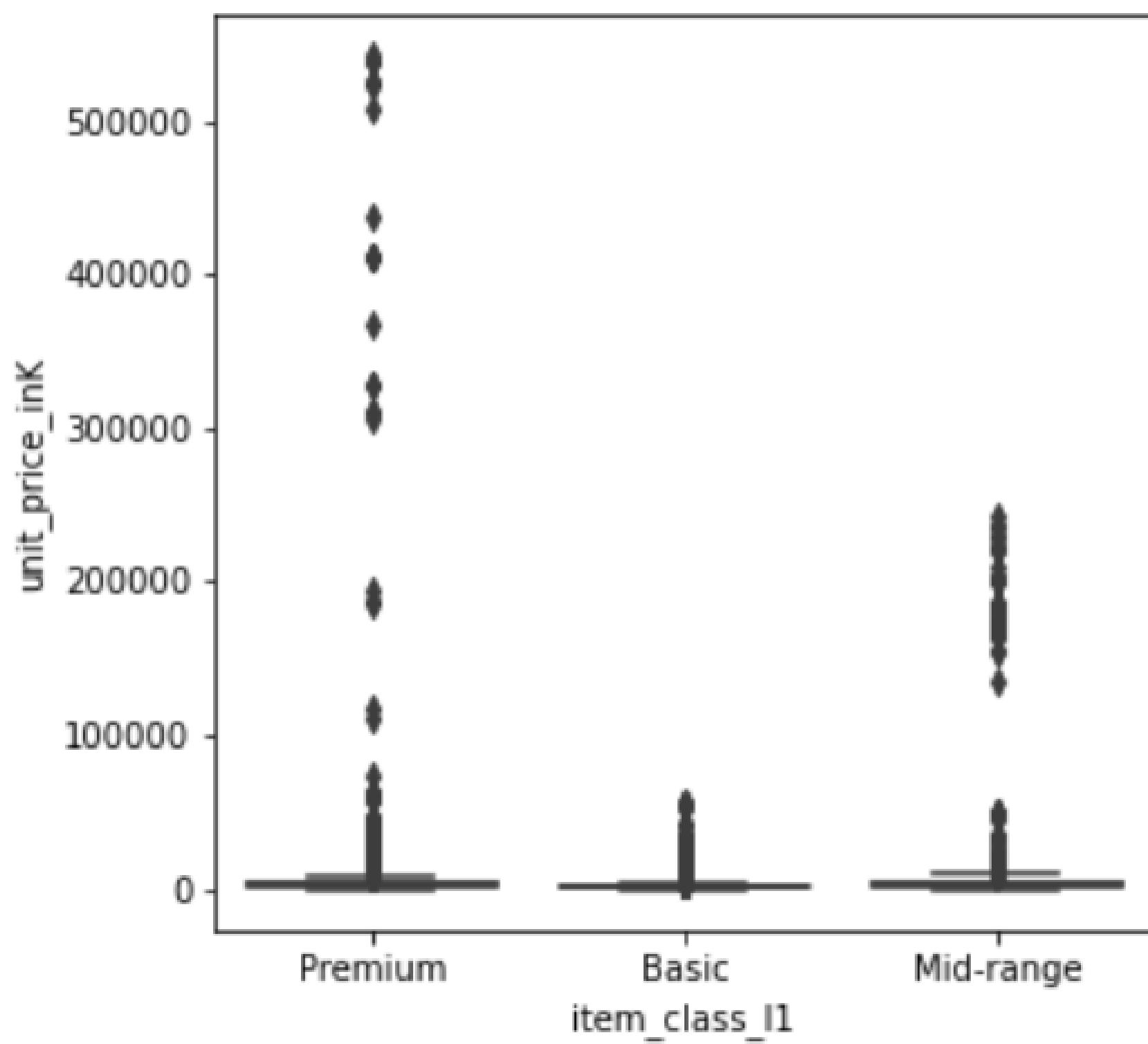
Are there outliers in the target price - Yes



