

# Car MSRP Prediction with Multiple Linear Regression

CIS 4321

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# Introduction

The problem for the manufacturers is to find a balance between what consumers want and pricing the particular vehicle

- Important to Know
  - which features and specs make the Manufacturer Suggested Retail Price
  - which features largely determine the price of the car
  - produce and sell cars that meet market demands for its target consumer group

The purpose of our group project is to figure out what features have an influence on a vehicle's Manufacturer Suggested Retail Price (MSRP) in the Car Features and MSRP dataset

- will be able to
  - know what features are most important to consumers by predicting an appropriate MSRP given a specific set of features
  - understand how those features affect the MSRP the most

# Data Collection and Variable Description

## Collecting Data:

- found our dataset on Kaggle
  - could not find the real source of the set
  - all the records were collected from Twitter and Edmunds

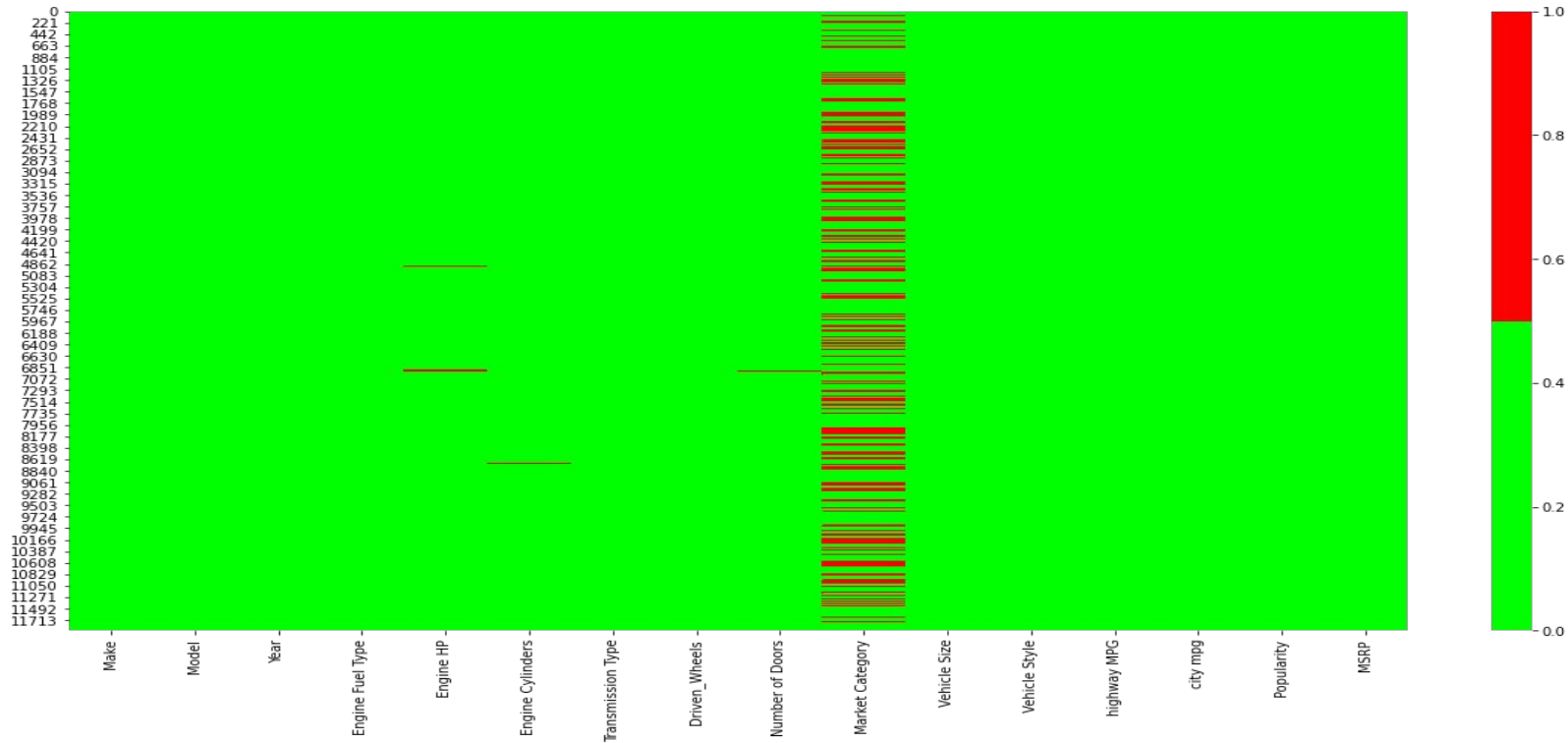
Dataset has 16 variables which are divided to 8 numerical and 8 categorical variables

- numerical variables: Year, Engine HP, Engine Cylinders, Number of Doors, highway mpg, city mpg, Popularity, and MSRP
- categorical variables: Make, Model, Engine Fuel Type, Transmission type, Driven Wheels, Market Category, Vehicle Size, and Vehicle Style
  - categorical variables such as Transmission type, Vehicle Size, and Vehicle Style, are transformed to dummy variables

