

Data Analytics 101 Course Project: Car Price Predictor

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2024-11-30

Brief introduction of the project: This project aims to build predictive models for car prices using two datasets: Cars93 and CarPrice_Assignment

Load both cars data

```
cars93 <- read.csv("Cars93.csv")
cars_assign = read.csv("CarPrice_Assignment.csv")
```

Standard deviations for cars__assign

```
sd(cars_assign$enginesize)
```

```
## [1] 41.64269
```

```
sd(cars_assign$horsepower)
```

```
## [1] 39.54417
```

```
sd(cars_assign$curbweight)
```

```
## [1] 520.6802
```

```
sd(cars_assign$price)
```

```
## [1] 7988.852
```

Standard deviations for cars93

```
sd(cars93$EngineSize)
```

```
## [1] 1.037363
```

```
sd(cars93$Horsepower)
```

```
## [1] 52.37441
```

```
sd(cars93$Weight)
```

```
## [1] 589.8965
```

```
knitr::opts_chunk$set(echo = TRUE)
library(ggplot2) # Ensure ggplot2 is loaded
```

Dataset 1 Summary

```
summary(cars_assign)
```

```
##      car_ID      symboling      CarName      fueltype
## Min.   : 1      Min.   :-2.0000      Length:205      Length:205
## 1st Qu.: 52      1st Qu.: 0.0000      Class :character  Class :character
## Median :103      Median : 1.0000      Mode  :character  Mode  :character
## Mean   :103      Mean   : 0.8341
## 3rd Qu.:154      3rd Qu.: 2.0000
## Max.   :205      Max.   : 3.0000
##      aspiration      doornumber      carbody      drivewheel
## Length:205      Length:205      Length:205      Length:205
## Class :character  Class :character  Class :character  Class :character
## Mode  :character  Mode  :character  Mode  :character  Mode  :character
##
##
##      enginelocation      wheelbase      carlength      carwidth
## Length:205      Min.   : 86.60      Min.   :141.1      Min.   :60.30
## Class :character  1st Qu.: 94.50      1st Qu.:166.3      1st Qu.:64.10
## Mode  :character  Median : 97.00      Median :173.2      Median :65.50
##      Mean   : 98.76      Mean   :174.0      Mean   :65.91
##      3rd Qu.:102.40      3rd Qu.:183.1      3rd Qu.:66.90
##      Max.   :120.90      Max.   :208.1      Max.   :72.30
##      carheight      curbweight      enginetype      cylindernumber
## Min.   :47.80      Min.   :1488      Length:205      Length:205
## 1st Qu.:52.00      1st Qu.:2145      Class :character  Class :character
## Median :54.10      Median :2414      Mode  :character  Mode  :character
## Mean   :53.72      Mean   :2556
## 3rd Qu.:55.50      3rd Qu.:2935
## Max.   :59.80      Max.   :4066
##      enginesize      fuelsystem      boreratio      stroke
## Min.   : 61.0      Length:205      Min.   :2.54      Min.   :2.070
## 1st Qu.: 97.0      Class :character  1st Qu.:3.15      1st Qu.:3.110
## Median :120.0      Mode  :character  Median :3.31      Median :3.290
## Mean   :126.9      Mean   :3.33      Mean   :3.255
## 3rd Qu.:141.0      3rd Qu.:3.58      3rd Qu.:3.410
## Max.   :326.0      Max.   :3.94      Max.   :4.170
##      compressionratio      horsepower      peakrpm      citympg
## Min.   : 7.00      Min.   : 48.0      Min.   :4150      Min.   :13.00
## 1st Qu.: 8.60      1st Qu.: 70.0      1st Qu.:4800      1st Qu.:19.00
## Median : 9.00      Median : 95.0      Median :5200      Median :24.00
## Mean   :10.14      Mean   :104.1      Mean   :5125      Mean   :25.22
## 3rd Qu.: 9.40      3rd Qu.:116.0      3rd Qu.:5500      3rd Qu.:30.00
```

```
## Max. :23.00 Max. :288.0 Max. :6600 Max. :49.00
## highwaympg price
## Min. :16.00 Min. : 5118
## 1st Qu.:25.00 1st Qu.: 7788
## Median :30.00 Median :10295
## Mean :30.75 Mean :13277
## 3rd Qu.:34.00 3rd Qu.:16503
## Max. :54.00 Max. :45400
```

Dataset 2 Summary

```
summary(cars93)
```

```
##      id      Manufacturer      Model      Type
## Min.   : 1  Length:93      Length:93      Length:93
## 1st Qu.:24  Class :character  Class :character  Class :character
## Median :47  Mode  :character  Mode  :character  Mode  :character
## Mean   :47
## 3rd Qu.:70
## Max.   :93
##
##      Min.Price      Price      Max.Price      MPG.city      MPG.highway
## Min.   : 6.70  Min.   : 7.40  Min.   : 7.9  Min.   :15.00  Min.   :20.00
## 1st Qu.:10.80  1st Qu.:12.20  1st Qu.:14.7  1st Qu.:18.00  1st Qu.:26.00
## Median :14.70  Median :17.70  Median :19.6  Median :21.00  Median :28.00
## Mean   :17.13  Mean   :19.51  Mean   :21.9  Mean   :22.37  Mean   :29.09
## 3rd Qu.:20.30  3rd Qu.:23.30  3rd Qu.:25.3  3rd Qu.:25.00  3rd Qu.:31.00
## Max.   :45.40  Max.   :61.90  Max.   :80.0  Max.   :46.00  Max.   :50.00
##
##      AirBags      DriveTrain      Cylinders      EngineSize
## Length:93      Length:93      Length:93      Min.   :1.000
## Class :character  Class :character  Class :character  1st Qu.:1.800
## Mode  :character  Mode  :character  Mode  :character  Median :2.400
##                                     Mean   :2.668
##                                     3rd Qu.:3.300
##                                     Max.   :5.700
##
##      Horsepower      RPM      Rev.per.mile  Man.trans.avail
## Min.   : 55.0  Min.   :3800  Min.   :1320  Length:93
## 1st Qu.:103.0  1st Qu.:4800  1st Qu.:1985  Class :character
## Median :140.0  Median :5200  Median :2340  Mode  :character
## Mean   :143.8  Mean   :5281  Mean   :2332
## 3rd Qu.:170.0  3rd Qu.:5750  3rd Qu.:2565
## Max.   :300.0  Max.   :6500  Max.   :3755
##
##      Fuel.tank.capacity  Passengers      Length      Wheelbase
## Min.   : 9.20  Min.   :2.000  Min.   :141.0  Min.   : 90.0
## 1st Qu.:14.50  1st Qu.:4.000  1st Qu.:174.0  1st Qu.: 98.0
## Median :16.40  Median :5.000  Median :183.0  Median :103.0
## Mean   :16.66  Mean   :5.086  Mean   :183.2  Mean   :103.9
## 3rd Qu.:18.80  3rd Qu.:6.000  3rd Qu.:192.0  3rd Qu.:110.0
## Max.   :27.00  Max.   :8.000  Max.   :219.0  Max.   :119.0
##
```

```
##      Width      Turn.circle  Rear.seat.room  Luggage.room      Weight
##  Min.   :60.00   Min.   :32.00   Min.   :19.00   Min.    : 6.00   Min.   :1695
##  1st Qu.:67.00   1st Qu.:37.00   1st Qu.:26.00   1st Qu.:12.00   1st Qu.:2620
##  Median :69.00   Median :39.00   Median :27.50   Median :14.00   Median :3040
##  Mean   :69.38   Mean   :38.96   Mean   :27.83   Mean   :13.89   Mean   :3073
##  3rd Qu.:72.00   3rd Qu.:41.00   3rd Qu.:30.00   3rd Qu.:15.00   3rd Qu.:3525
##  Max.   :78.00   Max.   :45.00   Max.   :36.00   Max.   :22.00   Max.   :4105
##
##      NA's      :2      NA's      :11
##      Origin      Make
##  Length:93      Length:93
##  Class :character  Class :character
##  Mode  :character  Mode  :character
##
##
##
##
```