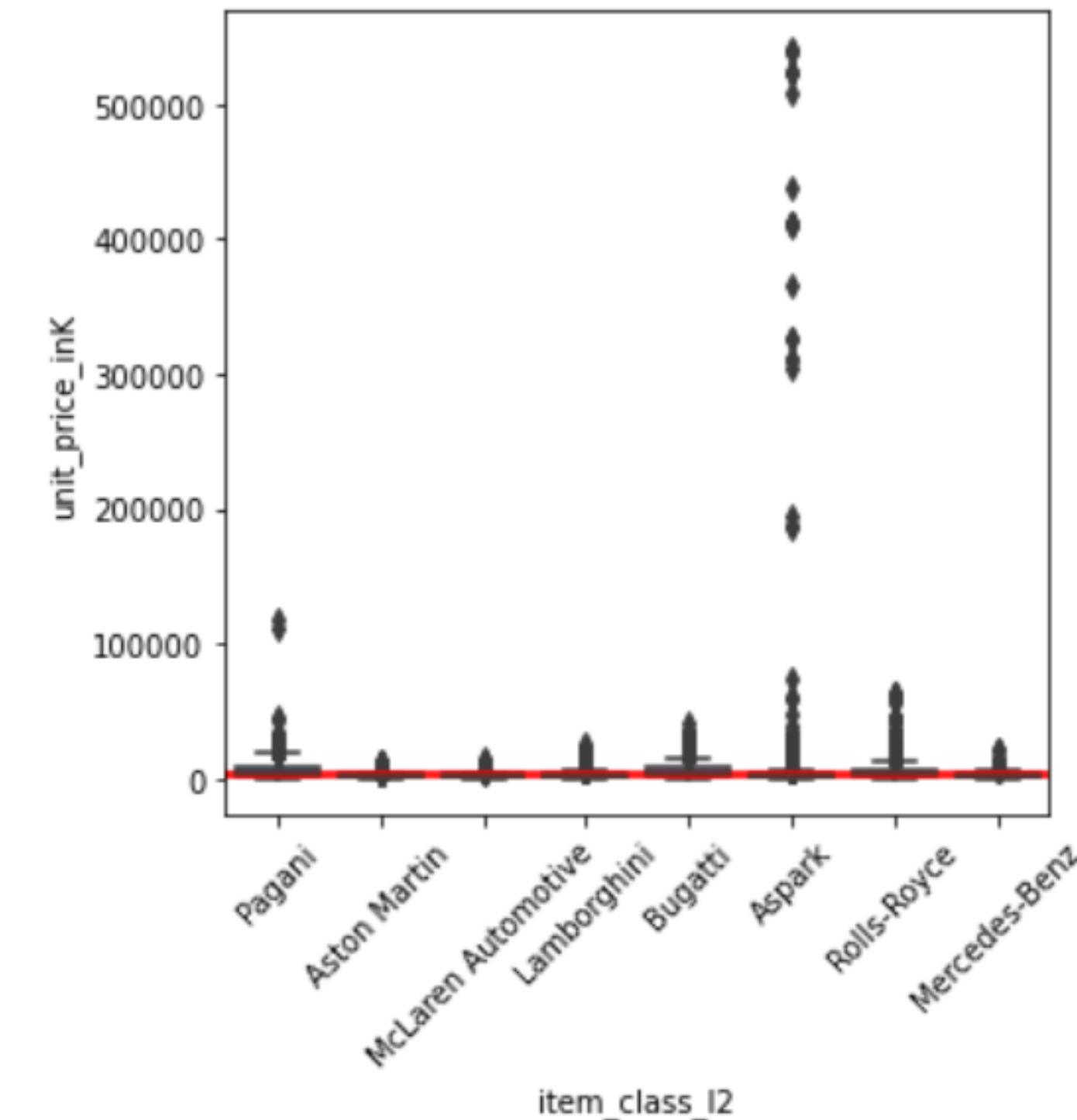
- Outliers in unit\_price could be due to higher-end cars.
- Mean value of unit price is seeing an upside trend



# Data Insights 2



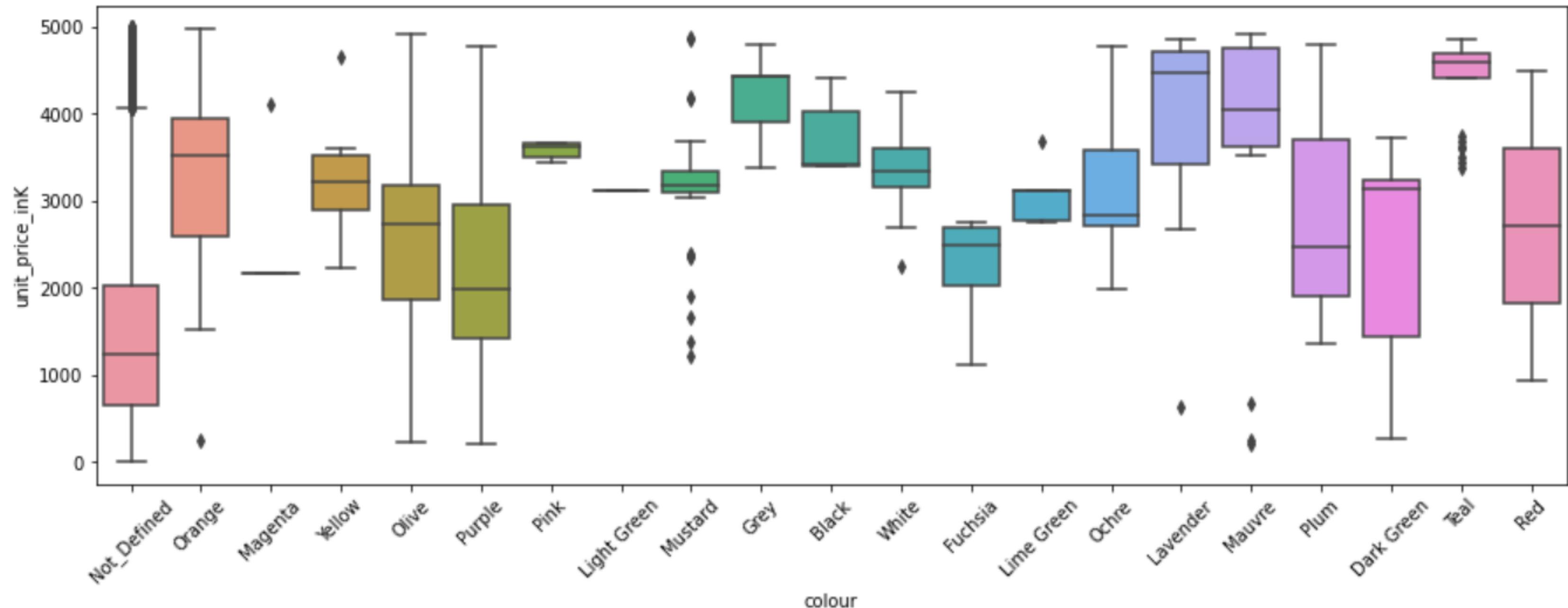
This is the premium range class of brand type Vs price



