

Fuel Economy Analysis

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Introduction

The dataset used in this analysis comes from <https://www.fueleconomy.gov/feg/download.shtml>. The particular file is <https://www.fueleconomy.gov/feg/epadata/vehicles.csv.zip> “Datasets for All Model Years (1984–2019)”. The data dictionary is here: <https://www.fueleconomy.gov/feg/ws/index.shtml#vehicle>

In this analysis, we are trying to find out which manufacturer produces the most efficient fleet of cars. Also, looking for some interesting trends or insights like how fuel economy changed over time.

Load Data

```
library(caret)

## Loading required package: lattice
## Loading required package: ggplot2
library(reshape2)
library(glm2)
library(corrplot)

## corrplot 0.84 loaded
library(ggplot2)
library(DataExplorer)
library(xtable)
library(car)

## Loading required package: carData
library(dplyr)

##
## Attaching package: 'dplyr'

## The following object is masked from 'package:car':
##
##   recode

## The following objects are masked from 'package:stats':
##
##   filter, lag

## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union

# library(randomForest)

setwd("/Users/jyoti/Project1/")
```

```
filepath <- "/Users/jyoti/Project1/vehicles.csv"
vehicles_data = read.csv(filepath)
```

Exploratory Analysis

Explore the raw data by checking number of rows, columns, printing and plotting some content of variables and finding is there are any missing values.

```
print(paste0(paste0("Number of rows of the raw dataset: ", nrow(vehicles_data)), paste0(" Number of columns of the raw dataset: ", ncol(vehicles_data))))
```

```
## [1] "Number of rows of the raw dataset: 40081 Number of columns of raw dataset: 83"
```

```
names(vehicles_data)
```

```
## [1] "barrels08"      "barrelsA08"      "charge120"
## [4] "charge240"      "city08"          "city08U"
## [7] "cityA08"        "cityA08U"        "cityCD"
## [10] "cityE"          "cityUF"          "co2"
## [13] "co2A"           "co2TailpipeAGpm" "co2TailpipeGpm"
## [16] "comb08"         "comb08U"         "combA08"
## [19] "combA08U"       "combE"           "combinedCD"
## [22] "combinedUF"     "cylinders"       "displ"
## [25] "drive"          "engId"           "eng_dscr"
## [28] "feScore"        "fuelCost08"      "fuelCostA08"
## [31] "fuelType"       "fuelType1"       "ghgScore"
## [34] "ghgScoreA"      "highway08"       "highway08U"
## [37] "highwayA08"     "highwayA08U"     "highwayCD"
## [40] "highwayE"       "highwayUF"       "hlv"
## [43] "hvp"           "id"              "lv2"
## [46] "lv4"           "make"            "model"
## [49] "mpgData"        "phevBlended"     "pv2"
## [52] "pv4"           "range"           "rangeCity"
## [55] "rangeCityA"     "rangeHwy"        "rangeHwyA"
## [58] "trany"          "UCity"           "UCityA"
## [61] "UHighway"       "UHighwayA"       "VClass"
## [64] "year"          "youSaveSpend"    "guzzler"
## [67] "trans_dscr"     "tCharger"        "sCharger"
## [70] "atvType"        "fuelType2"       "rangeA"
## [73] "evMotor"        "mfrCode"         "c240Dscr"
## [76] "charge240b"     "c240bDscr"       "createdOn"
## [79] "modifiedOn"     "startStop"       "phevCity"
## [82] "phevHwy"       "phevComb"
```

```
head(vehicles_data)
```

```
## barrels08 barrelsA08 charge120 charge240 city08 city08U cityA08 cityA08U
## 1 15.69571          0          0          0      19          0          0          0
## 2 29.96455          0          0          0       9          0          0          0
## 3 12.20778          0          0          0      23          0          0          0
## 4 29.96455          0          0          0      10          0          0          0
## 5 17.34789          0          0          0      17          0          0          0
## 6 14.98227          0          0          0      21          0          0          0
## cityCD cityE cityUF co2 co2A co2TailpipeAGpm co2TailpipeGpm comb08
## 1      0      0      0 -1  -1              0      423.1905      21
## 2      0      0      0 -1  -1              0      807.9091      11
```

```

## 3      0      0      0 -1 -1      0      329.1481      27
## 4      0      0      0 -1 -1      0      807.9091      11
## 5      0      0      0 -1 -1      0      467.7368      19
## 6      0      0      0 -1 -1      0      403.9545      22
##      comb08U combA08 combA08U combE combinedCD combinedUF cylinders displ
## 1      0      0      0      0      0      0      4      2.0
## 2      0      0      0      0      0      0      12      4.9
## 3      0      0      0      0      0      0      4      2.2
## 4      0      0      0      0      0      0      8      5.2
## 5      0      0      0      0      0      0      4      2.2
## 6      0      0      0      0      0      0      4      1.8
##
##              drive engId      eng_dscr feScore fuelCost08
## 1      Rear-Wheel Drive  9011      (FFS)      -1      2000
## 2      Rear-Wheel Drive 22020      (GUZZLER)      -1      3850
## 3      Front-Wheel Drive  2100      (FFS)      -1      1550
## 4      Rear-Wheel Drive  2850      -1      3850
## 5 4-Wheel or All-Wheel Drive 66031 (FFS,TRBO)      -1      2700
## 6      Front-Wheel Drive 66020      (FFS)      -1      1950
##      fuelCostA08 fuelType      fuelType1 ghgScore ghgScoreA highway08
## 1      0 Regular Regular Gasoline      -1      -1      25
## 2      0 Regular Regular Gasoline      -1      -1      14
## 3      0 Regular Regular Gasoline      -1      -1      33
## 4      0 Regular Regular Gasoline      -1      -1      12
## 5      0 Premium Premium Gasoline      -1      -1      23
## 6      0 Regular Regular Gasoline      -1      -1      24
##      highway08U highwayA08 highwayA08U highwayCD highwayE highwayUF hlv hpv
## 1      0      0      0      0      0      0      0      0      0
## 2      0      0      0      0      0      0      0      0      0
## 3      0      0      0      0      0      0      0      19      77
## 4      0      0      0      0      0      0      0      0      0
## 5      0      0      0      0      0      0      0      0      0
## 6      0      0      0      0      0      0      0      0      0
##
##      id lv2 lv4      make      model mpgData phevBlended pv2 pv4
## 1      1      0      0 Alfa Romeo Spider Veloce 2000      Y      false      0      0
## 2      10      0      0 Ferrari      Testarossa      N      false      0      0
## 3      100      0      0 Dodge      Charger      Y      false      0      0
## 4      1000      0      0 Dodge B150/B250 Wagon 2WD      N      false      0      0
## 5      10000      0      14 Subaru Legacy AWD Turbo      N      false      0      90
## 6      10001      0      15 Subaru      Loyale      N      false      0      88
##
##      range rangeCity rangeCityA rangeHwy rangeHwyA      tranny      UCity
## 1      0      0      0      0      0      Manual 5-spd 23.3333
## 2      0      0      0      0      0      Manual 5-spd 11.0000
## 3      0      0      0      0      0      Manual 5-spd 29.0000
## 4      0      0      0      0      0 Automatic 3-spd 12.2222
## 5      0      0      0      0      0      Manual 5-spd 21.0000
## 6      0      0      0      0      0 Automatic 3-spd 27.0000
##
##      UCityA UHighway UHighwayA      VClass year youSaveSpend guzzler
## 1      0      35.0000      0      Two Seaters 1985      -2250
## 2      0      19.0000      0      Two Seaters 1985      -11500      T
## 3      0      47.0000      0 Subcompact Cars 1985      0
## 4      0      16.6667      0      Vans 1985      -11500
## 5      0      32.0000      0 Compact Cars 1993      -5750
## 6      0      33.0000      0 Compact Cars 1993      -2000
##
##      trans_dscr tCharger sCharger atvType fuelType2 rangeA evMotor mfrCode