Scatter

```
cars93$Passengers
```

```
## [1] 5 5 5 6 4 6 6 6 5 6 5 5 5 4 6 7 8 6 2 6 6 6 5 5 6 7 6 4 5 6 4 5 5 4 4 7 5 6
## [39] 4 4 4 4 4 5 5 4 5 5 5 4 6 6 4 5 5 7 2 5 5 4 5 5 5 5 5 7 5 5 5 7 6 4 4 5 4 5
## [77] 6 5 5 4 5 5 4 5 4 5 7 4 7 5 4 5 5
```

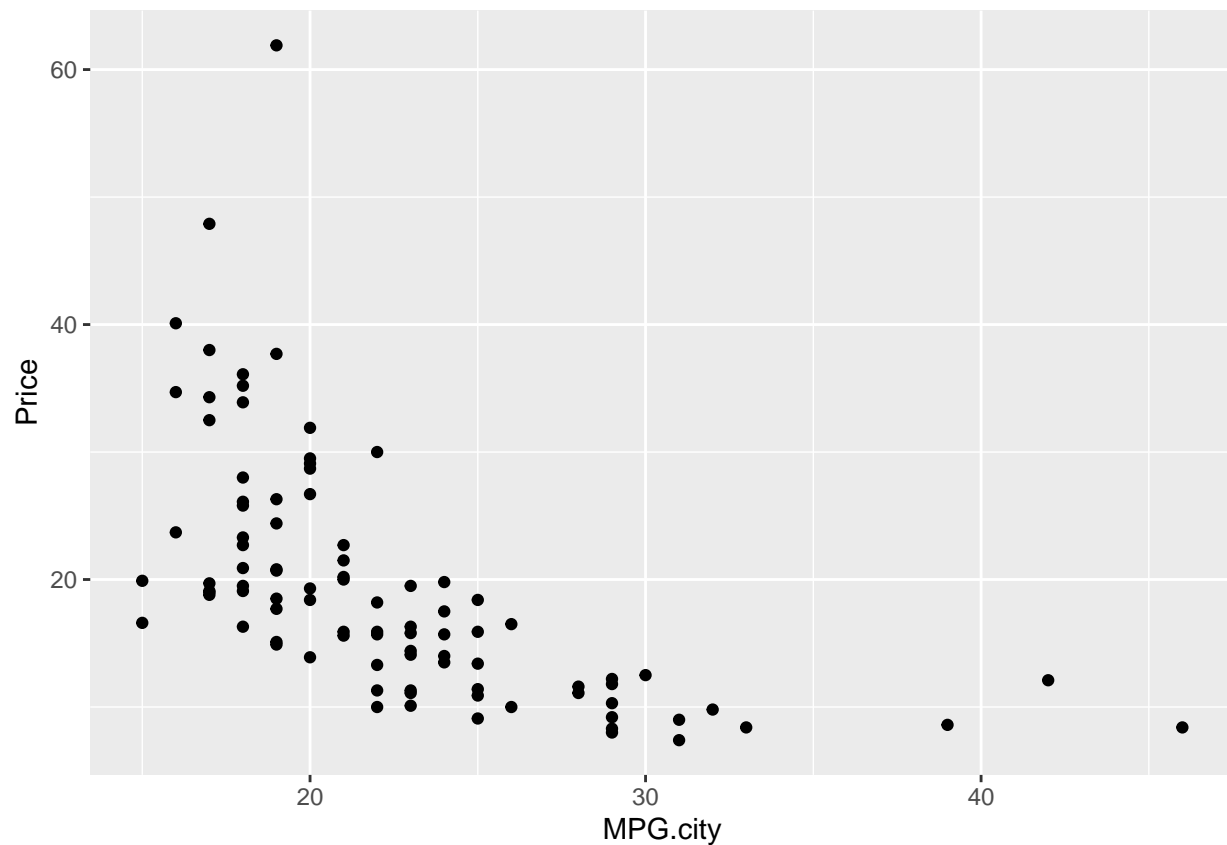
```
cars93$Cylinders
```

```
## [1] "4"      "6"      "6"      "6"      "4"      "4"      "6"      "6"
## [9] "6"      "8"      "8"      "4"      "4"      "6"      "4"      "6"
## [17] "6"      "8"      "8"      "6"      "4"      "6"      "4"      "4"
## [25] "4"      "6"      "4"      "6"      "4"      "6"      "4"      "4"
## [33] "4"      "4"      "4"      "6"      "6"      "8"      "3"      "4"
## [41] "4"      "4"      "4"      "4"      "4"      "4"      "4"      "8"
## [49] "6"      "6"      "6"      "8"      "4"      "4"      "4"      "6"
## [57] "rotary" "4"      "6"      "4"      "6"      "4"      "6"      "4"
## [65] "4"      "6"      "6"      "4"      "4"      "6"      "6"      "4"
## [73] "4"      "4"      "6"      "6"      "6"      "4"      "4"      "3"
## [81] "4"      "4"      "3"      "4"      "4"      "4"      "4"      "4"
## [89] "5"      "4"      "6"      "4"      "5"
```

```
columns93 <- c("MPG.city", "MPG.highway", "EngineSize", "Horsepower", "RPM", "Rev.per.mile", "Fuel.tank
length(columns93)
```

```
## [1] 13
```

```
ggplot(cars93, aes(x = MPG.city, y = Price)) +
  geom_point()
```



```
cars93[]
```