- Understanding more about the car classes



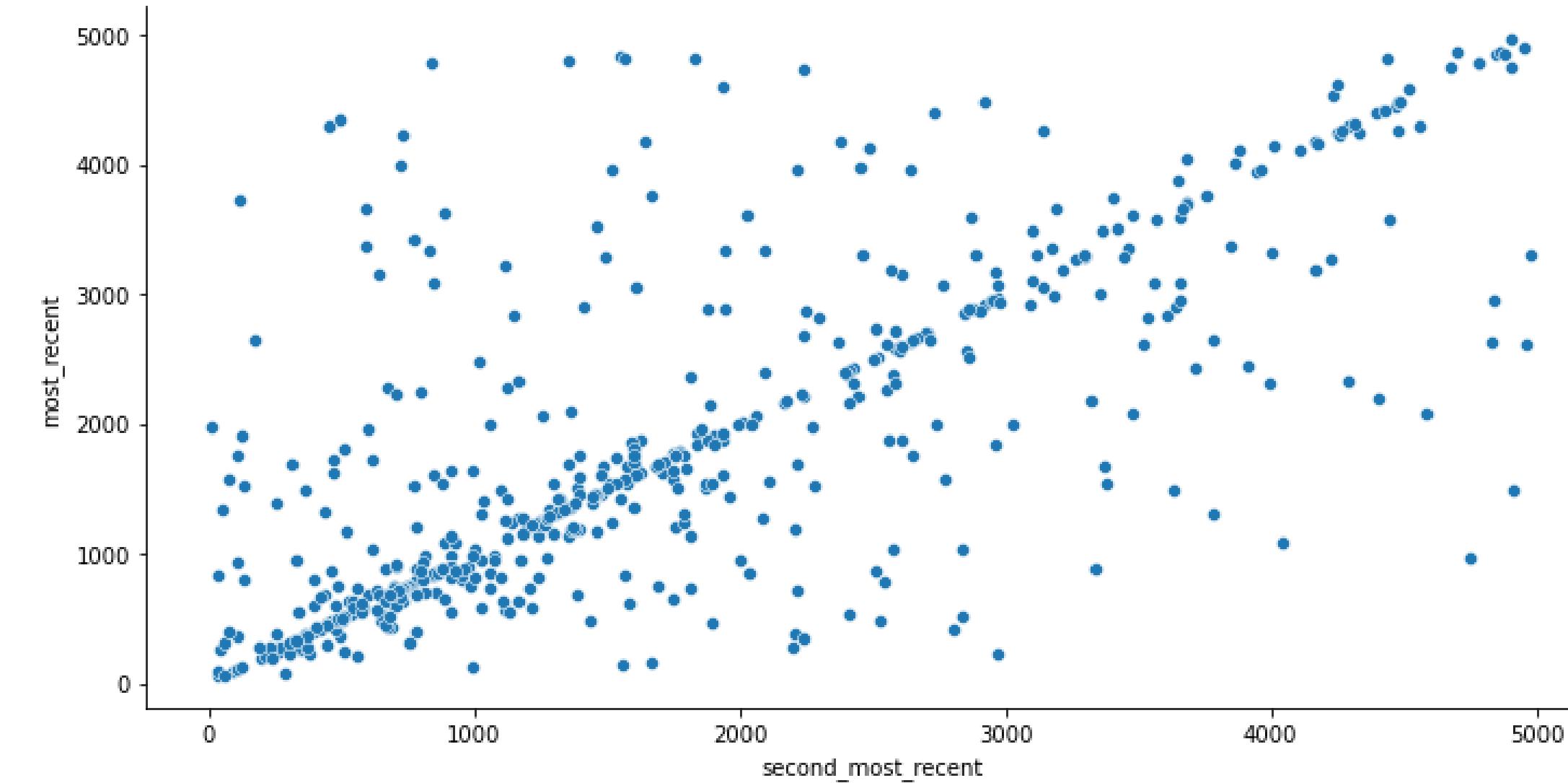
# Data Insights 3



- Filter based on lower car segment
- if color is defined, it tends to get better price