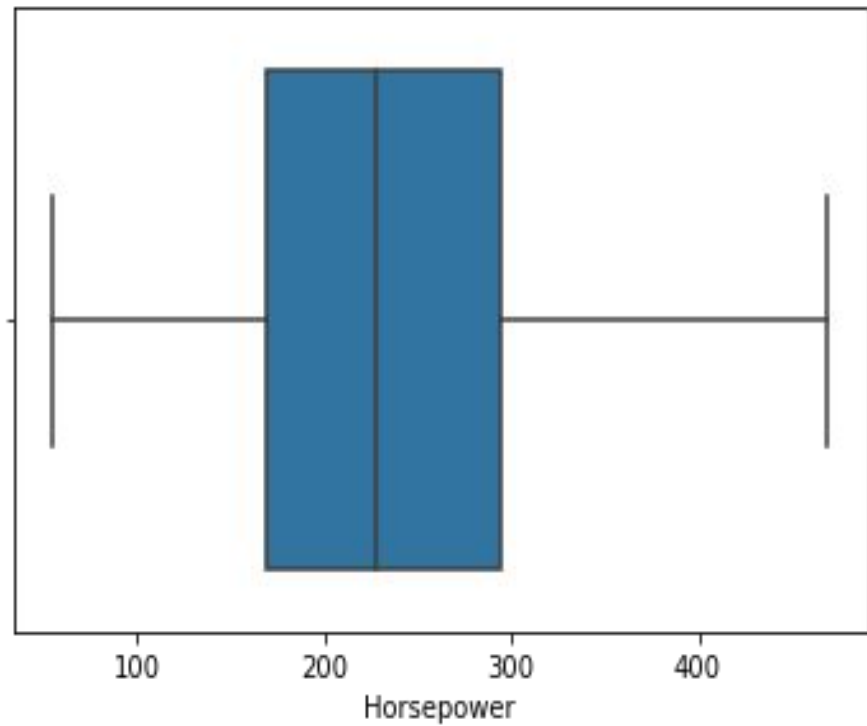
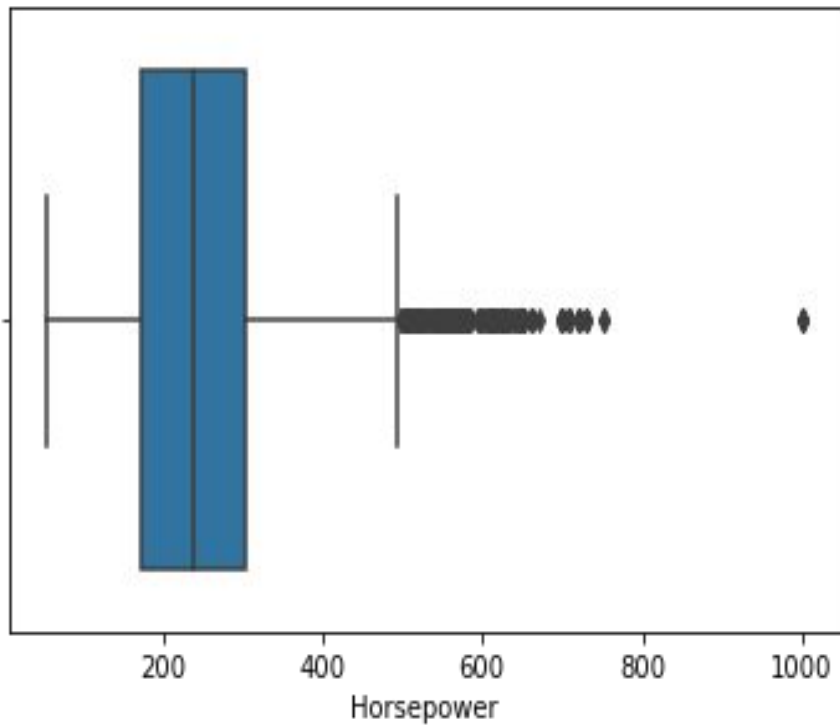
# Data Collection and Variable Description

