

### **Data and Business Ideas**

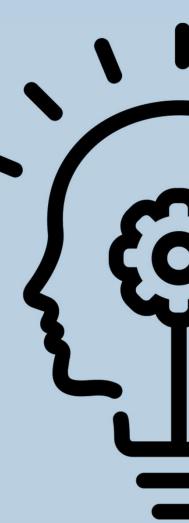
### Data

- Listings of Car Sales from craigslist
- Includes variety of attributes associated with the car and including the price of the car.

### **Business Ideas**

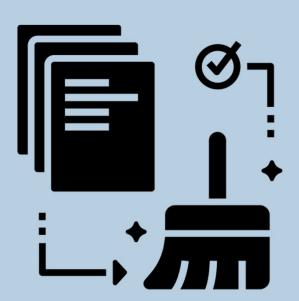
Aim to study these 3 main questions through our modelling:

- What is the degree of influence of different attributes towards the class -- price?
- What recommendations can we provide to future buyers when looking for cars that they may possibly buy?
- What recommendations can we provide to future sellers when making appropriate quotes for the cars they sell?



### **Data Cleaning and Preprocessing**

- Used a combination of Weka and Python Programming
- Initial Data: 25 columns and 423857 rows
- Removed Columns which were unique to every row
- Removed Rows with missing values
- Removed Rows with outlier values for continuous attributes [price and odometer distance]
- Removed Rows for models dated before 2000
- Removed Columns which created noise
- Cleaned Data: 13 columns and 53390 rows
- Discretized Price and Odometer attribute into 3 bins [Frequency and Size of bins varies according to model]



## **Unsupervised Learning: Association Rules**



- Metric Considered: Lift
- Minimum Lift: 1

### Trial 1: Price Discretized to 3 bins of equal size with non-equal frequency {condition\_attribute} -> {price\_attribute}

Price_Attribute	Condition_Attribute	Lift	Confidence	Number of Matched Rules
Mid Price Range	odometer='(-inf-91000.5]'	1.55	0.43	27
Mid Price Range	drive=4wd	1.27	0.35	27
Mid Price Range	condition=excellent	1.18	0.32	27
Mid Price Range	size=full-size	1.14	0.31	27
Mid Price Range	transmission=automatic	1.01	0.28	81

Price_Attribute	Condition_Attribute	Lift	Confidence	Number of Matched Rules
Lowest Price Range	odometer='(141002.5-inf)'	1.30	0.84	265
Lowest Price Range	drive=fwd	1.21	0.78	633
Lowest Price Range	type=sedan	1.20	0.77	303
Lowest Price Range	condition=good	1.19	0.77	197
Lowest Price Range	size=compact	1.19	0.76	5
Lowest Price Range	paint_color=silver	1.11	0.71	19
Lowest Price Range	cylinders=4_cylinders	1.10	0.71	383
Lowest Price Range	odometer='(91000.5-141002.5]'	1.08	0.69	153
Lowest Price Range	size=mid-size	1.08	0.70	163
Lowest Price Range	cylinders=6_cylinders	1.04	0.67	217
Lowest Price Range	fuel=gas	1.02	0.66	1169
Lowest Price Range	title_status=clean	1.01	0.65	1157

# Unsupervised Learning: Association Rules

# Trial 2: Price Discritized to 3 bins of non-equal size with equal frequency {condition\_attribute} -> {price\_attribute}

Price_Attribute	Condition_Attribute	Lift	Confidence	Number of Matched Rules
Mid Price Range	odometer='(91000.5-141002.5]'	1.30	0.43	27
Mid Price Range	cylinders=4_cylinders	1.15	0.38	41
Mid Price Range	drive=fwd	1.12	0.37	43
Mid Price Range	type=sedan	1.11	0.37	19
Mid Price Range	condition=excellent	1.10	0.36	29
Mid Price Range	type=SUV	1.04	0.34	1
Mid Price Range	size=mid-size	1.04	0.35	19
Mid Price Range	transmission=automatic	1.01	0.33	153
Mid Price Range	fuel=gas	1.01	0.34	153