# Data Insights 4



- Filter based on lower car segment
- Recent price has linear corelation with the last paid price

>>>

# Model Used

- Random Forest

## Outcome - Validation Data

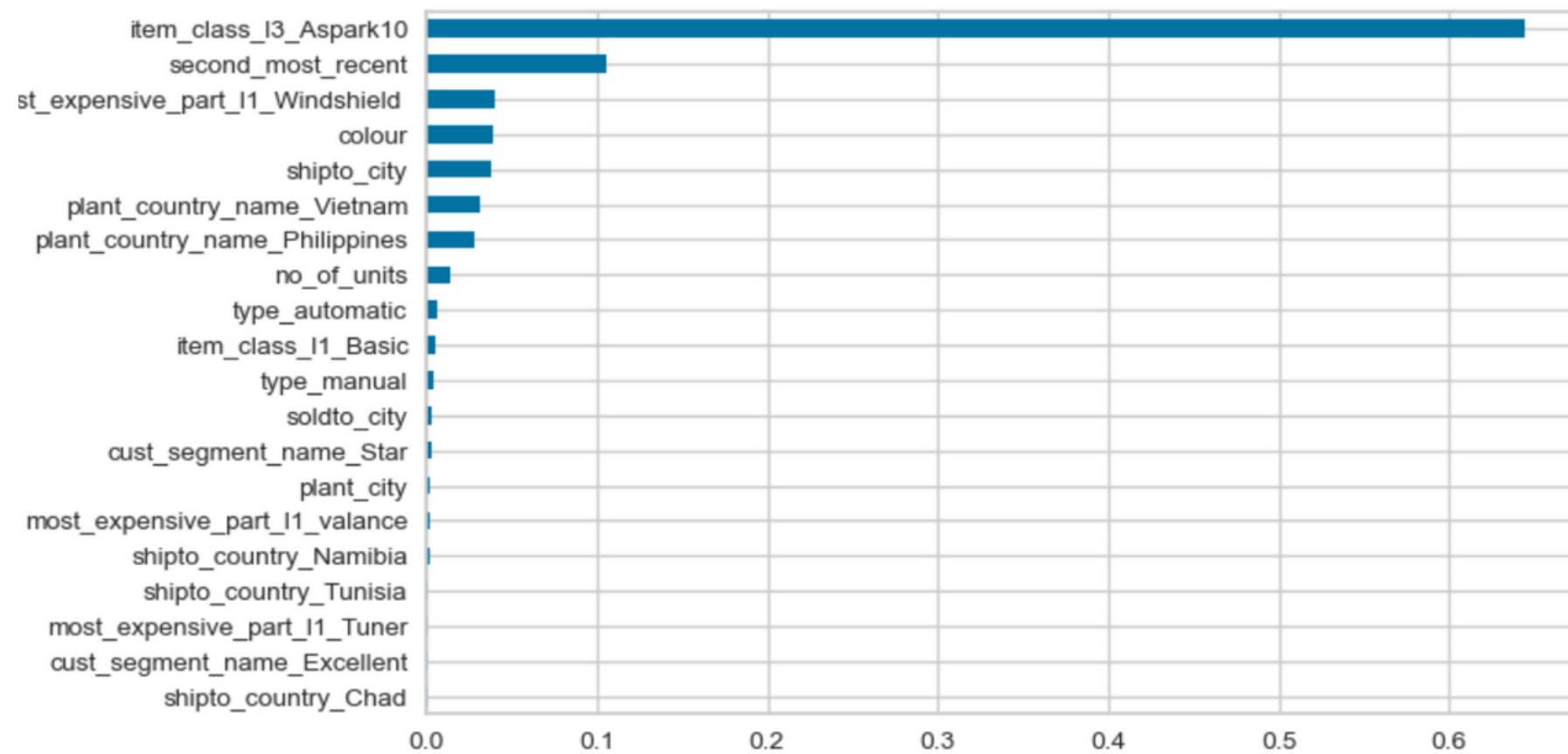
- 95% of the time, the model is able to get the price close to the actual price
- Model price brought more revenue as predicted revenues are higher than actual revenue