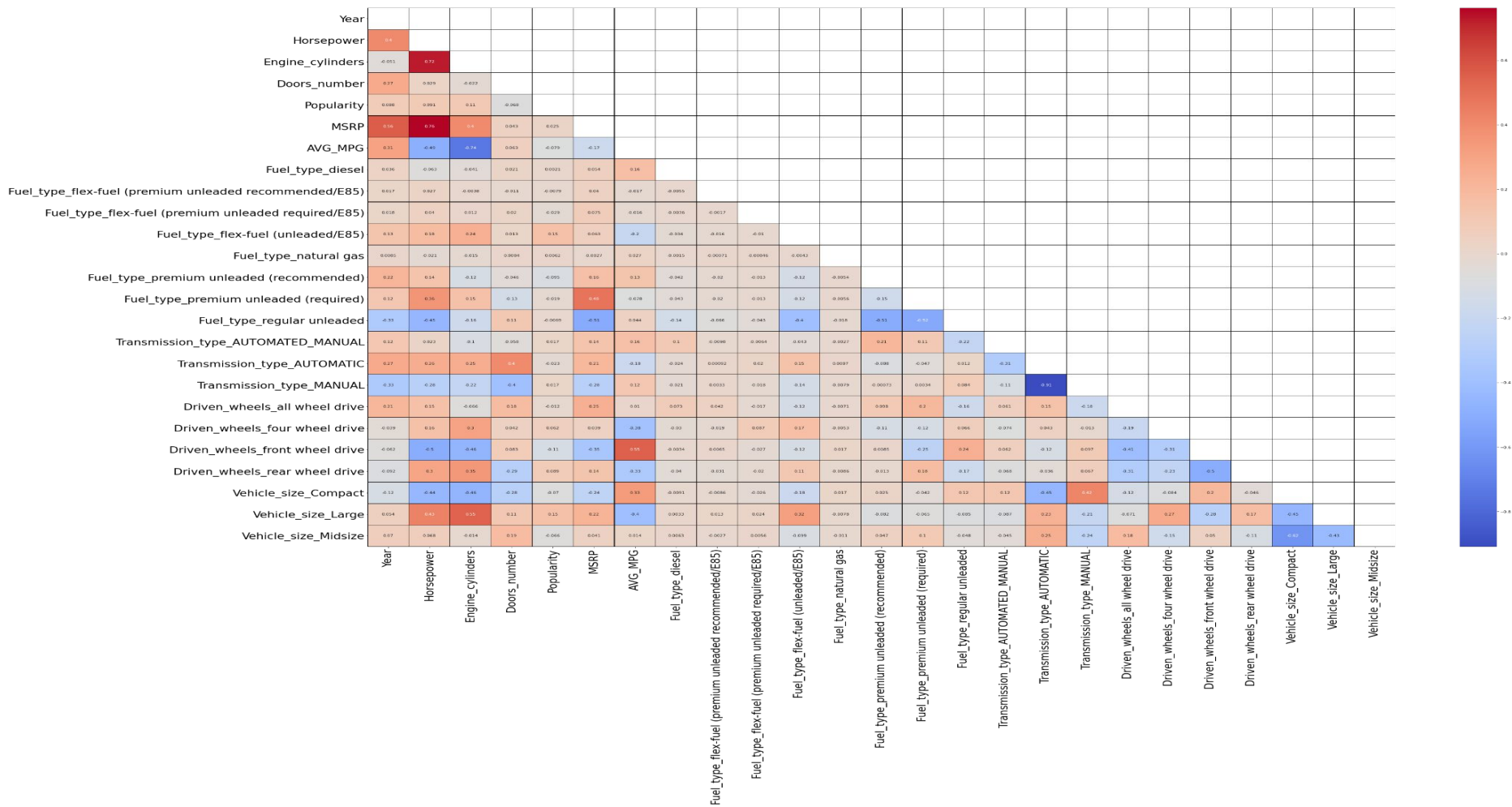
- Make : the brand of the vehicle
- Model : the model of the vehicle
- Year : the year of manufacture of the vehicle
- Engine fuel type : the type of fuel that the vehicle operates on
- Engine HP : the engine horsepower
- Engine cylinders : the number of cylinders in the combustion engine
- Transmission type: what kind of transmission the vehicle has
- Driven\_Wheels : the amount of wheels that receive power directly from the engine
- Number of doors : how many doors the vehicle has
- Market category : the market category to which the vehicle belongs
- Vehicle size : the size of the vehicle based on its volume capacity
- Vehicle style : the style of the vehicle based on its physical shape
- Highway MPG : the number of miles the vehicle drives per gallon at highway speed
- City MPG : the number of miles the vehicle drives per gallon at a city speed
- Popularity : how popular the car is among people
- MSRP : the manufacturer's suggested retail price of a vehicle