| ## | id | Manufacturer | Model | Type | Min.Price | Price | Max.Price | MPG.city |
|-------|----|--------------|------------|---------|-----------|-------|-----------|----------|
| ## 1 | 1 | Acura | Integra | Small | 12.9 | 15.9 | 18.8 | 25 |
| ## 2 | 2 | Acura | Legend | Midsize | 29.2 | 33.9 | 38.7 | 18 |
| ## 3 | 3 | Audi | 90 | Compact | 25.9 | 29.1 | 32.3 | 20 |
| ## 4 | 4 | Audi | 100 | Midsize | 30.8 | 37.7 | 44.6 | 19 |
| ## 5 | 5 | BMW | 535i | Midsize | 23.7 | 30.0 | 36.2 | 22 |
| ## 6 | 6 | Buick | Century | Midsize | 14.2 | 15.7 | 17.3 | 22 |
| ## 7 | 7 | Buick | LeSabre | Large | 19.9 | 20.8 | 21.7 | 19 |
| ## 8 | 8 | Buick | Roadmaster | Large | 22.6 | 23.7 | 24.9 | 16 |
| ## 9 | 9 | Buick | Riviera | Midsize | 26.3 | 26.3 | 26.3 | 19 |
| ## 10 | 10 | Cadillac | DeVille | Large | 33.0 | 34.7 | 36.3 | 16 |
| ## 11 | 11 | Cadillac | Seville | Midsize | 37.5 | 40.1 | 42.7 | 16 |
| ## 12 | 12 | Chevrolet | Cavalier | Compact | 8.5 | 13.4 | 18.3 | 25 |
| ## 13 | 13 | Chevrolet | Corsica | Compact | 11.4 | 11.4 | 11.4 | 25 |
| ## 14 | 14 | Chevrolet | Camaro | Sporty | 13.4 | 15.1 | 16.8 | 19 |
| ## 15 | 15 | Chevrolet | Lumina | Midsize | 13.4 | 15.9 | 18.4 | 21 |
| ## 16 | 16 | Chevrolet | Lumina_APV | Van | 14.7 | 16.3 | 18.0 | 18 |
| ## 17 | 17 | Chevrolet | Astro | Van | 14.7 | 16.6 | 18.6 | 15 |
| ## 18 | 18 | Chevrolet | Caprice | Large | 18.0 | 18.8 | 19.6 | 17 |
| ## 19 | 19 | Chevrolet | Corvette | Sporty | 34.6 | 38.0 | 41.5 | 17 |
| ## 20 | 20 | Chrysler | Concorde | Large | 18.4 | 18.4 | 18.4 | 20 |
| ## 21 | 21 | Chrysler | LeBaron | Compact | 14.5 | 15.8 | 17.1 | 23 |
| ## 22 | 22 | Chrysler | Imperial | Large | 29.5 | 29.5 | 29.5 | 20 |
| ## 23 | 23 | Dodge | Colt | Small | 7.9 | 9.2 | 10.6 | 29 |

| | | | | | | | | | |
|----|----|----|---------------|----------------|---------|------|------|------|----|
| ## | 24 | 24 | Dodge | Shadow | Small | 8.4 | 11.3 | 14.2 | 23 |
| ## | 25 | 25 | Dodge | Spirit | Compact | 11.9 | 13.3 | 14.7 | 22 |
| ## | 26 | 26 | Dodge | Caravan | Van | 13.6 | 19.0 | 24.4 | 17 |
| ## | 27 | 27 | Dodge | Dynasty | Midsize | 14.8 | 15.6 | 16.4 | 21 |
| ## | 28 | 28 | Dodge | Stealth | Sporty | 18.5 | 25.8 | 33.1 | 18 |
| ## | 29 | 29 | Eagle | Summit | Small | 7.9 | 12.2 | 16.5 | 29 |
| ## | 30 | 30 | Eagle | Vision | Large | 17.5 | 19.3 | 21.2 | 20 |
| ## | 31 | 31 | Ford | Festiva | Small | 6.9 | 7.4 | 7.9 | 31 |
| ## | 32 | 32 | Ford | Escort | Small | 8.4 | 10.1 | 11.9 | 23 |
| ## | 33 | 33 | Ford | Tempo | Compact | 10.4 | 11.3 | 12.2 | 22 |
| ## | 34 | 34 | Ford | Mustang | Sporty | 10.8 | 15.9 | 21.0 | 22 |
| ## | 35 | 35 | Ford | Probe | Sporty | 12.8 | 14.0 | 15.2 | 24 |
| ## | 36 | 36 | Ford | Aerostar | Van | 14.5 | 19.9 | 25.3 | 15 |
| ## | 37 | 37 | Ford | Taurus | Midsize | 15.6 | 20.2 | 24.8 | 21 |
| ## | 38 | 38 | Ford | Crown_Victoria | Large | 20.1 | 20.9 | 21.7 | 18 |
| ## | 39 | 39 | Geo | Metro | Small | 6.7 | 8.4 | 10.0 | 46 |
| ## | 40 | 40 | Geo | Storm | Sporty | 11.5 | 12.5 | 13.5 | 30 |
| ## | 41 | 41 | Honda | Prelude | Sporty | 17.0 | 19.8 | 22.7 | 24 |
| ## | 42 | 42 | Honda | Civic | Small | 8.4 | 12.1 | 15.8 | 42 |
| ## | 43 | 43 | Honda | Accord | Compact | 13.8 | 17.5 | 21.2 | 24 |
| ## | 44 | 44 | Hyundai | Excel | Small | 6.8 | 8.0 | 9.2 | 29 |
| ## | 45 | 45 | Hyundai | Elantra | Small | 9.0 | 10.0 | 11.0 | 22 |
| ## | 46 | 46 | Hyundai | Scoupe | Sporty | 9.1 | 10.0 | 11.0 | 26 |
| ## | 47 | 47 | Hyundai | Sonata | Midsize | 12.4 | 13.9 | 15.3 | 20 |
| ## | 48 | 48 | Infiniti | Q45 | Midsize | 45.4 | 47.9 | 50.4 | 17 |
| ## | 49 | 49 | Lexus | ES300 | Midsize | 27.5 | 28.0 | 28.4 | 18 |
| ## | 50 | 50 | Lexus | SC300 | Midsize | 34.7 | 35.2 | 35.6 | 18 |
| ## | 51 | 51 | Lincoln | Continental | Midsize | 33.3 | 34.3 | 35.3 | 17 |
| ## | 52 | 52 | Lincoln | Town_Car | Large | 34.4 | 36.1 | 37.8 | 18 |
| ## | 53 | 53 | Mazda | 323 | Small | 7.4 | 8.3 | 9.1 | 29 |
| ## | 54 | 54 | Mazda | Protege | Small | 10.9 | 11.6 | 12.3 | 28 |
| ## | 55 | 55 | Mazda | 626 | Compact | 14.3 | 16.5 | 18.7 | 26 |
| ## | 56 | 56 | Mazda | MPV | Van | 16.6 | 19.1 | 21.7 | 18 |
| ## | 57 | 57 | Mazda | RX-7 | Sporty | 32.5 | 32.5 | 32.5 | 17 |
| ## | 58 | 58 | Mercedes-Benz | 190E | Compact | 29.0 | 31.9 | 34.9 | 20 |
| ## | 59 | 59 | Mercedes-Benz | 300E | Midsize | 43.8 | 61.9 | 80.0 | 19 |
| ## | 60 | 60 | Mercury | Capri | Sporty | 13.3 | 14.1 | 15.0 | 23 |
| ## | 61 | 61 | Mercury | Cougar | Midsize | 14.9 | 14.9 | 14.9 | 19 |
| ## | 62 | 62 | Mitsubishi | Mirage | Small | 7.7 | 10.3 | 12.9 | 29 |
| ## | 63 | 63 | Mitsubishi | Diamante | Midsize | 22.4 | 26.1 | 29.9 | 18 |
| ## | 64 | 64 | Nissan | Sentra | Small | 8.7 | 11.8 | 14.9 | 29 |
| ## | 65 | 65 | Nissan | Altima | Compact | 13.0 | 15.7 | 18.3 | 24 |
| ## | 66 | 66 | Nissan | Quest | Van | 16.7 | 19.1 | 21.5 | 17 |
| ## | 67 | 67 | Nissan | Maxima | Midsize | 21.0 | 21.5 | 22.0 | 21 |
| ## | 68 | 68 | Oldsmobile | Achieva | Compact | 13.0 | 13.5 | 14.0 | 24 |
| ## | 69 | 69 | Oldsmobile | Cutlass_Ciera | Midsize | 14.2 | 16.3 | 18.4 | 23 |
| ## | 70 | 70 | Oldsmobile | Silhouette | Van | 19.5 | 19.5 | 19.5 | 18 |
| ## | 71 | 71 | Oldsmobile | Eighty-Eight | Large | 19.5 | 20.7 | 21.9 | 19 |
| ## | 72 | 72 | Plymouth | Laser | Sporty | 11.4 | 14.4 | 17.4 | 23 |
| ## | 73 | 73 | Pontiac | LeMans | Small | 8.2 | 9.0 | 9.9 | 31 |
| ## | 74 | 74 | Pontiac | Sunbird | Compact | 9.4 | 11.1 | 12.8 | 23 |
| ## | 75 | 75 | Pontiac | Firebird | Sporty | 14.0 | 17.7 | 21.4 | 19 |
| ## | 76 | 76 | Pontiac | Grand_Prix | Midsize | 15.4 | 18.5 | 21.6 | 19 |
| ## | 77 | 77 | Pontiac | Bonneville | Large | 19.4 | 24.4 | 29.4 | 19 |