## Test Model only on retail data

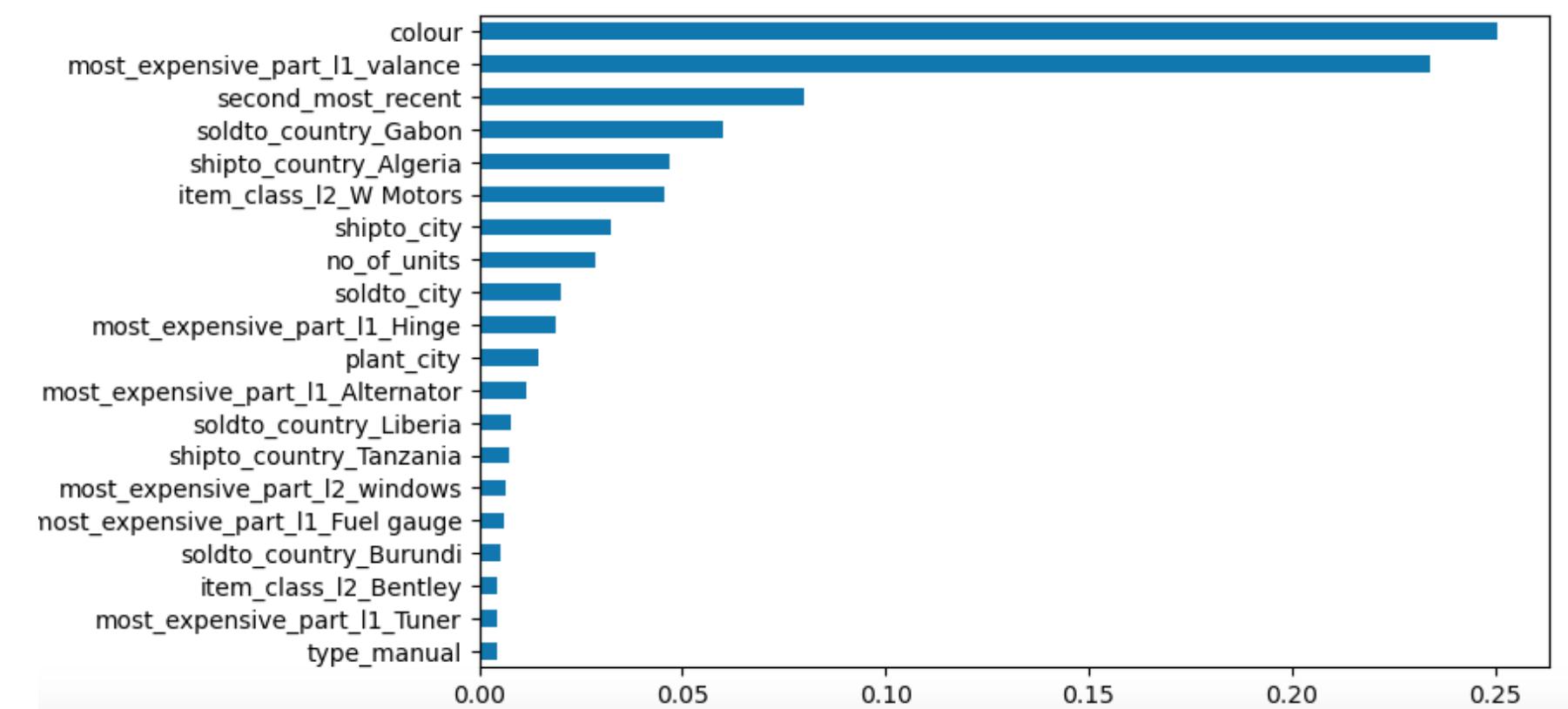
- 65% times model predicts the price closer to the actual price
- Predicted price are in the range of +/- 322 of actual price

Experiment No.	Model	R2_Score	RMSE	Decision
1	Linear Regression	0.14	266454	Model does not fit, we need to have complex models as we have higher dimensions
2	Random Forest (default)	0.72	4867	Model has high RMSE value, removed outlier class features from the model
3	Random Forest (Tuned)	0.95	1445	Current Model – It is overfitting