```

```
## 1      NA
## 2      NA
## 3      SIL      NA
## 4      NA
## 5      TRUE
## 6      NA
##      c240Dscr charge240b c240bDscr      createdOn
## 1      0      Tue Jan 01 00:00:00 EST 2013
## 2      0      Tue Jan 01 00:00:00 EST 2013
## 3      0      Tue Jan 01 00:00:00 EST 2013
## 4      0      Tue Jan 01 00:00:00 EST 2013
## 5      0      Tue Jan 01 00:00:00 EST 2013
## 6      0      Tue Jan 01 00:00:00 EST 2013
##      modifiedOn startStop phевCity phевHwy phевComb
## 1 Tue Jan 01 00:00:00 EST 2013      0      0      0
## 2 Tue Jan 01 00:00:00 EST 2013      0      0      0
## 3 Tue Jan 01 00:00:00 EST 2013      0      0      0
## 4 Tue Jan 01 00:00:00 EST 2013      0      0      0
## 5 Tue Jan 01 00:00:00 EST 2013      0      0      0
## 6 Tue Jan 01 00:00:00 EST 2013      0      0      0
```

```
# str(vehicles_data)
# summary(vehicles_data)
glimpse(vehicles_data)
```

```
## Observations: 40,081
## Variables: 83
## $ barrels08      <dbl> 15.69571, 29.96455, 12.20778, 29.96455, 17.347...
## $ barrelsA08     <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ charge120      <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ charge240      <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ city08         <int> 19, 9, 23, 10, 17, 21, 22, 23, 23, 23, 23, 18,...
## $ city08U        <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ cityA08        <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ cityA08U       <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ cityCD         <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ cityE          <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ cityUF         <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ co2            <int> -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1...
## $ co2A           <int> -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1...
## $ co2TailpipeAGpm <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ co2TailpipeGpm <dbl> 423.1905, 807.9091, 329.1481, 807.9091, 467.73...
## $ comb08         <int> 21, 11, 27, 11, 19, 22, 25, 24, 26, 25, 26, 21...
## $ comb08U        <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ combA08        <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ combA08U       <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ combE          <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ combinedCD     <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ combinedUF     <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ cylinders      <int> 4, 12, 4, 8, 4, 4, 4, 4, 4, 4, 4, 4, 4, 4, 8, ...
## $ displ          <dbl> 2.0, 4.9, 2.2, 5.2, 2.2, 1.8, 1.8, 1.6, 1.6, 1...
## $ drive          <fct> Rear-Wheel Drive, Rear-Wheel Drive, Front-Whee...
## $ engId          <int> 9011, 22020, 2100, 2850, 66031, 66020, 66020, ...
## $ eng_dscr       <fct> (FFS), (GUZZLER), (FFS), , (FFS,TRBO), (FFS), ...
## $ feScore        <int> -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1...
```

```

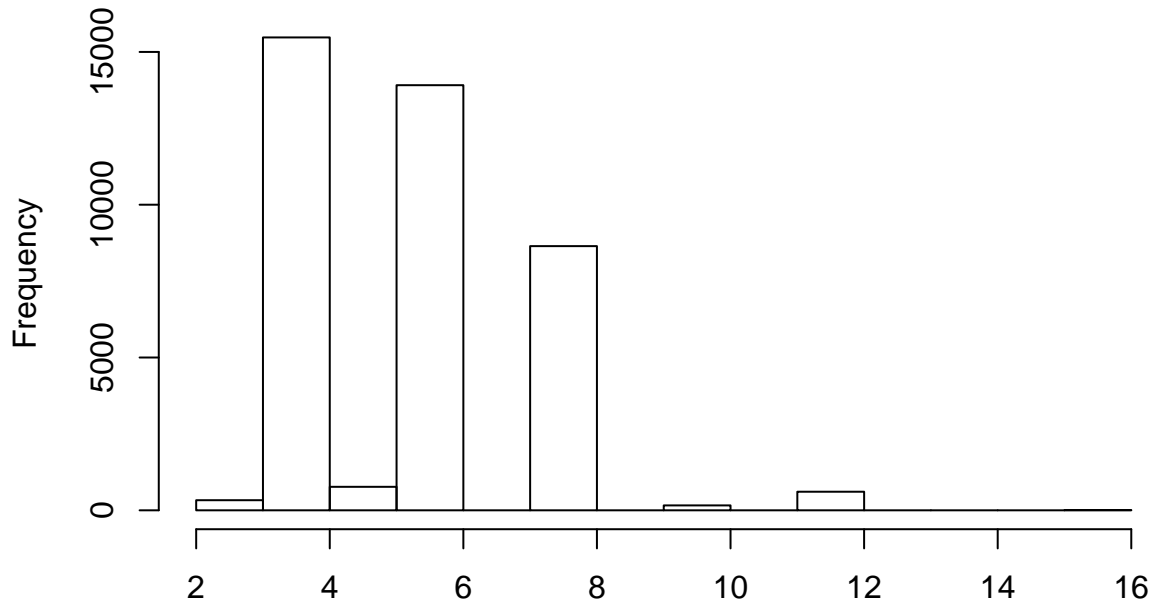