| | | | | | | | |
|----------|-------------|--------------------|------------|-----------|------------|------------|----|
| ## 78 78 | Saab | 900 | Compact | 20.3 | 28.7 | 37.1 | 20 |
| ## 79 79 | Saturn | SL | Small | 9.2 | 11.1 | 12.9 | 28 |
| ## 80 80 | Subaru | Justy | Small | 7.3 | 8.4 | 9.5 | 33 |
| ## 81 81 | Subaru | Loyale | Small | 10.5 | 10.9 | 11.3 | 25 |
| ## 82 82 | Subaru | Legacy | Compact | 16.3 | 19.5 | 22.7 | 23 |
| ## 83 83 | Suzuki | Swift | Small | 7.3 | 8.6 | 10.0 | 39 |
| ## 84 84 | Toyota | Tercel | Small | 7.8 | 9.8 | 11.8 | 32 |
| ## 85 85 | Toyota | Celica | Sporty | 14.2 | 18.4 | 22.6 | 25 |
| ## 86 86 | Toyota | Camry | Midsize | 15.2 | 18.2 | 21.2 | 22 |
| ## 87 87 | Toyota | Previa | Van | 18.9 | 22.7 | 26.6 | 18 |
| ## 88 88 | Volkswagen | Fox | Small | 8.7 | 9.1 | 9.5 | 25 |
| ## 89 89 | Volkswagen | Eurovan | Van | 16.6 | 19.7 | 22.7 | 17 |
| ## 90 90 | Volkswagen | Passat | Compact | 17.6 | 20.0 | 22.4 | 21 |
| ## 91 91 | Volkswagen | Corrado | Sporty | 22.9 | 23.3 | 23.7 | 18 |
| ## 92 92 | Volvo | 240 | Compact | 21.8 | 22.7 | 23.5 | 21 |
| ## 93 93 | Volvo | 850 | Midsize | 24.8 | 26.7 | 28.5 | 20 |
| ## | MPG.highway | AirBags | DriveTrain | Cylinders | EngineSize | Horsepower | |
| ## 1 | 31 | None | Front | 4 | 1.8 | 140 | |
| ## 2 | 25 | Driver & Passenger | Front | 6 | 3.2 | 200 | |
| ## 3 | 26 | Driver only | Front | 6 | 2.8 | 172 | |
| ## 4 | 26 | | Front | 6 | 2.8 | 172 | |
| ## 5 | 30 | Driver only | Rear | 4 | 3.5 | 208 | |
| ## 6 | 31 | Driver only | Front | 4 | 2.2 | 110 | |
| ## 7 | 28 | Driver only | Front | 6 | 3.8 | 170 | |
| ## 8 | 25 | Driver only | Rear | 6 | 5.7 | 180 | |
| ## 9 | 27 | Driver only | Front | 6 | 3.8 | 170 | |
| ## 10 | 25 | Driver only | Front | 8 | 4.9 | 200 | |
| ## 11 | 25 | Driver & Passenger | Front | 8 | 4.6 | 295 | |
| ## 12 | 36 | None | Front | 4 | 2.2 | 110 | |
| ## 13 | 34 | Driver only | Front | 4 | 2.2 | 110 | |
| ## 14 | 28 | Driver & Passenger | Rear | 6 | 3.4 | 160 | |
| ## 15 | 29 | None | Front | 4 | 2.2 | 110 | |
| ## 16 | 23 | None | Front | 6 | 3.8 | 170 | |
| ## 17 | 20 | None | 4WD | 6 | 4.3 | 165 | |
| ## 18 | 26 | Driver only | Rear | 8 | 5.0 | 170 | |
| ## 19 | 25 | Driver only | Rear | 8 | 5.7 | 300 | |
| ## 20 | 28 | Driver & Passenger | Front | 6 | 3.3 | 153 | |
| ## 21 | 28 | Driver & Passenger | Front | 4 | 3.0 | 141 | |
| ## 22 | 26 | | Front | 6 | 3.3 | 147 | |
| ## 23 | 33 | None | Front | 4 | 1.5 | 92 | |
| ## 24 | 29 | Driver only | Front | 4 | 2.2 | 93 | |
| ## 25 | 27 | Driver only | Front | 4 | 2.5 | 100 | |
| ## 26 | 21 | Driver only | 4WD | 6 | 3.0 | 142 | |
| ## 27 | 27 | Driver only | Front | 4 | 2.5 | 100 | |
| ## 28 | 24 | Driver only | 4WD | 6 | 3.0 | 300 | |
| ## 29 | 33 | None | Front | 4 | 1.5 | 92 | |
| ## 30 | 28 | Driver & Passenger | Front | 6 | 3.5 | 214 | |
| ## 31 | 33 | None | Front | 4 | 1.3 | 63 | |
| ## 32 | 30 | None | Front | 4 | 1.8 | 127 | |
| ## 33 | 27 | None | Front | 4 | 2.3 | 96 | |
| ## 34 | 29 | Driver only | Rear | 4 | 2.3 | 105 | |
| ## 35 | 30 | Driver only | Front | 4 | 2.0 | 115 | |
| ## 36 | 20 | Driver only | 4WD | 6 | 3.0 | 145 | |
| ## 37 | 30 | Driver only | Front | 6 | 3.0 | 140 | |