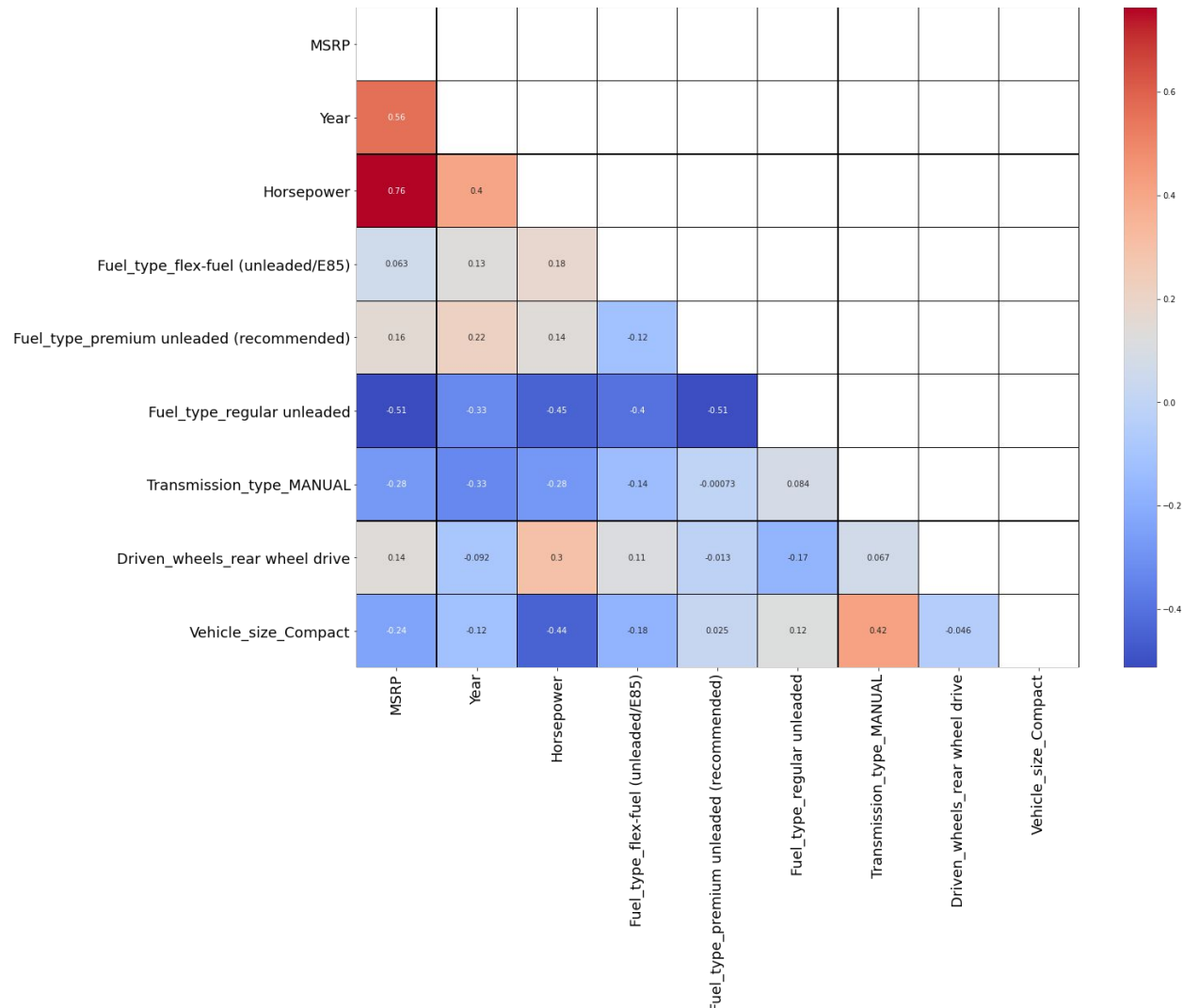
## Descriptive Analysis



## Descriptive Analysis









# Data Analysis

- Use multiple linear regression
  - to predict one variable based on the known information about other variables, and
  - examine how multiple independent variables are related to one dependent variable
- Used recursive model elimination (RFE) to compare models with different sizes and selected Model 8\* with highest adjusted  $R^2$  of 0.75
- Compared the performance of the model using the training and test sets

Model 8\* (Year, Horsepower, Fuel\_type\_flex-fuel (unleaded/E85), Fuel\_type\_premium unleaded (recommended), Fuel\_type\_regular unleaded, Transmission\_type\_MANUAL, Driven\_wheels\_rear wheel drive, Vehicle\_size\_Compact)

## Model 8

Final Features For The Best Model(Model 8)	Coefficient
Year:	745.196675
Horsepower:	150.668859
Fuel_type_flex_fuel(unleaded/E85):	-17650.075979
Fuel_type_premium unleaded (recommended):	-12529.747319
Fuel_type_regular unleaded:	-16154.122128
Transmission_type_MANUAL:	-2799.322581
Driven_wheel_rear wheel drive:	-2139.430713
Vehicle_size_Compact:	3808.463290

# Summary of Findings

- Result of the analysis: predict the MSRP of a car with a 75% accuracy
- We were able to predict a car's MSRP with a given set of features
- We also concluded that the RFE is not the best model to predict a car's MSRP based on the 75% accuracy that the RFE method yielded.
  - We were able to accurately predict a car's msrp after plugging in values for the features listed in Model 8

Car Features	Inputs
Year:	2011
Horsepower:	335
Fuel type:	Premium
Transmission type:	Manual
Driven Wheels:	Rear
Vehicle Size:	Compact
Predicted MSRP:	\$46,549.31

# Implications

- At 75% accuracy in predicting MSRP, manufacturers can:
  - Adjust features and specs according to the preferred MSRP,
  - Budget early on,
  - Decide on how to market the car, and
  - Have a better understanding and control of the factors contributing to the MSRP