## $ fuelCost08      <int> 2000, 3850, 1550, 3850, 2700, 1950, 1700, 1750...
## $ fuelCostA08     <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ fuelType        <fct> Regular, Regular, Regular, Regular, Premium, R...
## $ fuelType1       <fct> Regular Gasoline, Regular Gasoline, Regular Ga...
## $ ghgScore        <int> -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1...
## $ ghgScoreA       <int> -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1...
## $ highway08       <int> 25, 14, 33, 12, 23, 24, 29, 26, 31, 30, 30, 26...
## $ highway08U      <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ highwayA08      <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ highwayA08U     <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ highwayCD       <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ highwayE        <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ highwayUF       <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ hlv             <int> 0, 0, 19, 0, 0, 0, 0, 0, 0, 0, 0, 17, 17, 0, 0...
## $ hpv             <int> 0, 0, 77, 0, 0, 0, 0, 0, 0, 0, 0, 88, 88, 0, 0...
## $ id              <int> 1, 10, 100, 1000, 10000, 10001, 10002, 10003, ...
## $ lv2             <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ lv4             <int> 0, 0, 0, 0, 14, 15, 15, 13, 13, 13, 13, 0, 0, ...
## $ make            <fct> Alfa Romeo, Ferrari, Dodge, Dodge, Subaru, Sub...
## $ model           <fct> Spider Veloce 2000, Testarossa, Charger, B150/...
## $ mpgData         <fct> Y, N, Y, N, N, N, Y, Y, Y, Y, Y, N, Y, N, N...
## $ phevBlended     <fct> false, false, false, false, false, false, false, fals...
## $ pv2             <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ pv4             <int> 0, 0, 0, 0, 90, 88, 88, 89, 89, 89, 89, 0, 0, ...
## $ range           <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ rangeCity       <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ rangeCityA      <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ rangeHwy        <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ rangeHwyA       <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ trany           <fct> Manual 5-spd, Manual 5-spd, Manual 5-spd, Auto...
## $ UCity           <dbl> 23.3333, 11.0000, 29.0000, 12.2222, 21.0000, 2...
## $ UCityA          <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ UHighway        <dbl> 35.0000, 19.0000, 47.0000, 16.6667, 32.0000, 3...
## $ UHighwayA       <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ VClass          <fct> Two Seaters, Two Seaters, Subcompact Cars, Van...
## $ year            <int> 1985, 1985, 1985, 1985, 1993, 1993, 1993, 1993...
## $ youSaveSpend    <int> -2250, -11500, 0, -11500, -5750, -2000, -750, ...
## $ guzzler         <fct> , T, , , , , , , , , , , , , , , , T, T,...
## $ trans_dscr      <fct> , , SIL, , , , , , , , , 2MODE CLKUP, , 2MODE ...
## $ tCharger        <lgl> NA, NA, NA, NA, TRUE, NA, NA, NA, NA, NA, NA, ...
## $ sCharger        <fct> , , , , , , , , , , , , , , , , , , , ,
## $ atvType         <fct> , , , , , , , , , , , , , , , , , , , ,
## $ fuelType2       <fct> , , , , , , , , , , , , , , , , , , , ,
## $ rangeA          <fct> , , , , , , , , , , , , , , , , , , , ,
## $ evMotor         <fct> , , , , , , , , , , , , , , , , , , , ,
## $ mfrCode         <fct> , , , , , , , , , , , , , , , , , , , ,
## $ c240Dscr        <fct> , , , , , , , , , , , , , , , , , , , ,
## $ charge240b      <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ c240bDscr       <fct> , , , , , , , , , , , , , , , , , , , ,
## $ createdOn       <fct> Tue Jan 01 00:00:00 EST 2013, Tue Jan 01 00:00...
## $ modifiedOn      <fct> Tue Jan 01 00:00:00 EST 2013, Tue Jan 01 00:00...
## $ startStop       <fct> , , , , , , , , , , , , , , , , , , , ,
## $ phevCity        <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
## $ phevHwy         <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...