| | | | | | | |
|-------|----|--------------------|-------|--------|-----|-----|
| ## 38 | 26 | Driver only | Rear | 8 | 4.6 | 190 |
| ## 39 | 50 | None | Front | 3 | 1.0 | 55 |
| ## 40 | 36 | Driver only | Front | 4 | 1.6 | 90 |
| ## 41 | 31 | Driver & Passenger | Front | 4 | 2.3 | 160 |
| ## 42 | 46 | Driver only | Front | 4 | 1.5 | 102 |
| ## 43 | 31 | Driver & Passenger | Front | 4 | 2.2 | 140 |
| ## 44 | 33 | None | Front | 4 | 1.5 | 81 |
| ## 45 | 29 | None | Front | 4 | 1.8 | 124 |
| ## 46 | 34 | | Front | 4 | 1.5 | 92 |
| ## 47 | 27 | None | Front | 4 | 2.0 | 128 |
| ## 48 | 22 | Driver only | Rear | 8 | 4.5 | 278 |
| ## 49 | 24 | Driver only | Front | 6 | 3.0 | 185 |
| ## 50 | 23 | Driver & Passenger | Rear | 6 | 3.0 | 225 |
| ## 51 | 26 | Driver & Passenger | Front | 6 | 3.8 | 160 |
| ## 52 | 26 | | Rear | 8 | 4.6 | 210 |
| ## 53 | 37 | None | Front | 4 | 1.6 | 82 |
| ## 54 | 36 | None | Front | 4 | 1.8 | 103 |
| ## 55 | 34 | Driver only | Front | 4 | 2.5 | 164 |
| ## 56 | 24 | None | 4WD | 6 | 3.0 | 155 |
| ## 57 | 25 | Driver only | Rear | rotary | 1.3 | 255 |
| ## 58 | 29 | Driver only | Rear | 4 | 2.3 | 130 |
| ## 59 | 25 | Driver & Passenger | Rear | 6 | 3.2 | 217 |
| ## 60 | 26 | Driver only | Front | 4 | 1.6 | 100 |
| ## 61 | 26 | None | Rear | 6 | 3.8 | 140 |
| ## 62 | 33 | None | Front | 4 | 1.5 | 92 |
| ## 63 | 24 | Driver only | Front | 6 | 3.0 | 202 |
| ## 64 | 33 | Driver only | Front | 4 | 1.6 | 110 |
| ## 65 | 30 | Driver only | Front | 4 | 2.4 | 150 |
| ## 66 | 23 | None | Front | 6 | 3.0 | 151 |
| ## 67 | 26 | Driver only | Front | 6 | 3.0 | 160 |
| ## 68 | 31 | None | Front | 4 | 2.3 | 155 |
| ## 69 | 31 | Driver only | Front | 4 | 2.2 | 110 |
| ## 70 | 23 | None | Front | 6 | 3.8 | 170 |
| ## 71 | 28 | Driver only | Front | 6 | 3.8 | 170 |
| ## 72 | 30 | None | 4WD | 4 | 1.8 | 92 |
| ## 73 | 41 | None | Front | 4 | 1.6 | 74 |
| ## 74 | 31 | None | Front | 4 | 2.0 | 110 |
| ## 75 | 28 | Driver & Passenger | Rear | 6 | 3.4 | 160 |
| ## 76 | 27 | None | Front | 6 | 3.4 | 200 |
| ## 77 | 28 | Driver & Passenger | Front | 6 | 3.8 | 170 |
| ## 78 | 26 | Driver only | Front | 4 | 2.1 | 140 |
| ## 79 | 38 | Driver only | Front | 4 | 1.9 | 85 |
| ## 80 | 37 | None | 4WD | 3 | 1.2 | 73 |
| ## 81 | 30 | None | 4WD | 4 | 1.8 | 90 |
| ## 82 | 30 | Driver only | 4WD | 4 | 2.2 | 130 |
| ## 83 | 43 | None | Front | 3 | 1.3 | 70 |
| ## 84 | 37 | Driver only | Front | 4 | 1.5 | 82 |
| ## 85 | 32 | Driver only | Front | 4 | 2.2 | 135 |
| ## 86 | 29 | Driver only | Front | 4 | 2.2 | 130 |
| ## 87 | 22 | Driver only | 4WD | 4 | 2.4 | 138 |
| ## 88 | 33 | None | Front | 4 | 1.8 | 81 |
| ## 89 | 21 | None | Front | 5 | 2.5 | 109 |
| ## 90 | 30 | None | Front | 4 | 2.0 | 134 |
| ## 91 | 25 | None | Front | 6 | 2.8 | 178 |