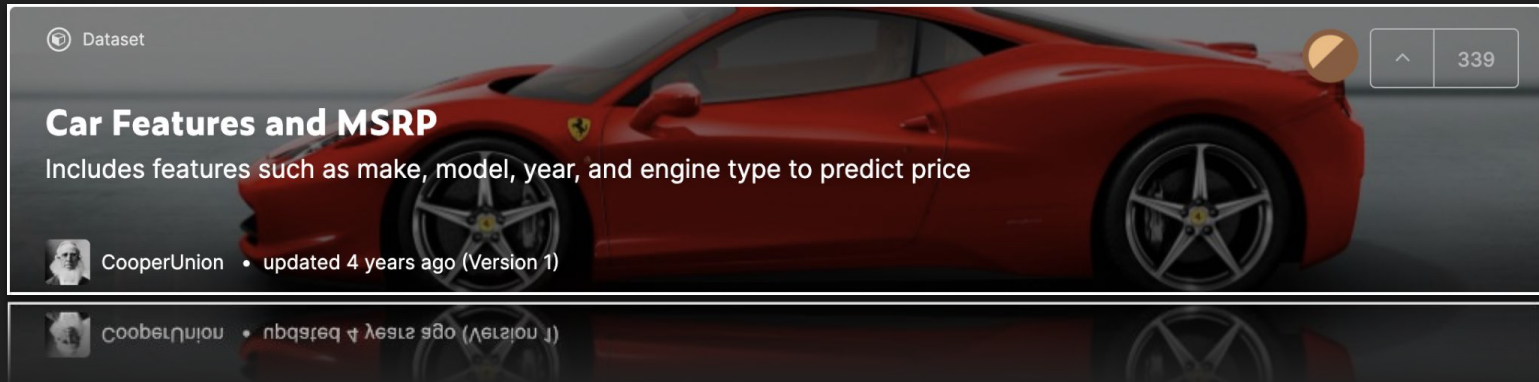


# Limitations

- Lack of numerical variables
- Unknown values which needed to be dropped individually
- Irrelevant variables such as market category, vehicle style, and popularity.

# Future steps

- Improve our results with a dataset that has more complete data and more relevant variables
- Use a different model for our prediction



# Classification

The problem: Filling a missing Engine Cylinder value on a car shopping website.

The approach: Using 3 different Binary Classification methods to solve this problem.

Used Methods: Decision Tree, Random Forest, and Naive Bayes

Results: All Methods yielded impressive Classification abilities of 97% and above accuracy scores.