```

```
## $ phevComb      <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
```

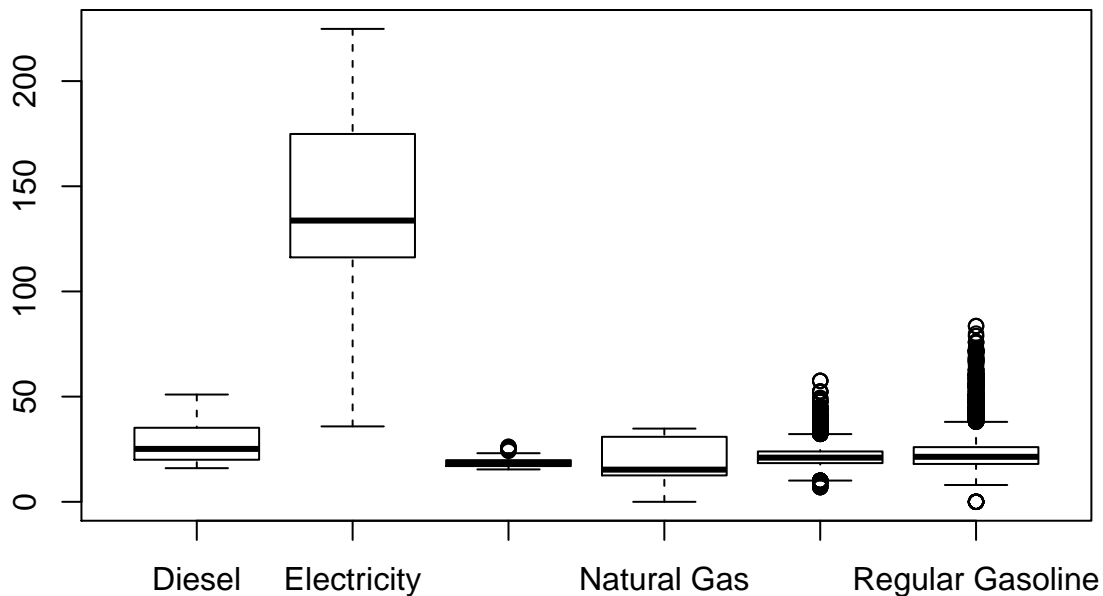
```
hist(vehicles_data$cylinders)
```

Histogram of vehicles_data\$cylinders



vehicles_data\$cylinders

```
plot(vehicles_data$fuelType1, vehicles_data$UCity)
```



```
#Find missing values in data
```

```
missing_values <- sapply(vehicles_data, function(x) sum(is.na(x)))
```

```
missing_values[missing_values > 0]
```

```
## cylinders      displ  tCharger
```

```
##          171          169          33779
```