Price_Attribute	Condition_Attribute	Lift	Confidence	Number of Matched Rules
Lowest Price Range	odometer='(141002.5-inf)'	1.63	0.54	27
Lowest Price Range	condition=good	1.38	0.46	27
Lowest Price Range	drive=fwd	1.28	0.43	75
Lowest Price Range	type=sedan	1.27	0.42	57
Lowest Price Range	size=mid-size	1.11	0.37	27
Lowest Price Range	cylinders=6_cylinders	1.11	0.37	27
Lowest Price Range	cylinders=4_cylinders	1.05	0.35	45
Lowest Price Range	fuel=gas	1.03	0.35	203
Lowest Price Range	title_status=clean	1.02	0.34	195

Price_Attribute	Condition_Attribute	Lift	Confidence	Number of Matched Rules
Max Price Range	odometer='(-inf-91000.5]'	1,73	0.58	49
Max Price Range	cylinders=8_cylinders	1.48	0.49	5
Max Price Range	drive=4wd	1.38	0.46	59
Max Price Range	size=full-size	1.22	0.41	75
Max Price Range	condition=excellent	1.18	0.39	69
Max Price Range	type=SUV	1.09	0.36	9
Max Price Range	transmission=automatic	1.01	0.34	135

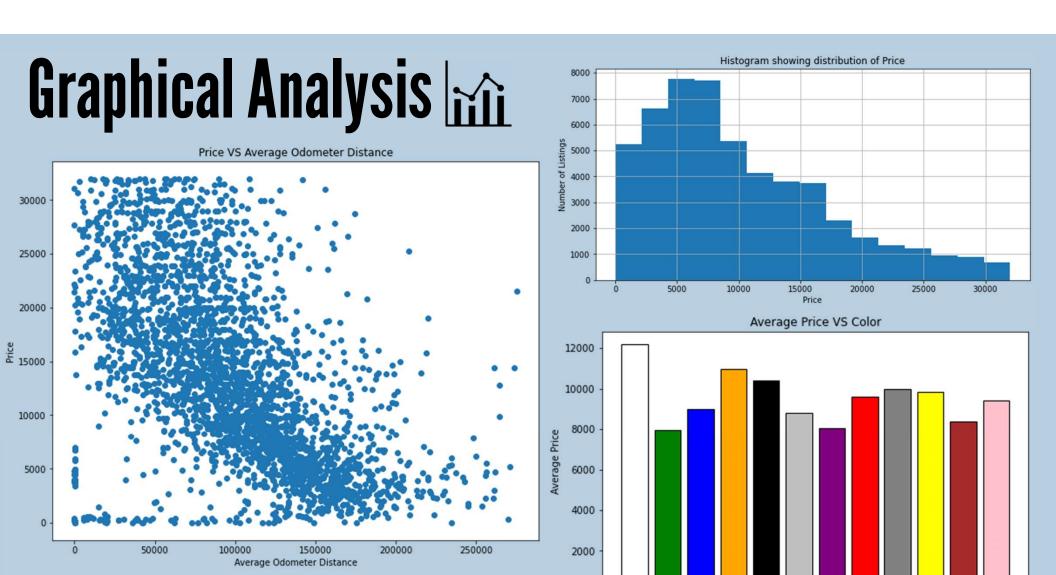
# **Correlation Analysis**

Attribute: Price

### Top 3 Correlated Attributes

- Odometer [Negative Correlation]:
   -0.4335
- Drive [Positive Correlation]:0.2686
- Fuel [Positive Correlation]: 0.1827

```
Ranked attributes:
 0.2686
        10 drive
 0.1827
          6 fuel
 0.1729 11 size
 0.1499
          4 condition
 0.1439
        12 type
 0.1424
          5 cylinders
 0.1109
          2 year
 0.0783
         13 paint color
 0.0625
           manufacturer
 0.0545
          9 transmission
 0.0196
           title status
-0.4335
            odometer
```



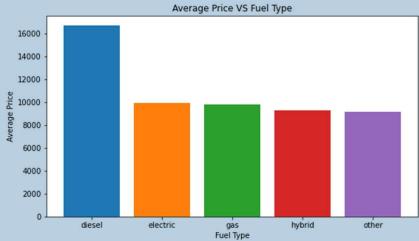
white green blue orange black silver purple red