| | | | | | | |
|-------|------|--------------------|-----------------|--------------------|------------|--------|
| ## 92 | 28 | Driver only | Rear | 4 | 2.3 | 114 |
| ## 93 | 28 | Driver & Passenger | Front | 5 | 2.4 | 168 |
| ## | RPM | Rev.per.mile | Man.trans.avail | Fuel.tank.capacity | Passengers | Length |
| ## 1 | 6300 | 2890 | Yes | 13.2 | 5 | 177 |
| ## 2 | 5500 | 2335 | Yes | 18.0 | 5 | 195 |
| ## 3 | 5500 | 2280 | Yes | 16.9 | 5 | 180 |
| ## 4 | 5500 | 2535 | Yes | 21.1 | 6 | 193 |
| ## 5 | 5700 | 2545 | Yes | 21.1 | 4 | 186 |
| ## 6 | 5200 | 2565 | No | 16.4 | 6 | 189 |
| ## 7 | 4800 | 1570 | No | 18.0 | 6 | 200 |
| ## 8 | 4000 | 1320 | No | 23.0 | 6 | 216 |
| ## 9 | 4800 | 1690 | No | 18.8 | 5 | 198 |
| ## 10 | 4100 | 1510 | No | 18.0 | 6 | 206 |
| ## 11 | 6000 | 1985 | No | 20.0 | 5 | 204 |
| ## 12 | 5200 | 2380 | Yes | 15.2 | 5 | 182 |
| ## 13 | 5200 | 2665 | Yes | 15.6 | 5 | 184 |
| ## 14 | 4600 | 1805 | Yes | 15.5 | 4 | 193 |
| ## 15 | 5200 | 2595 | No | 16.5 | 6 | 198 |
| ## 16 | 4800 | 1690 | No | 20.0 | 7 | 178 |
| ## 17 | 4000 | 1790 | No | 27.0 | 8 | 194 |
| ## 18 | 4200 | 1350 | No | 23.0 | 6 | 214 |
| ## 19 | 5000 | 1450 | Yes | 20.0 | 2 | 179 |
| ## 20 | 5300 | 1990 | No | 18.0 | 6 | 203 |
| ## 21 | 5000 | 2090 | No | 16.0 | 6 | 183 |
| ## 22 | 4800 | 1785 | No | 16.0 | 6 | 203 |
| ## 23 | 6000 | 3285 | Yes | 13.2 | 5 | 174 |
| ## 24 | 4800 | 2595 | Yes | 14.0 | 5 | 172 |
| ## 25 | 4800 | 2535 | Yes | 16.0 | 6 | 181 |
| ## 26 | 5000 | 1970 | No | 20.0 | 7 | 175 |
| ## 27 | 4800 | 2465 | No | 16.0 | 6 | 192 |
| ## 28 | 6000 | 2120 | Yes | 19.8 | 4 | 180 |
| ## 29 | 6000 | 2505 | Yes | 13.2 | 5 | 174 |
| ## 30 | 5800 | 1980 | No | 18.0 | 6 | 202 |
| ## 31 | 5000 | 3150 | Yes | 10.0 | 4 | 141 |
| ## 32 | 6500 | 2410 | Yes | 13.2 | 5 | 171 |
| ## 33 | 4200 | 2805 | Yes | 15.9 | 5 | 177 |
| ## 34 | 4600 | 2285 | Yes | 15.4 | 4 | 180 |
| ## 35 | 5500 | 2340 | Yes | 15.5 | 4 | 179 |
| ## 36 | 4800 | 2080 | Yes | 21.0 | 7 | 176 |
| ## 37 | 4800 | 1885 | No | 16.0 | 5 | 192 |
| ## 38 | 4200 | 1415 | No | 20.0 | 6 | 212 |
| ## 39 | 5700 | 3755 | Yes | 10.6 | 4 | 151 |
| ## 40 | 5400 | 3250 | Yes | 12.4 | 4 | 164 |
| ## 41 | 5800 | 2855 | Yes | 15.9 | 4 | 175 |
| ## 42 | 5900 | 2650 | Yes | 11.9 | 4 | 173 |
| ## 43 | 5600 | 2610 | Yes | 17.0 | 4 | 185 |
| ## 44 | 5500 | 2710 | Yes | 11.9 | 5 | 168 |
| ## 45 | 6000 | 2745 | Yes | 13.7 | 5 | 172 |
| ## 46 | 5550 | 2540 | Yes | 11.9 | 4 | 166 |
| ## 47 | 6000 | 2335 | Yes | 17.2 | 5 | 184 |
| ## 48 | 6000 | 1955 | No | 22.5 | 5 | 200 |
| ## 49 | 5200 | 2325 | Yes | 18.5 | 5 | 188 |
| ## 50 | 6000 | 2510 | Yes | 20.6 | 4 | 191 |
| ## 51 | 4400 | 1835 | No | 18.4 | 6 | 205 |