Subset and Clean Data

Looking at the data, the data dictionary and the problem statement, we can now select the features that we intuitively find relevant to the analysis. Below are the columns selected from dataset:

- “cylinders” - engine cylinders
- “displ” - engine displacement in liters
- “drive” - drive axle type
- “feScore” - EPA Fuel Economy Score (-1 = Not available)
- “make” - manufacturer (division)
- “trany” - transmission
- “fuelType1” - fuel type 1. For single fuel vehicles, this will be the only fuel. For dual fuel vehicles, this will be the conventional fuel
- “phevBlended” - if true, this vehicle operates on a blend of gasoline and electricity in charge depleting mode
- “VClass” - EPA vehicle size class
- “UCity” - unadjusted city MPG for fuelType1
- “year” - model year

```
required <- c('cylinders', 'displ', 'drive', 'feScore', 'make', 'trany', 'fuelType1', 'phevBlended', 'V
```

```
#Subset the data to get desired features
```

```
vehicles_desired <- vehicles_data[, (names(vehicles_data) %in% required)]
```

```
# names(vehicles_desired)
```

```
#get count of NAs in feScore column (with value = -1)
```

```
sum(vehicles_desired$feScore==-1)
```

```
## [1] 32027
```

```
#drop feScore from features
```

```
vehicles_desired <- vehicles_desired[, !(names(vehicles_desired) %in% c("feScore"))]
```

```
#get rid of NAs from Cylinders and displ columns by dropping rows (since the NAs are few)
```

```
vehicles_desired <- vehicles_desired[complete.cases(vehicles_desired),] #This also gets rid of Electric
```

```
#Treating 0 UCity as bad data
```

```
sum(vehicles_desired$UCity==0)
```

```
## [1] 25
```

```
#Drop rows with 0 Ucity
```

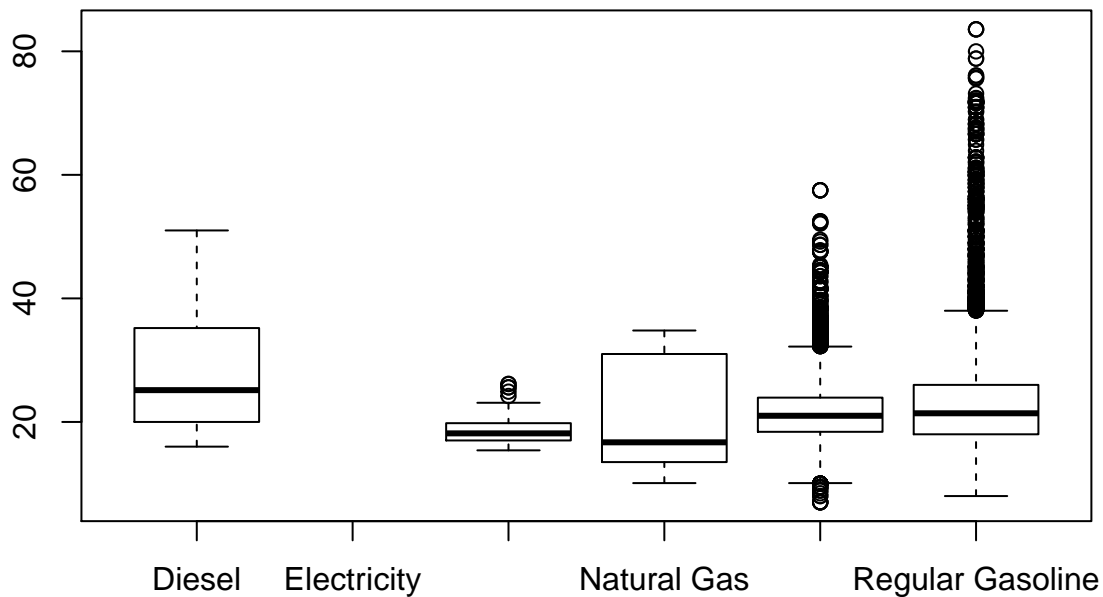
```
vehicles_desired <- vehicles_desired[!vehicles_desired$UCity==0,]
```

```
summary(vehicles_desired)
```

```
##      cylinders      displ      drive
##  Min.   : 2.000   Min.   :0.6   Front-Wheel Drive   :13876
##  1st Qu.: 4.000   1st Qu.:2.2   Rear-Wheel Drive    :13475
##  Median : 6.000   Median :3.0   4-Wheel or All-Wheel Drive: 6642
##  Mean   : 5.721   Mean   :3.3   All-Wheel Drive     : 2675
##  3rd Qu.: 6.000   3rd Qu.:4.3   4-Wheel Drive       : 1325
##  Max.   :16.000   Max.   :8.4   (Other)             :  711
```

```
##           fuelType1           make           phevBlended
## Diesel      : 1142   Chevrolet: 3933   false:39809
## Electricity :    0    Ford      : 3257   true :    76
## Midgrade Gasoline: 100 Dodge      : 2555
## Natural Gas   :   54   GMC        : 2465
## Premium Gasoline :11267 Toyota    : 2003
## Regular Gasoline :27322 BMW        : 1848
##              (Other) :23824
##           tranny           UCity
## Automatic 4-spd:11021   Min.    : 7.00
## Manual 5-spd  : 8348   1st Qu.:18.10
## Automatic 3-spd: 3150   Median :21.20
## Automatic (S6) : 2984   Mean    :22.50
## Manual 6-spd   : 2671   3rd Qu.:25.59
## Automatic 5-spd: 2198   Max.    :33.56
## (Other)       : 9513
##           VClass           year
## Compact Cars      : 5738   Min.    :1984
## Subcompact Cars   : 5016   1st Qu.:1991
## Midsize Cars      : 4675   Median :2002
## Standard Pickup Trucks : 2354 Mean    :2001
## Sport Utility Vehicle - 4WD: 2090 3rd Qu.:2011
## Large Cars        : 2032   Max.    :2019
## (Other)           :17980
```

```
plot(vehicles_desired$fuelType1, vehicles_desired$UCity)
```