# Model Decision Interpretation



Model with complete dataset

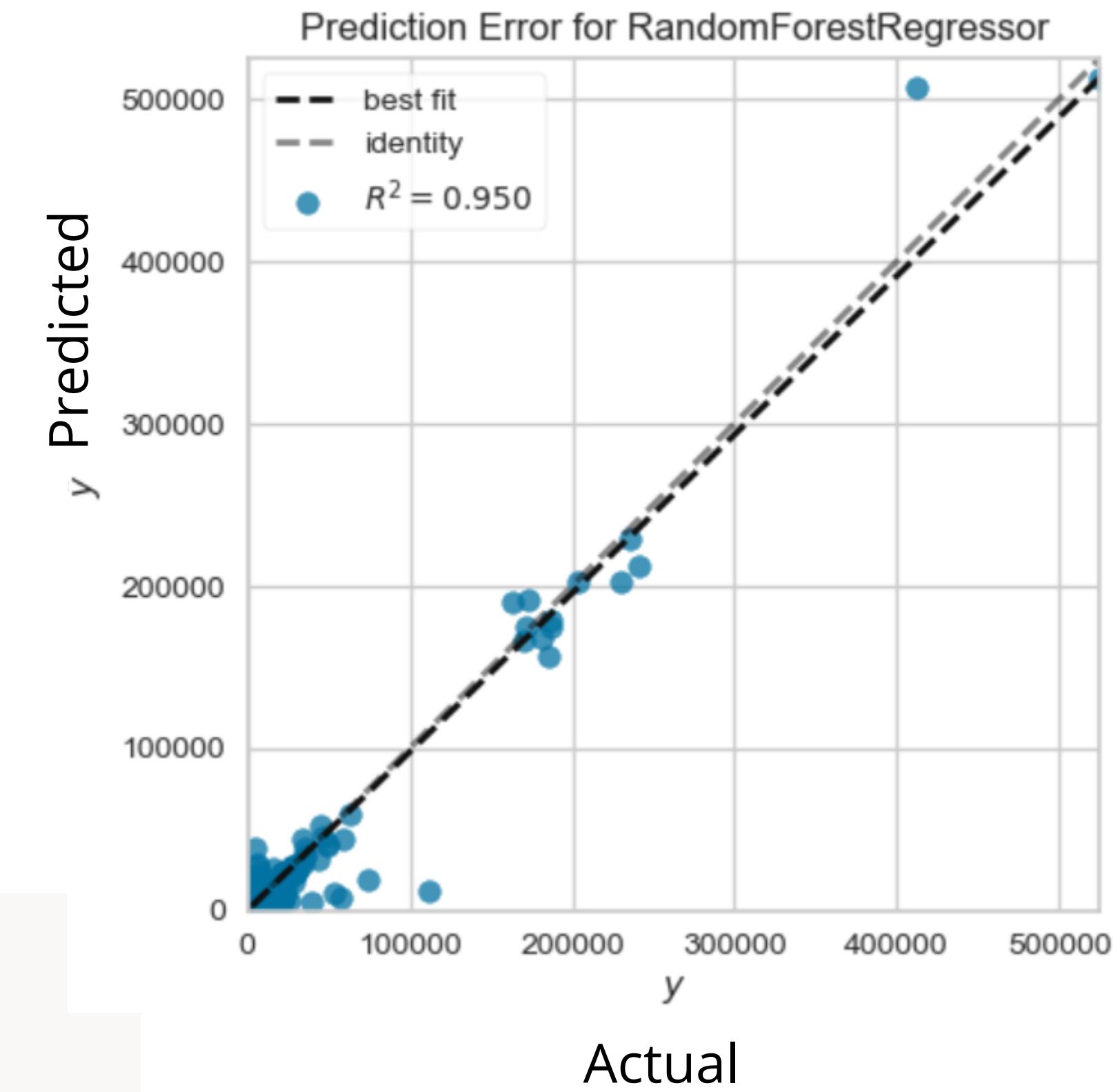
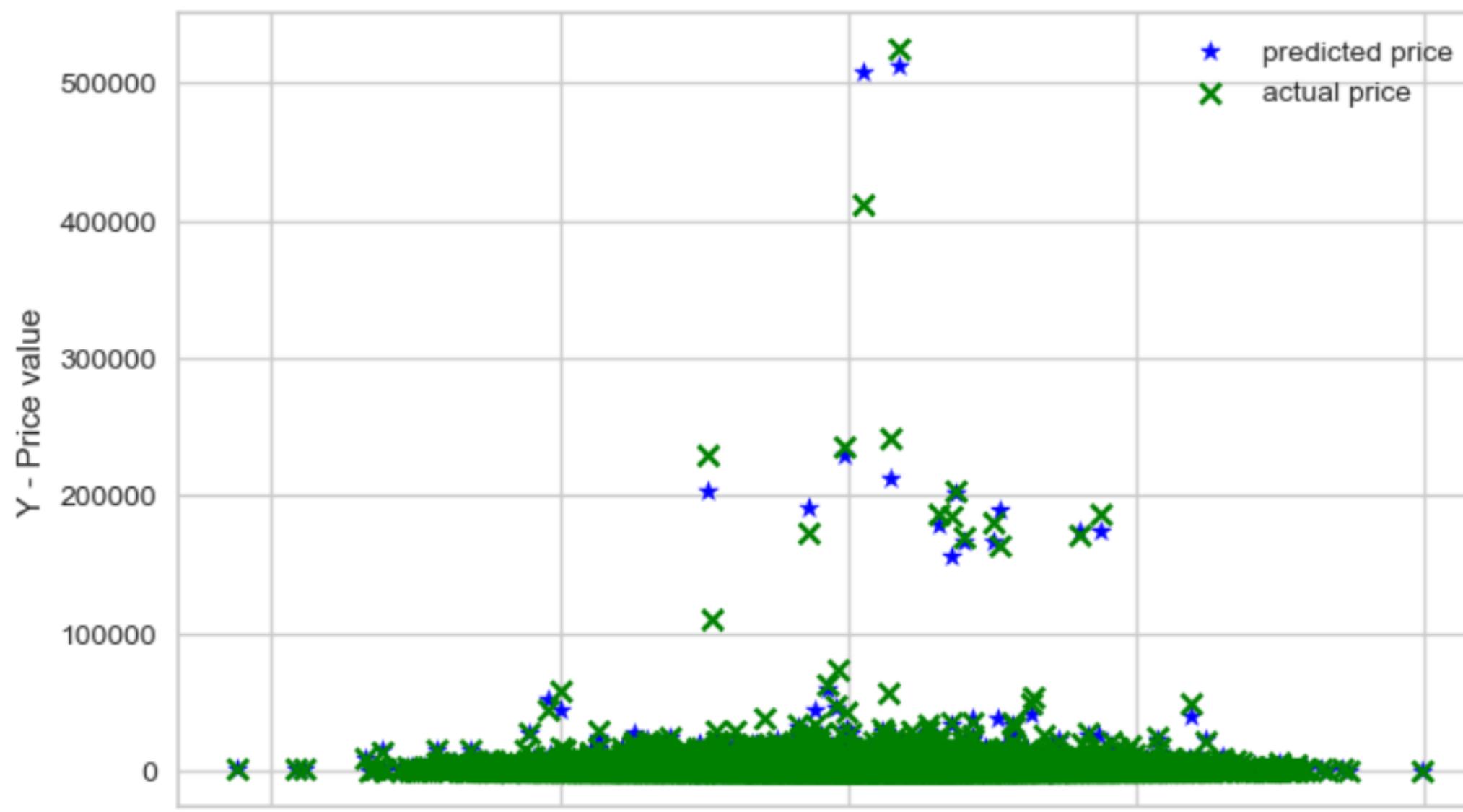