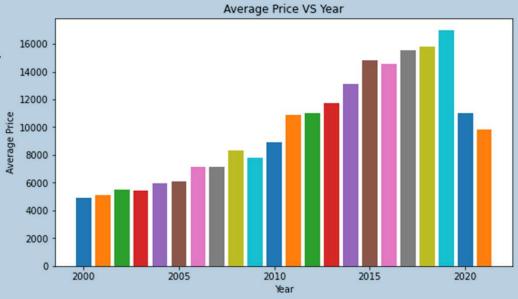
Color

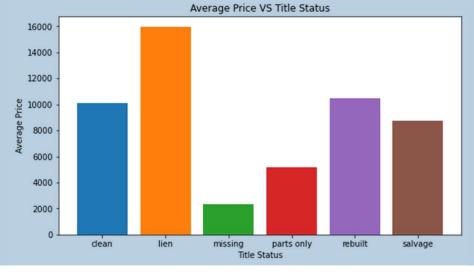
grey yellow brown custom

# Graphical Analysis initial





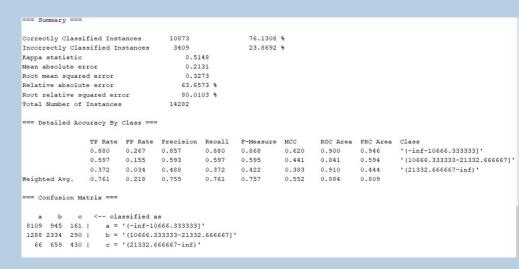




# Supervised Learning: Naive Bayes

### Dependent Variable: Price

- Price discretized into 3 bins of equal size
- Odometer discretized into 3 bins of equal frequency
- Duplicates Rows Removed
- Accuracy: 76.13% [Percentage Split 66% Test Option]



# Supervised Learning: Naive Bayes Contingency table and probability chart 1. Odometer

Count of odometer	Column Labels			
Row Labels	'\'(-inf-10666.333333]\'' '\'(1	0666.333333-21332.666667]\" '\	'(21332.666667-inf)\"	<b>Grand Total</b>
'\'(141002.5-inf)\''	11758	1932	289	13979
'\'(91000.5-141002.5]\''	9718	3620	676	14014
'\'(-inf-91000.5]\''	5619	5967	2427	14013
Grand Total	27095	11519	3392	42006
Probability				
Odometer/Price	'\'(-inf-10666.333333)	]\" '\'(10666.333333-21332.66	6667]\'' '\'(21332.666	6667-inf)\"
'\'(141002.5-inf)\''	0.4	43	0.17	0.09
'\'(91000.5-141002.5]	\" 0.3	36	0.31	0.20
"\'(-inf-91000.5]\"	0.2	21	0.52	0.72

### 3. Fuel

Count of fuel C	Column Labels			
Row Labels 💌 "	\'(-inf-10666.333333]\'' '\'(1	0666.333333-21332.666667]\" '\'(	(21332.666667-inf)\"	<b>Grand Total</b>
diesel	450	746	410	1606
electric	9	4	1	14
gas	26122	10559	2956	39637
hybrid	477	189	22	688
other	37	21	3	61
<b>Grand Total</b>	27095	11519	3392	42006
Probability				
Fuel/Price	'\'(-inf-10666.333333]\	" '\'(10666.333333-21332.6666	67]\'' '\'(21332.6666	667-inf)\''
diesel	0.02	2	0.06	0.12
electric	0.00	)	0.00	0.00
gas	0.96		0.92	0.87
hybrid	0.02		0.02	0.01
other	0.00		0.00	0.00