| | | | | | | | |
|-------|-----------|-------|-------------|----------------|--------------|--------|---------|
| ## 52 | 4600 | 1840 | No | 20.0 | 6 | 219 | |
| ## 53 | 5000 | 2370 | Yes | 13.2 | 4 | 164 | |
| ## 54 | 5500 | 2220 | Yes | 14.5 | 5 | 172 | |
| ## 55 | 5600 | 2505 | Yes | 15.5 | 5 | 184 | |
| ## 56 | 5000 | 2240 | No | 19.6 | 7 | 190 | |
| ## 57 | 6500 | 2325 | Yes | 20.0 | 2 | 169 | |
| ## 58 | 5100 | 2425 | Yes | 14.5 | 5 | 175 | |
| ## 59 | 5500 | 2220 | No | 18.5 | 5 | 187 | |
| ## 60 | 5750 | 2475 | Yes | 11.1 | 4 | 166 | |
| ## 61 | 3800 | 1730 | No | 18.0 | 5 | 199 | |
| ## 62 | 6000 | 2505 | Yes | 13.2 | 5 | 172 | |
| ## 63 | 6000 | 2210 | No | 19.0 | 5 | 190 | |
| ## 64 | 6000 | 2435 | Yes | 13.2 | 5 | 170 | |
| ## 65 | 5600 | 2130 | Yes | 15.9 | 5 | 181 | |
| ## 66 | 4800 | 2065 | No | 20.0 | 7 | 190 | |
| ## 67 | 5200 | 2045 | No | 18.5 | 5 | 188 | |
| ## 68 | 6000 | 2380 | No | 15.2 | 5 | 188 | |
| ## 69 | 5200 | 2565 | No | 16.5 | 5 | 190 | |
| ## 70 | 4800 | 1690 | No | 20.0 | 7 | 194 | |
| ## 71 | 4800 | 1570 | No | 18.0 | 6 | 201 | |
| ## 72 | 5000 | 2360 | Yes | 15.9 | 4 | 173 | |
| ## 73 | 5600 | 3130 | Yes | 13.2 | 4 | 177 | |
| ## 74 | 5200 | 2665 | Yes | 15.2 | 5 | 181 | |
| ## 75 | 4600 | 1805 | Yes | 15.5 | 4 | 196 | |
| ## 76 | 5000 | 1890 | Yes | 16.5 | 5 | 195 | |
| ## 77 | 4800 | 1565 | No | 18.0 | 6 | 177 | |
| ## 78 | 6000 | 2910 | Yes | 18.0 | 5 | 184 | |
| ## 79 | 5000 | 2145 | Yes | 12.8 | 5 | 176 | |
| ## 80 | 5600 | 2875 | Yes | 9.2 | 4 | 146 | |
| ## 81 | 5200 | 3375 | Yes | 15.9 | 5 | 175 | |
| ## 82 | 5600 | 2330 | Yes | 15.9 | 5 | 179 | |
| ## 83 | 6000 | 3360 | Yes | 10.6 | 4 | 161 | |
| ## 84 | 5200 | 3505 | Yes | 11.9 | 5 | 162 | |
| ## 85 | 5400 | 2405 | Yes | 15.9 | 4 | 174 | |
| ## 86 | 5400 | 2340 | Yes | 18.5 | 5 | 188 | |
| ## 87 | 5000 | 2515 | Yes | 19.8 | 7 | 187 | |
| ## 88 | 5500 | 2550 | Yes | 12.4 | 4 | 163 | |
| ## 89 | 4500 | 2915 | Yes | 21.1 | 7 | 187 | |
| ## 90 | 5800 | 2685 | Yes | 18.5 | 5 | 180 | |
| ## 91 | 5800 | 2385 | Yes | 18.5 | 4 | 159 | |
| ## 92 | 5400 | 2215 | Yes | 15.8 | 5 | 190 | |
| ## 93 | 6200 | 2310 | Yes | 19.3 | 5 | 184 | |
| ## | Wheelbase | Width | Turn.circle | Rear.seat.room | Luggage.room | Weight | Origin |
| ## 1 | 102 | 68 | 37 | 26.5 | 11 | 2705 | non-USA |
| ## 2 | 115 | 71 | 38 | 30.0 | 15 | 3560 | non-USA |
| ## 3 | 102 | 67 | 37 | 28.0 | 14 | 3375 | non-USA |
| ## 4 | 106 | 70 | 37 | 31.0 | 17 | 3405 | non-USA |
| ## 5 | 109 | 69 | 39 | 27.0 | 13 | 3640 | non-USA |
| ## 6 | 105 | 69 | 41 | 28.0 | 16 | 2880 | USA |
| ## 7 | 111 | 74 | 42 | 30.5 | 17 | 3470 | USA |
| ## 8 | 116 | 78 | 45 | 30.5 | 21 | 4105 | USA |
| ## 9 | 108 | 73 | 41 | 26.5 | 14 | 3495 | USA |
| ## 10 | 114 | 73 | 43 | 35.0 | 18 | 3620 | USA |
| ## 11 | 111 | 74 | 44 | 31.0 | 14 | 3935 | USA |

| | | | | | | | |
|-------|-----|----|----|------|----|------|---------|
| ## 12 | 101 | 66 | 38 | 25.0 | 13 | 2490 | USA |
| ## 13 | 103 | 68 | 39 | 26.0 | 14 | 2785 | USA |
| ## 14 | 101 | 74 | 43 | 25.0 | 13 | 3240 | USA |
| ## 15 | 108 | 71 | 40 | 28.5 | 16 | 3195 | USA |
| ## 16 | 110 | 74 | 44 | 30.5 | NA | 3715 | USA |
| ## 17 | 111 | 78 | 42 | 33.5 | NA | 4025 | USA |
| ## 18 | 116 | 77 | 42 | 29.5 | 20 | 3910 | USA |
| ## 19 | 96 | 74 | 43 | NA | NA | 3380 | USA |
| ## 20 | 113 | 74 | 40 | 31.0 | 15 | 3515 | USA |
| ## 21 | 104 | 68 | 41 | 30.5 | 14 | 3085 | USA |
| ## 22 | 110 | 69 | 44 | 36.0 | 17 | 3570 | USA |
| ## 23 | 98 | 66 | 32 | 26.5 | 11 | 2270 | USA |
| ## 24 | 97 | 67 | 38 | 26.5 | 13 | 2670 | USA |
| ## 25 | 104 | 68 | 39 | 30.5 | 14 | 2970 | USA |
| ## 26 | 112 | 72 | 42 | 26.5 | NA | 3705 | USA |
| ## 27 | 105 | 69 | 42 | 30.5 | 16 | 3080 | USA |
| ## 28 | 97 | 72 | 40 | 20.0 | 11 | 3805 | USA |
| ## 29 | 98 | 66 | 36 | 26.5 | 11 | 2295 | USA |
| ## 30 | 113 | 74 | 40 | 30.0 | 15 | 3490 | USA |
| ## 31 | 90 | 63 | 33 | 26.0 | 12 | 1845 | USA |
| ## 32 | 98 | 67 | 36 | 28.0 | 12 | 2530 | USA |
| ## 33 | 100 | 68 | 39 | 27.5 | 13 | 2690 | USA |
| ## 34 | 101 | 68 | 40 | 24.0 | 12 | 2850 | USA |
| ## 35 | 103 | 70 | 38 | 23.0 | 18 | 2710 | USA |
| ## 36 | 119 | 72 | 45 | 30.0 | NA | 3735 | USA |
| ## 37 | 106 | 71 | 40 | 27.5 | 18 | 3325 | USA |
| ## 38 | 114 | 78 | 43 | 30.0 | 21 | 3950 | USA |
| ## 39 | 93 | 63 | 34 | 27.5 | 10 | 1695 | non-USA |
| ## 40 | 97 | 67 | 37 | 24.5 | 11 | 2475 | non-USA |
| ## 41 | 100 | 70 | 39 | 23.5 | 8 | 2865 | non-USA |
| ## 42 | 103 | 67 | 36 | 28.0 | 12 | 2350 | non-USA |
| ## 43 | 107 | 67 | 41 | 28.0 | 14 | 3040 | non-USA |
| ## 44 | 94 | 63 | 35 | 26.0 | 11 | 2345 | non-USA |
| ## 45 | 98 | 66 | 36 | 28.0 | 12 | 2620 | non-USA |
| ## 46 | 94 | 64 | 34 | 23.5 | 9 | 2285 | non-USA |
| ## 47 | 104 | 69 | 41 | 31.0 | 14 | 2885 | non-USA |
| ## 48 | 113 | 72 | 42 | 29.0 | 15 | 4000 | non-USA |
| ## 49 | 103 | 70 | 40 | 27.5 | 14 | 3510 | non-USA |
| ## 50 | 106 | 71 | 39 | 25.0 | 9 | 3515 | non-USA |
| ## 51 | 109 | 73 | 42 | 30.0 | 19 | 3695 | USA |
| ## 52 | 117 | 77 | 45 | 31.5 | 22 | 4055 | USA |
| ## 53 | 97 | 66 | 34 | 27.0 | 16 | 2325 | non-USA |
| ## 54 | 98 | 66 | 36 | 26.5 | 13 | 2440 | non-USA |
| ## 55 | 103 | 69 | 40 | 29.5 | 14 | 2970 | non-USA |
| ## 56 | 110 | 72 | 39 | 27.5 | NA | 3735 | non-USA |
| ## 57 | 96 | 69 | 37 | NA | NA | 2895 | non-USA |
| ## 58 | 105 | 67 | 34 | 26.0 | 12 | 2920 | non-USA |
| ## 59 | 110 | 69 | 37 | 27.0 | 15 | 3525 | non-USA |
| ## 60 | 95 | 65 | 36 | 19.0 | 6 | 2450 | USA |
| ## 61 | 113 | 73 | 38 | 28.0 | 15 | 3610 | USA |
| ## 62 | 98 | 67 | 36 | 26.0 | 11 | 2295 | non-USA |
| ## 63 | 107 | 70 | 43 | 27.5 | 14 | 3730 | non-USA |
| ## 64 | 96 | 66 | 33 | 26.0 | 12 | 2545 | non-USA |
| ## 65 | 103 | 67 | 40 | 28.5 | 14 | 3050 | non-USA |