Model with removing outlier of car class

\*Feature weights in taking decisions



# Results Interpretation



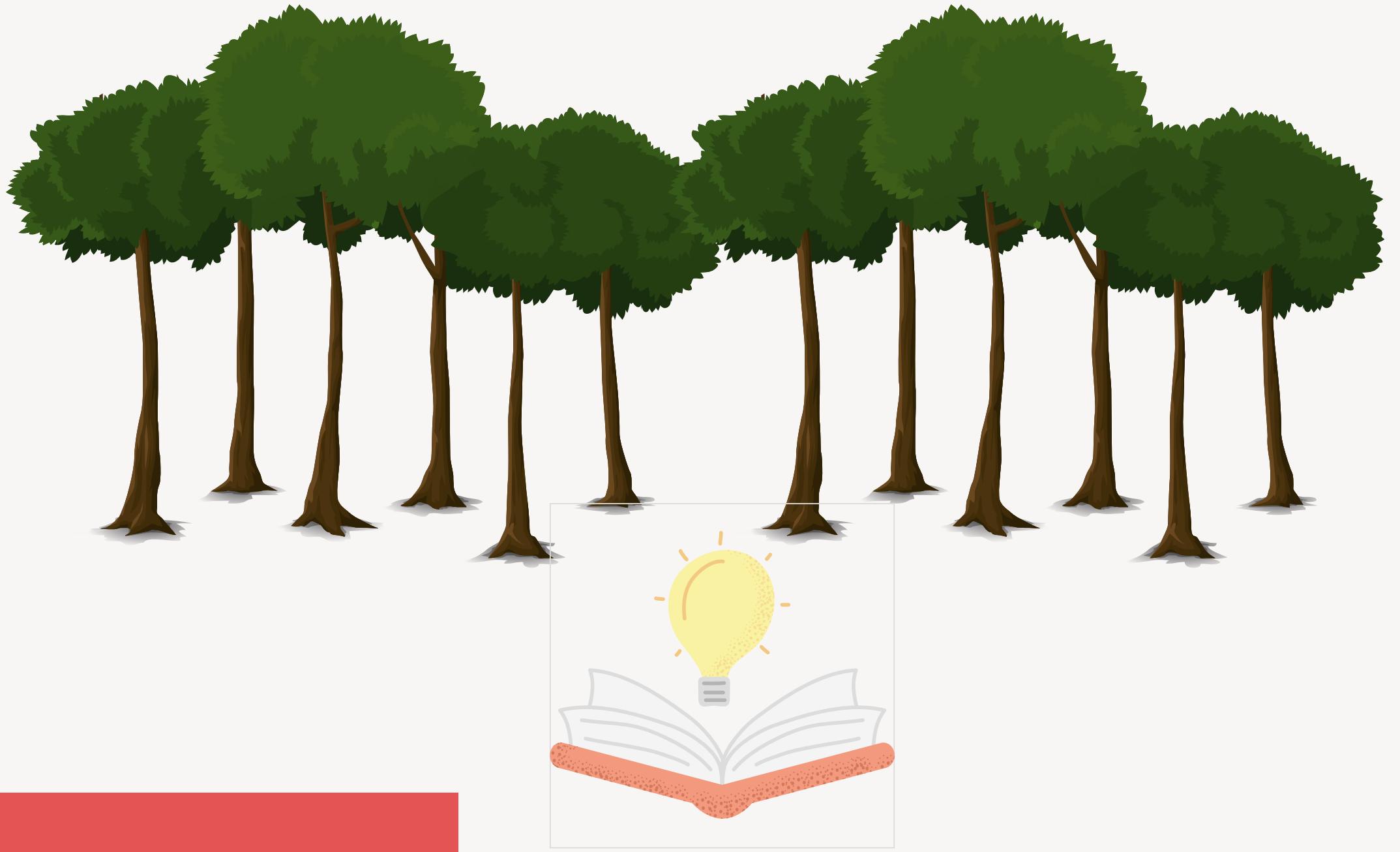


# How Random Forest works?





# How Random Forest works?

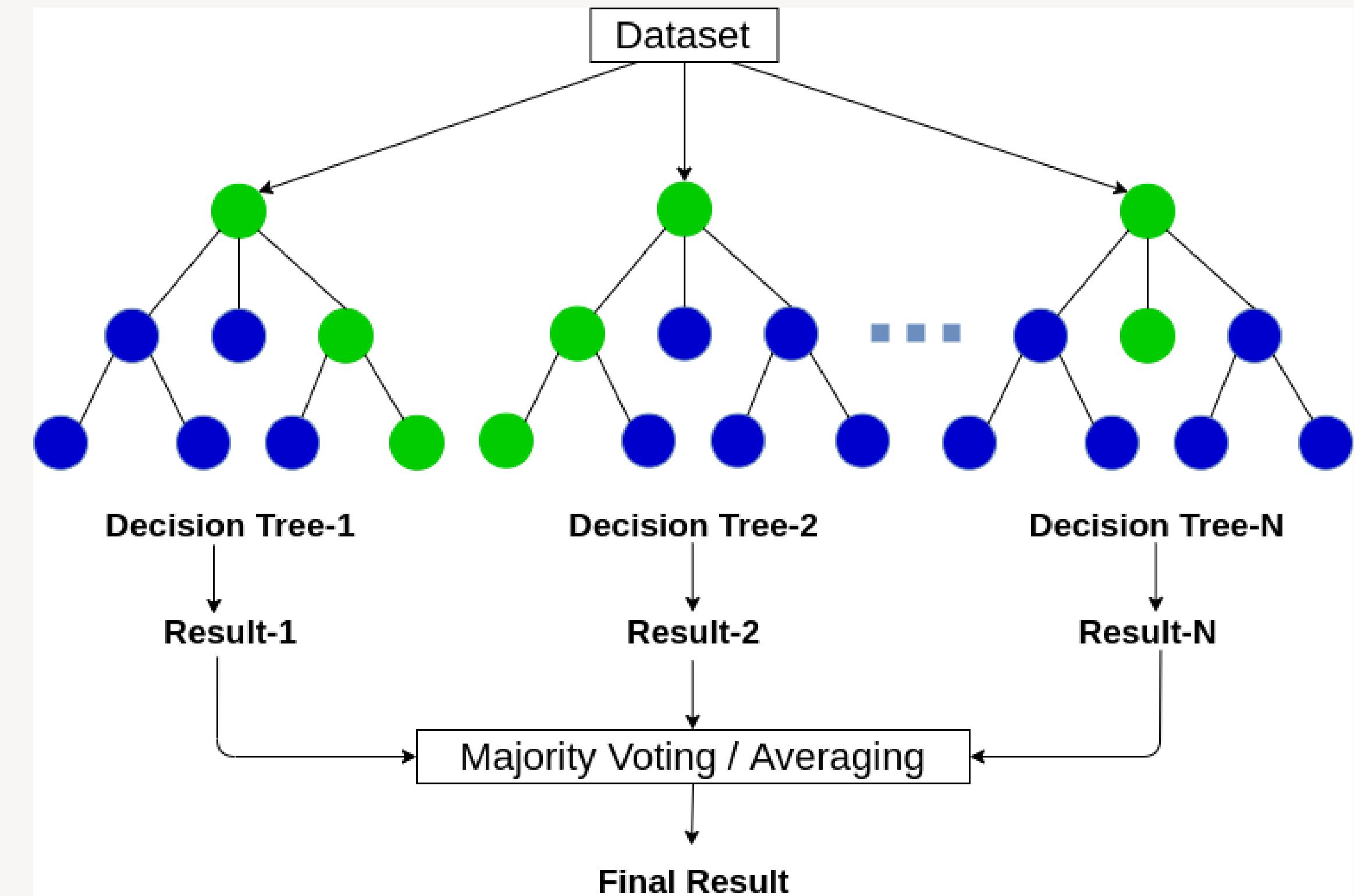


# **Everyone has different :**

- **Experiences**
- **Bias**
- **Choices**
- **Ways to predict**

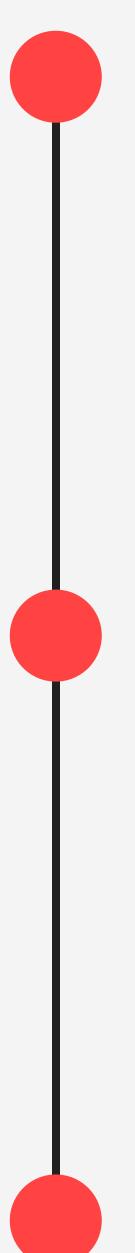
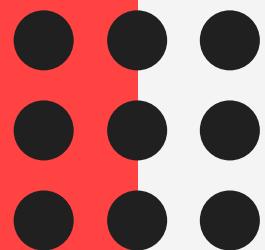


**Each tree is a department  
that provides price as an  
output**



# Next Steps

## To Make it Better



### REAL TIME TESTING

Model shall be deployed and tested in live supervised environment

### EXPERIMENT MORE

Improve the model with better experiments such as incorporating feedbacks from SMEs

### BUILD TRUST WITH USERS

Build more local explanations of the model

# Thank You.