### 2. Drive

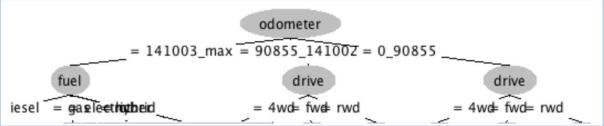
Count of drive Co	olumn Labels			
Row Labels " '\'(	(-inf-10666.333333]\" '\'(106	66.333333-21332.666667]\"	'\'(21332.666667-inf)\"	<b>Grand Total</b>
4wd	8223	5553	2179	15955
fwd	14869	3789	431	19089
rwd	4003	2177	782	6962
Grand Total	27095	11519	3392	42006
Probability				
Drive/Price	'\'(-inf-10666.333333]\'	' '\'(10666.333333-21332.66	6667]\" '\'(21332.666	667-inf)\"
4wd	0.30		0.48	0.64
fwd	0.55		0.33	0.13
rwd	0.15		0.19	0.23

# Supervised Learning: Decision Trees

### Dependent Variable: Price

- Price discretized into 3 bins of equal size
- Odometer discretized into 3 bins of equal frequency
- Duplicates Rows NOT removed
- Accuracy: 80.99% [Percentage Split 66% Test Option
- First Node Split on Odometer Variable
- Improved Accuracy from Naive Bayes Model

```
=== Summary ===
Correctly Classified Instances
                                    14702
                                                       80.9894 %
Incorrectly Classified Instances
                                     3451
                                                       19.0106 %
Kappa statistic
                                        0.641
Mean absolute error
                                        0.1711
Root mean squared error
                                        0.314
Relative absolute error
                                       48.2972 %
Root relative squared error
                                       74.4099 %
Total Number of Instances
                                    18153
=== Confusion Matrix ===
                    <-- classified as
                       a = '(-inf-10666.3333333)'
 9796 1069 177 |
 1234 3816
            283
                       b = (10666.333333-21332.666667)
  132 556 1090
                       c = '(21332.666667-inf)'
```



# Supervised Learning: Random Forests Andrews Burning: Random Forests

### Dependent Variable: Price

- Price discretized into 3 bins of equal size
- Odometer discretized into 3 bins of equal frequency
- Duplicates Rows <u>NOT</u> removed
- 3 Features selection gave most accurate model
- Accuracy: 84.95% [Percentage Split 66% Test Option]
- Removed Overfitting from Decision Trees
- Highly correlated attributes weren't used rather year, manufacturer and paint\_color helped most, so more attributes included

```
Attribute importance based on average impurity decrease (and number of nodes using that attribute)

0.76 ( 46042) year

0.63 ( 65740) manufacturer

0.58 ( 90404) paint_color

0.52 ( 46887) type

0.5 ( 52738) odometer

0.48 ( 76137) condition

0.47 ( 51734) cylinders

0.45 ( 62963) size

0.43 ( 48174) drive

0.42 ( 23730) fuel

0.37 ( 32408) title_status

0.36 ( 23289) transmission
```

```
Correctly Classified Instances
Incorrectly Classified Instances
Root mean squared error
Relative absolute error
oot relative squared error
               TP Rate FP Rate Precision Recall F-Measure MCC
              0.918 0.165 0.896
                                                                                       '(-inf-10666.3333333]'
               0.759 0.096
                              0.767
                                                 0.763
                                                            0.666
                                                                   0.927
                                                                             0.840
                                                                                       '(10666.333333-21332.6666671'
              0.696 0.020
                               0.788
                                         0.696
                                                 0.739
                                                            0.715
                                                                    0.969
                                                                             0.830
                                                                                       '(21332.666667=inf)'
```

```
=== Confusion Matrix ===

a b c <-- classified as

10134 804 104 | a = '(-inf-10666.333333]'

1055 4049 229 | b = '(10666.333333-21332.666667]'

116 424 1238 | c = '(21332.666667-inf)'
```

# Conclusions and Model Benefits

- Best Performing Model: Random Forests: 84.9501% accuracy
- Major Predictors for Price of Used Car: Odometer Distance, Drive, Fuel Type
- Benefits for seller: set a price range
- Benefits for buyer: set a budget

# **Limitations and Improvements**



- Numerical Data
- Details of Condition of Car and proper regional data
- Limited to only US Market region
- Uneven Distribution of certain attributes
- Seller/buyer information



# QUESTIONS 22