Finding manufacturer with most fuel efficient fleet

```
#Helper function for plotting multiple plots together
multiplot <- function(..., plotlist=NULL, cols=1, layout=NULL) {
  library(grid)

  # Make a list from the ... arguments and plotlist
```



```

plots <- c(list(...), plotlist)

numPlots = length(plots)

# If layout is NULL, then use 'cols' to determine layout
if (is.null(layout)) {
  # Make the panel
  # ncol: Number of columns of plots
  # nrow: Number of rows needed, calculated from # of cols
  layout <- matrix(seq(1, cols * ceiling(numPlots/cols)),
                    ncol = cols, nrow = ceiling(numPlots/cols))
}

if (numPlots==1) {
  print(plots[[1]])
} else {
  # Set up the page
  grid.newpage()
  pushViewport(viewport(layout = grid.layout(nrow(layout), ncol(layout))))

  # Make each plot, in the correct location
  for (i in 1:numPlots) {
    # Get the i,j matrix positions of the regions that contain this subplot
    matchidx <- as.data.frame(which(layout == i, arr.ind = TRUE))

    print(plots[[i]], vp = viewport(layout.pos.row = matchidx$row,
                                    layout.pos.col = matchidx$col))
  }
}

```

To compare fuel efficiency among different car manufacturers, it only makes sense to do so over a particular model year. The following comparison thus is done for 2018 model year. Also, our data cleaning step excludes the 'Electricity' fueltype as it does not make sense to compare electric vehicle with other fuel type vehicles for MPG.

```

#Subsetting for year 2018
vehicles_2018 <- vehicles_desired[vehicles_desired$year==2018,]

# mean(vehicles_2018$UCity)
mean_mpg_per_make <- aggregate(vehicles_2018$UCity, list(vehicles_2018$make), mean)
median_mpg_per_make <- aggregate(vehicles_2018$UCity, list(vehicles_2018$make), median)

top_mean <- mean_mpg_per_make[order(mean_mpg_per_make$x,decreasing=T)[1:3],]
top_median <- median_mpg_per_make[order(median_mpg_per_make$x,decreasing=T)[1:3],]

barplot1 <- ggplot(data=top_mean, aes(x=Group.1, y=x)) +
  geom_bar(stat="identity", fill="steelblue") +
  xlab("Maker") +
  ylab("Avg. MPG") +
  ggtitle("Top Avg. MPG Makers")
# print(barplot1)

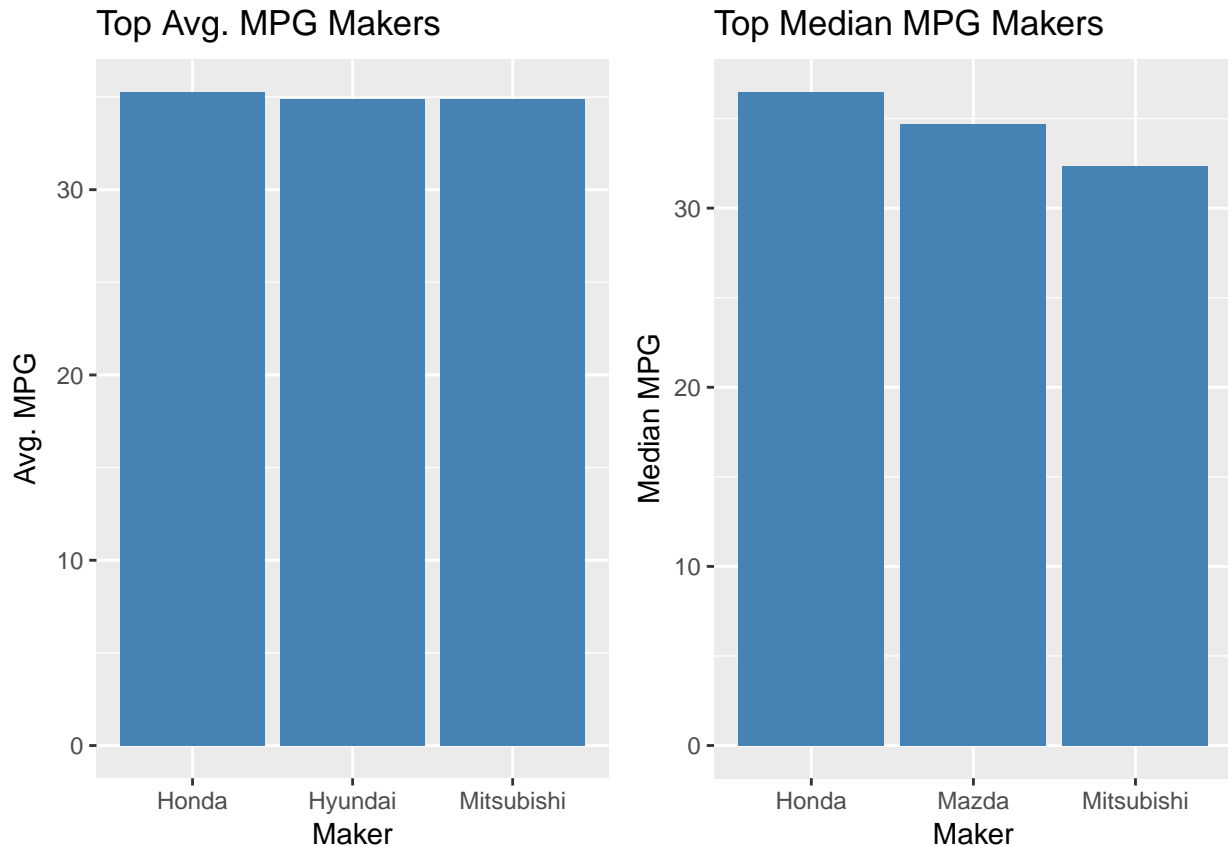
```

```

barplot2 <- ggplot(data=top_median, aes(x=Group.1, y=x)) +
  geom_bar(stat="identity", fill="steelblue") +
  xlab("Maker") +
  ylab("Median MPG") +
  ggtitle("Top Median MPG Makers")
# print(barplot2)

multiplot(barplot1, barplot2, cols =2)