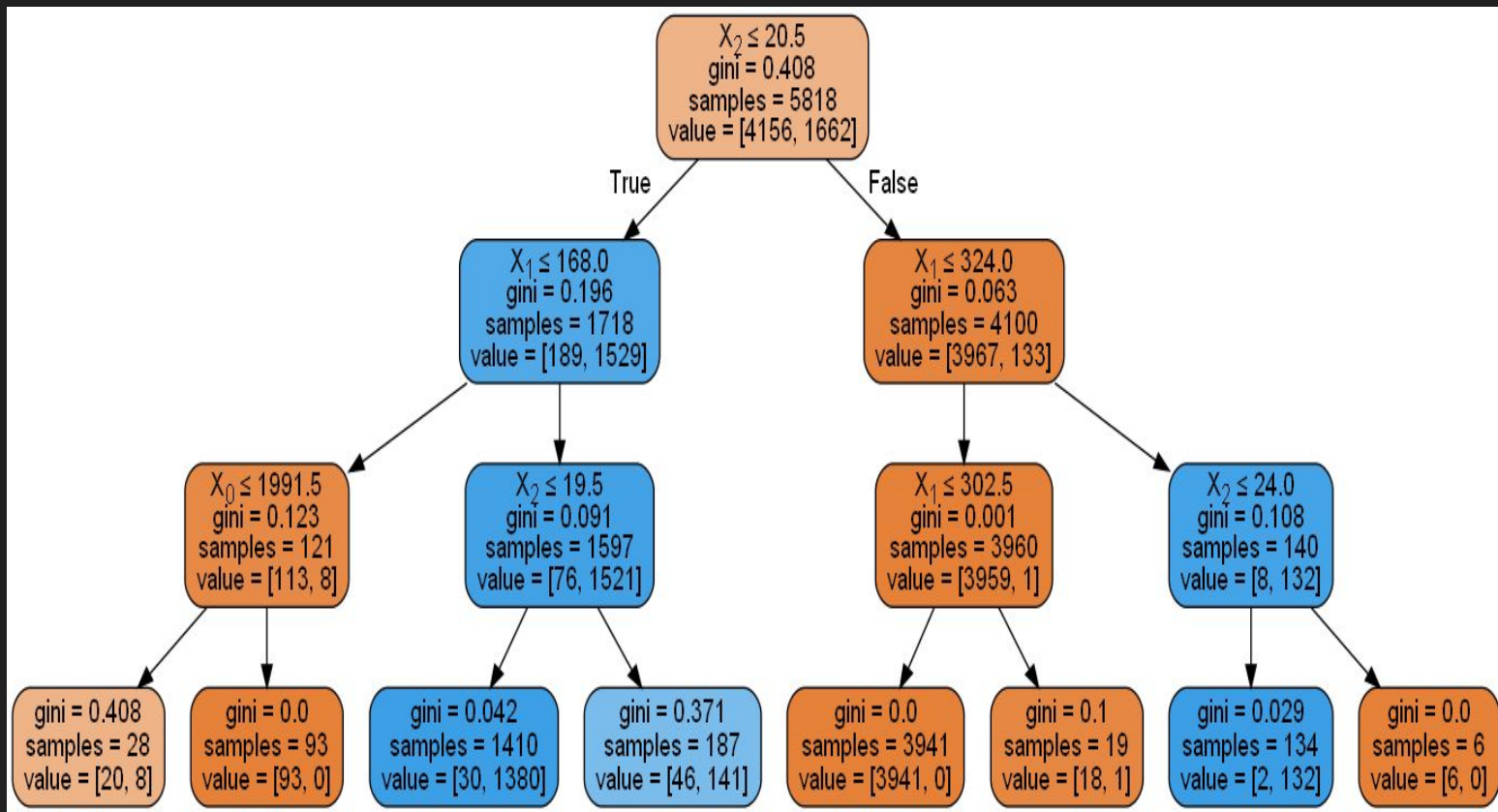
Decision tree for the 'son' variable in the 'son' dataset. The root node splits on  $X_1 \leq 20.5$ . The tree contains 45 nodes, with orange nodes representing splits and blue nodes representing leaf nodes. Each node displays the split condition, gini index, number of samples, and the resulting values for the two branches.

```

graph TD
    Root["X1 ≤ 20.5  
gini = 0.408  
samples = 5818  
value = [4156, 1662]"]
    Root -- True --> Node1["X1 ≤ 168.0  
gini = 0.195  
samples = 1718  
value = [168, 1529]"]
    Root -- False --> Node2["X1 ≤ 324.0  
gini = 0.063  
samples = 4100  
value = [3867, 133]"]
    
    Node1 -- True --> Node3["X1 ≤ 1991.5  
gini = 0.123  
samples = 121  
value = [113, 8]"]
    Node1 -- False --> Node4["X2 ≤ 19.5  
gini = 0.091  
samples = 1597  
value = [78, 1521]"]
    
    Node2 -- True --> Node5["X1 ≤ 502.5  
gini = 0.001  
samples = 3960  
value = [3959, 1]"]
    Node2 -- False --> Node6["X2 ≤ 24.0  
gini = 0.108  
samples = 140  
value = [8, 132]"]
    
    Node3 -- True --> Node7["X1 ≤ 139.0  
gini = 0.408  
samples = 28  
value = [20, 8]"]
    Node3 -- False --> Node8["X1 ≤ 151.5  
gini = 0.486  
samples = 7  
value = [5, 7]"]
    
    Node4 -- True --> Node9["gini = 0.0  
samples = 93  
value = [93, 0]"]
    Node4 -- False --> Node10["X2 ≤ 213.5  
gini = 0.042  
samples = 1410  
value = [50, 1380]"]
    
    Node5 -- True --> Node11["gini = 0.0  
samples = 3941  
value = [3941, 0]"]
    Node5 -- False --> Node12["X1 ≤ 306.5  
gini = 0.1  
samples = 19  
value = [18, 1]"]
    
    Node6 -- True --> Node13["X1 ≤ 37422.5  
gini = 0.029  
samples = 134  
value = [2, 132]"]
    Node6 -- False --> Node14["gini = 0.0  
samples = 6  
value = [6, 0]"]
    
    Node7 -- True --> Node15["X2 ≤ 13.0  
gini = 0.117  
samples = 16  
value = [15, 1]"]
    Node7 -- False --> Node16["gini = 0.0  
samples = 12  
value = [0, 1]"]
    
    Node8 -- True --> Node17["gini = 0.0  
samples = 5  
value = [5, 0]"]
    Node8 -- False --> Node18["gini = 0.0  
samples = 2  
value = [0, 2]"]
    
    Node9 -- True --> Node19["gini = 0.0  
samples = 11  
value = [11, 0]"]
    Node9 -- False --> Node20["X2 ≤ 17.5  
gini = 0.153  
samples = 48  
value = [4, 44]"]
    
    Node10 -- True --> Node21["X1 ≤ 18.5  
gini = 0.149  
samples = 149  
value = [12, 137]"]
    Node10 -- False --> Node22["X1 ≤ 291.5  
gini = 0.005  
samples = 1202  
value = [3, 1199]"]
    
    Node11 -- True --> Node23["X2 ≤ 2011.5  
gini = 0.444  
samples = 69  
value = [46, 23]"]
    Node11 -- False --> Node24["gini = 0.0  
samples = 100  
value = [0, 100]"]
    
    Node12 -- True --> Node25["gini = 0.0  
samples = 18  
value = [0, 18]"]
    Node12 -- False --> Node26["X1 ≤ 274.5  
gini = 0.47  
samples = 37  
value = [14, 23]"]
    
    Node13 -- True --> Node27["gini = 0.0  
samples = 118  
value = [0, 118]"]
    Node13 -- False --> Node28["X1 ≤ 2016.5  
gini = 0.015  
samples = 133  
value = [1, 132]"]
    
    Node14 -- True --> Node29["gini = 0.0  
samples = 1  
value = [1, 0]"]
    Node14 -- False --> Node30["X1 ≤ 376.0  
gini = 0.124  
samples = 15  
value = [1, 14]"]
    
    Node15 -- True --> Node31["gini = 0.0  
samples = 15  
value = [15, 0]"]
    Node15 -- False --> Node32["gini = 0.0  
samples = 1  
value = [0, 1]"]
    
    Node16 -- True --> Node33["gini = 0.0  
samples = 39  
value = [0, 39]"]
    Node16 -- False --> Node34["X1 ≤ 177.5  
gini = 0.494  
samples = 9  
value = [4, 5]"]
    
    Node17 -- True --> Node35["gini = 0.0  
samples = 139  
value = [2, 137]"]
    Node17 -- False --> Node36["X2 ≤ 2015.5  
gini = 0.028  
samples = 10  
value = [10, 0]"]
    
    Node18 -- True --> Node37["gini = 0.0  
samples = 2  
value = [2, 0]"]
    Node18 -- False --> Node38["X1 ≤ 290.5  
gini = 0.033  
samples = 180  
value = [3, 177]"]
    
    Node19 -- True --> Node39["gini = 0.0  