```



```

print(paste0(top_mean[1,1], paste0(" has the most fuel efficient fleet with average MPG as ", round(top
## [1] "Honda has the most fuel efficient fleet with average MPG as 35.27"

```

Correlation Analysis

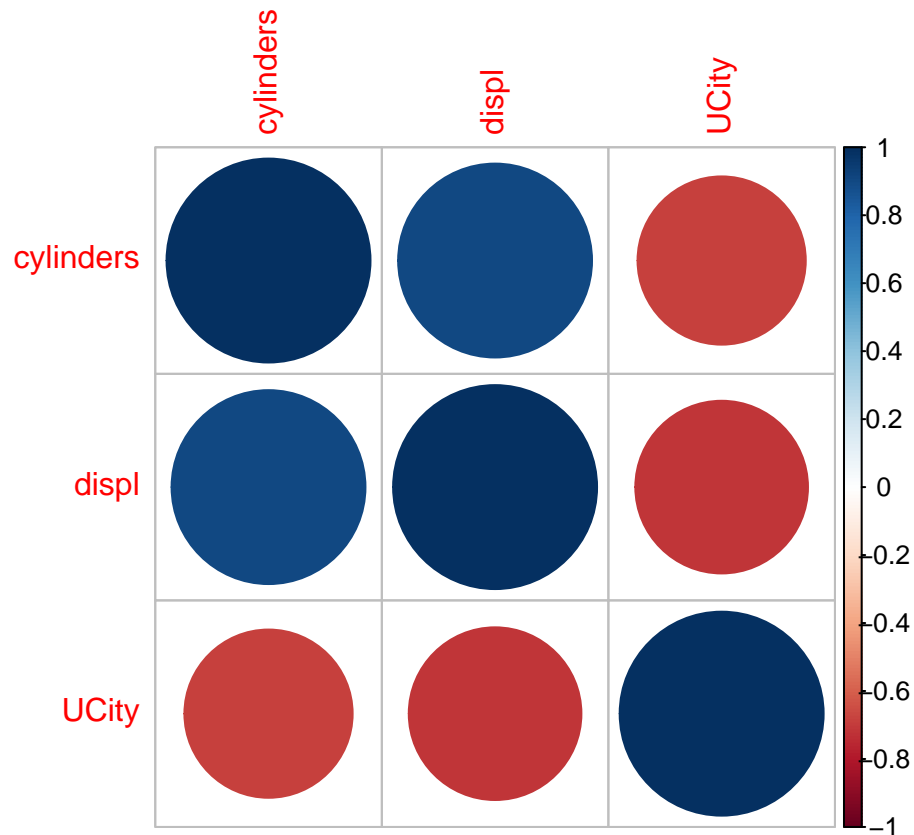
Checking for correlations between different variables.

```

numeric_features <- c('cylinders', 'displ', 'UCity')
vehicles_numeric_data <- vehicles_desired[, names(vehicles_desired) %in% numeric_features]

correlations <- cor(vehicles_numeric_data)
corrplot(correlations)