samples = 1022  
value = [0, 1022]"]
    Node19 -- False --> Node40["X1 ≤ 274.5  
gini = 0.47  
samples = 37  
value = [14, 23]"]
    
    Node20 -- True --> Node41["gini = 0.0  
samples = 177  
value = [0, 177]"]
    Node20 -- False --> Node42["gini = 0.0  
samples = 3  
value = [3, 0]"]
    
    Node21 -- True --> Node43["X1 ≤ 3.0  
gini = 0.444  
samples = 21  
value = [14, 7]"]
    Node21 -- False --> Node44["gini = 0.0  
samples = 16  
value = [0, 16]"]
    
    Node22 -- True --> Node45["X2 ≤ 2003.5  
gini = 0.492  
samples = 16  
value = [9, 7]"]
    Node22 -- False --> Node46["X2 ≤ 2008.0  
gini = 0.375  
samples = 12  
value = [5, 9]"]
    
    Node23 -- True --> Node47["gini = 0.0  
samples = 4  
value = [0, 4]"]
    Node23 -- False --> Node48["X1 ≤ 241.5  
gini = 0.48  
samples = 5  
value = [2, 3]"]
    
    Node24 -- True --> Node49["gini = 0.0  
samples = 7  
value = [7, 0]"]
    Node24 -- False --> Node50["gini = 0.0  
samples = 3  
value = [0, 3]"]
    
    Node25 -- True --> Node51["gini = 0.0  
samples = 2  
value = [2, 0]"]
    Node25 -- False --> Node52["gini = 0.0  
samples = 1  
value = [1, 0]"]
    
    Node26 -- True --> Node53["gini = 0.0  
samples = 1  
value = [1, 0]"]
    Node26 -- False --> Node54["gini = 0.0  
samples = 1  
value = [1, 0]"]
    
    Node27 -- True --> Node55["gini = 0.0  
samples = 1  
value = [1, 0]"]
    Node27 -- False --> Node56["gini = 0.0  
samples = 1  
value = [1, 0]"]
    
    Node28 -- True --> Node57["gini = 0.0  
samples = 1  
value = [1, 0]"]
    Node28 -- False --> Node58["gini = 0.0  
samples = 1  
value = [1, 0]"]
    
    Node29 -- True --> Node59["gini = 0.0  
samples = 1  
value = [1, 0]"]
    Node29 -- False --> Node60["gini = 0.0  
samples = 1  
value = [1, 0]"]
    
    Node30 -- True --> Node61["gini = 0.0  
samples = 1  
value = [1, 0]"]
    Node30 -- False --> Node62["gini = 0.0  
samples = 1  
value = [1, 0]"]
    
    Node31 -- True --> Node63["gini = 0.0  
samples = 1  
value = [1, 0]"]
    Node31 -- False --> Node64["gini = 0.0  
samples = 1  
value = [1, 0]"]
    
    Node32 -- True --> Node65["gini = 0.0  
samples = 1  
value = [1, 0]"]
    Node32 -- False --> Node66["gini = 0.0  
samples = 1  
value = [1, 0]"]
    
    Node33 -- True --> Node67["gini = 0.0  
samples = 1  
value = [1, 0]"]
    Node33 -- False --> Node68["gini = 0.0  
samples = 1  
value = [1, 0]"]
    
    Node34 -- True --> Node69["gini = 0.0  
samples = 1  
value = [1, 0]"]
    Node34 -- False --> Node70["gini = 0.0  
samples = 1  
value = [1, 0]"]
    
    Node35 -- True --> Node71["gini = 0.0  
samples = 1  
value = [1, 0]"]
    Node35 -- False --> Node72["gini = 0.0  
samples = 1  
value = [1, 0]"]
    
    Node36 -- True --> Node73["gini = 0.0  
samples = 1  
value = [1, 0]"]
    Node36 -- False --> Node74["gini = 0.0  
samples = 1  
value = [1, 0]"]
    
    Node37 -- True --> Node75["gini = 0.0  
samples = 1  
value = [1, 0]"]
    Node37 -- False --> Node76["gini = 0.0  
samples = 1  
value = [1, 0]"]
    
    Node38 -- True --> Node77["gini = 0.0  
samples = 1  
value = [1, 0]"]
    Node38 -- False --> Node78["gini = 0.0  
samples = 1  
value = [1, 0]"]
    
    Node39 -- True --> Node79["gini = 0.0  
samples = 1  
value = [1, 0]"]
    Node39 -- False --> Node80["gini = 0.0  
samples = 1  
value = [1, 0]"]
    
    Node40 -- True --> Node81["gini = 0.0  
samples = 1  
value = [1, 0]"]
    Node40 -- False --> Node82["gini = 0.0  
samples = 1  
value = [1, 0]"]
    
    Node41 -- True --> Node83["gini = 0.0  
samples = 1  
value = [1, 0]"]
    Node41 -- False --> Node84["gini = 0.0  
samples = 1  
value = [1, 0]"]
    
    Node42 -- True --> Node85["gini = 0.0  
samples = 1  
value = [1, 0]"]
    Node42 -- False --> Node86["gini = 0.0  
samples = 1  
value = [1, 0]"]
    
    Node43 -- True --> Node87["gini = 0.0  
samples = 1  
value =
```