```

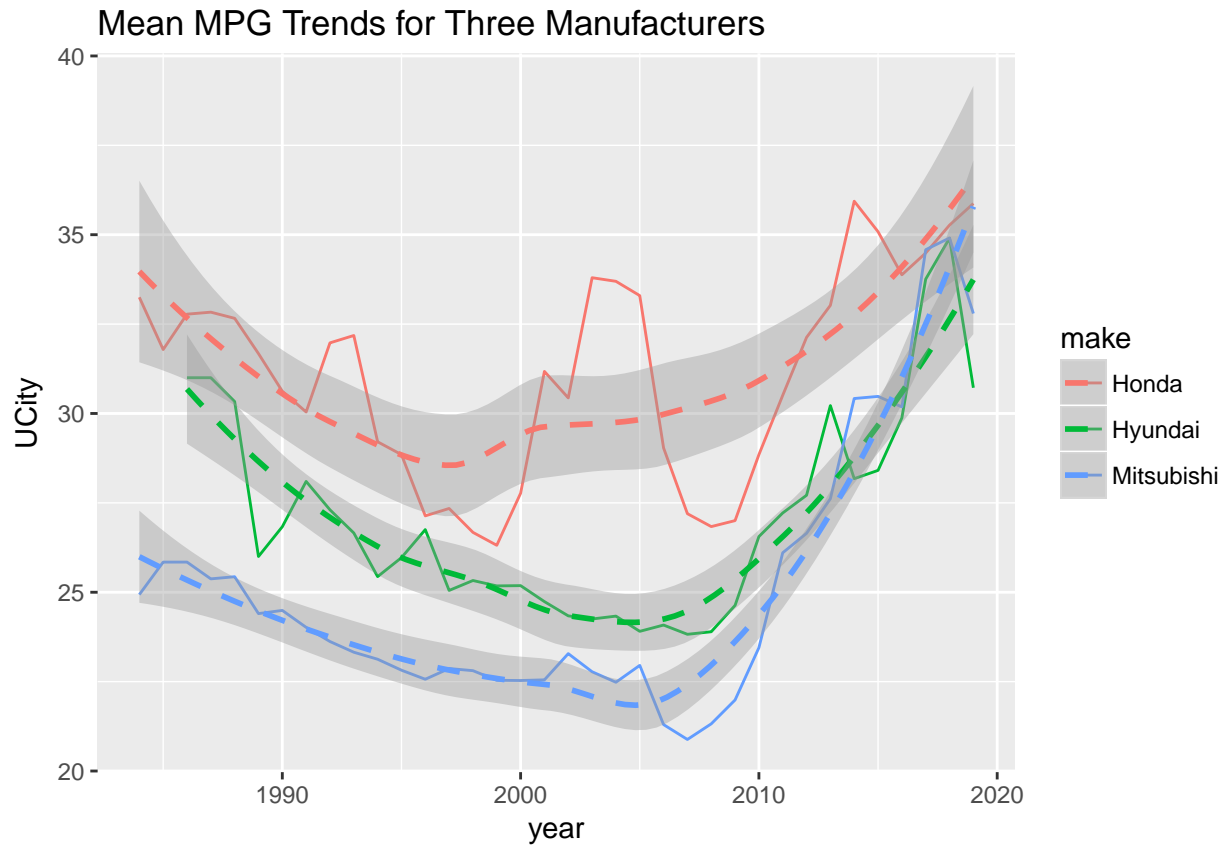


The two numeric features, cylinders and displ, both are well correlated with target UCity (MPG) but also highly correlated with each other as seen from the above correlation graph. So, one of the two variables might be required to be removed depending on Variance Inflation Factor of final model.

Other Trends

```
#Subsetting for only top 3 manufacturers as our previous analysis
vehicles_trends <- vehicles_desired[vehicles_desired$make %in% c("Honda", "Hyundai", "Mitsubishi"),]

mean_mpg <- aggregate( UCity ~ make + year, data = vehicles_trends, mean)
plot1 <- ggplot(data = mean_mpg, aes(x=year, y=UCity, colour = make)) +
  geom_line(aes(group = make)) +
  geom_smooth(method = 'loess', linetype = 2) +
  ggtitle("Mean MPG Trends for Three Manufacturers")
plot1
```



As seen from the graph, the average MPG is generally going up for the shown manufacturers since 2005, probably because of external factors like recent push from authorities to reduce fuel consumption or rising fuel costs. The future MPG thus should be higher than current values. One can further explore this with a time series model.