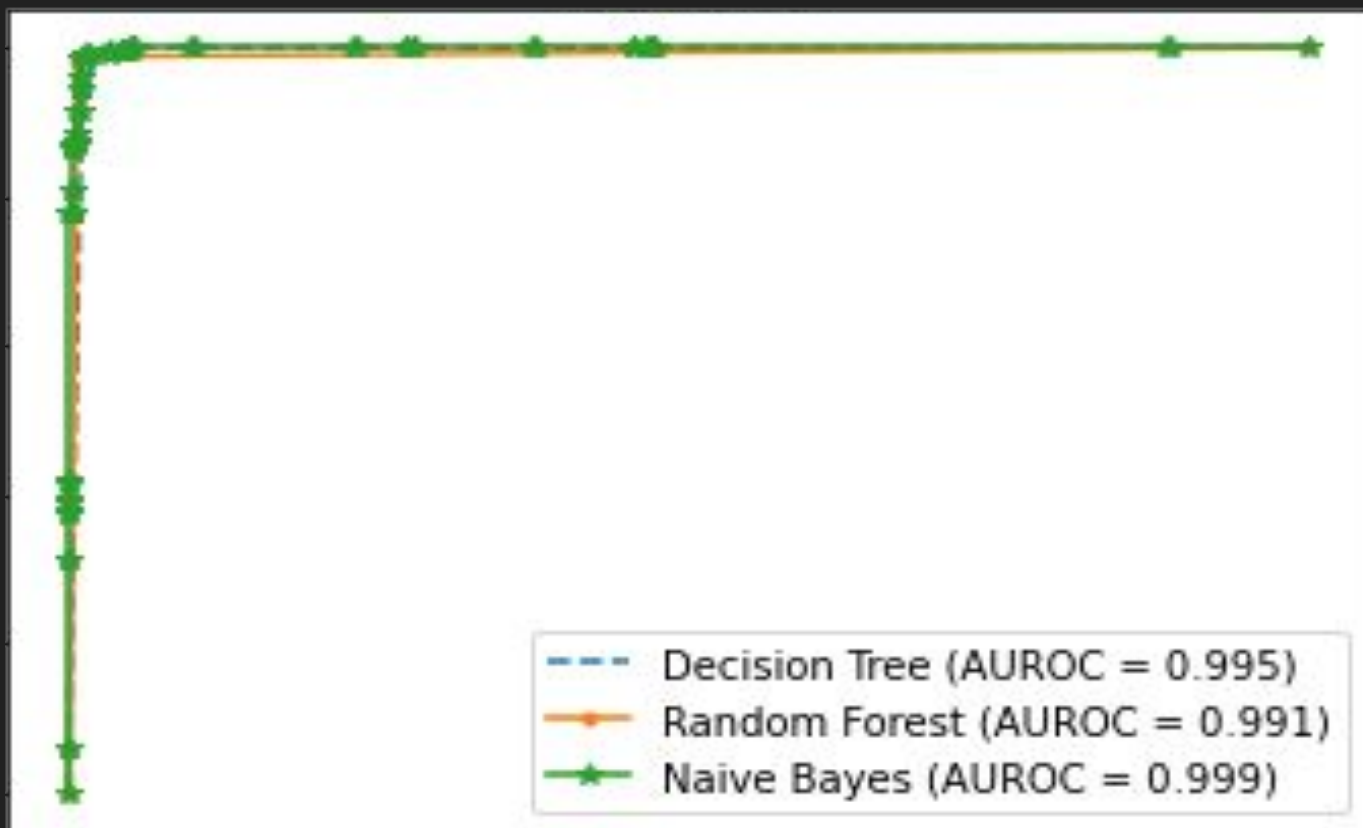
# Full Decision Tree





Pruned Decision Tree

Cross Validation									
Decision Tree					Random Forest				
	Precision	Recall	F1-Score	Support		Precision	Recall	F1-Score	Support
4 Cylinders	99%	99%	99%	4156	4 Cylinders	98%	98%	98%	4156
8 Cylinders	97%	97%	97%	1662	8 Cylinders	96%	94%	95%	1662
Accuracy			98%	5818	Accuracy			97%	5818
Macro Avg	98%	98%	98%	5818	Macro Avg	97%	96%	97%	5818
Weighted Avg	98%	98%	98%	5818	Weighted Avg	97%	97%	97%	5818
Pruned Decision Tree					Naïve Bayes				
	Precision	Recall	F1-Score	Support		Precision	Recall	F1-Score	Support
4 Cylinders	100%	98%	99%	4156	4 Cylinders	99%	99%	99%	4156
8 Cylinders	96%	99%	97%	1662	8 Cylinders	96%	98%	97%	1662
Accuracy			98%	5818	Accuracy			98%	5818
Macro Avg	98%	99%	99%	5818	Macro Avg	98%	98%	98%	5818
Weighted Avg	98%	98%	98%	5818	Weighted Avg	98%	98%	98%	5818



ROC Curve