| | | | | | | | |
|-------|-----|----|----------------------|------|----|------|---------|
| ## 66 | 112 | 74 | 41 | 27.0 | NA | 4100 | non-USA |
| ## 67 | 104 | 69 | 41 | 28.5 | 14 | 3200 | non-USA |
| ## 68 | 103 | 67 | 39 | 28.0 | 14 | 2910 | USA |
| ## 69 | 105 | 70 | 42 | 28.0 | 16 | 2890 | USA |
| ## 70 | 110 | 74 | 44 | 30.5 | NA | 3715 | USA |
| ## 71 | 111 | 74 | 42 | 31.5 | 17 | 3470 | USA |
| ## 72 | 97 | 67 | 39 | 24.5 | 8 | 2640 | USA |
| ## 73 | 99 | 66 | 35 | 25.5 | 17 | 2350 | USA |
| ## 74 | 101 | 66 | 39 | 25.0 | 13 | 2575 | USA |
| ## 75 | 101 | 75 | 43 | 25.0 | 13 | 3240 | USA |
| ## 76 | 108 | 72 | 41 | 28.5 | 16 | 3450 | USA |
| ## 77 | 111 | 74 | 43 | 30.5 | 18 | 3495 | USA |
| ## 78 | 99 | 67 | 37 | 26.5 | 14 | 2775 | non-USA |
| ## 79 | 102 | 68 | 40 | 26.5 | 12 | 2495 | USA |
| ## 80 | 90 | 60 | 32 | 23.5 | 10 | 2045 | non-USA |
| ## 81 | 97 | 65 | 35 | 27.5 | 15 | 2490 | non-USA |
| ## 82 | 102 | 67 | 37 | 27.0 | 14 | 3085 | non-USA |
| ## 83 | 93 | 63 | 34 | 27.5 | 10 | 1965 | non-USA |
| ## 84 | 94 | 65 | 36 | 24.0 | 11 | 2055 | non-USA |
| ## 85 | 99 | 69 | 39 | 23.0 | 13 | 2950 | non-USA |
| ## 86 | 103 | 70 | 38 | 28.5 | 15 | 3030 | non-USA |
| ## 87 | 113 | 71 | 41 | 35.0 | NA | 3785 | non-USA |
| ## 88 | 93 | 63 | 34 | 26.0 | 10 | 2240 | non-USA |
| ## 89 | 115 | 72 | 38 | 34.0 | NA | 3960 | non-USA |
| ## 90 | 103 | 67 | 35 | 31.5 | 14 | 2985 | non-USA |
| ## 91 | 97 | 66 | 36 | 26.0 | 15 | 2810 | non-USA |
| ## 92 | 104 | 67 | 37 | 29.5 | 14 | 2985 | non-USA |
| ## 93 | 105 | 69 | 38 | 30.0 | 15 | 3245 | non-USA |
| ## | | | Make | | | | |
| ## 1 | | | Acura Integra | | | | |
| ## 2 | | | Acura Legend | | | | |
| ## 3 | | | Audi 90 | | | | |
| ## 4 | | | Audi 100 | | | | |
| ## 5 | | | BMW 535i | | | | |
| ## 6 | | | Buick Century | | | | |
| ## 7 | | | Buick LeSabre | | | | |
| ## 8 | | | Buick Roadmaster | | | | |
| ## 9 | | | Buick Riviera | | | | |
| ## 10 | | | Cadillac DeVille | | | | |
| ## 11 | | | Cadillac Seville | | | | |
| ## 12 | | | Chevrolet Cavalier | | | | |
| ## 13 | | | Chevrolet Corsica | | | | |
| ## 14 | | | Chevrolet Camaro | | | | |
| ## 15 | | | Chevrolet Lumina | | | | |
| ## 16 | | | Chevrolet Lumina_APV | | | | |
| ## 17 | | | Chevrolet Astro | | | | |
| ## 18 | | | Chevrolet Caprice | | | | |
| ## 19 | | | Chevrolet Corvette | | | | |
| ## 20 | | | Chrysler Concorde | | | | |
| ## 21 | | | Chrysler LeBaron | | | | |
| ## 22 | | | Chrysler Imperial | | | | |
| ## 23 | | | Dodge Colt | | | | |
| ## 24 | | | Dodge Shadow | | | | |
| ## 25 | | | Dodge Spirit | | | | |

| | |
|-------|--------------------------|
| ## 26 | Dodge Caravan |
| ## 27 | Dodge Dynasty |
| ## 28 | Dodge Stealth |
| ## 29 | Eagle Summit |
| ## 30 | Eagle Vision |
| ## 31 | Ford Festiva |
| ## 32 | Ford Escort |
| ## 33 | Ford Tempo |
| ## 34 | Ford Mustang |
| ## 35 | Ford Probe |
| ## 36 | Ford Aerostar |
| ## 37 | Ford Taurus |
| ## 38 | Ford Crown_Victoria |
| ## 39 | Geo Metro |
| ## 40 | Geo Storm |
| ## 41 | Honda Prelude |
| ## 42 | Honda Civic |
| ## 43 | Honda Accord |
| ## 44 | Hyundai Excel |
| ## 45 | Hyundai Elantra |
| ## 46 | Hyundai Scoupe |
| ## 47 | Hyundai Sonata |
| ## 48 | Infiniti Q45 |
| ## 49 | Lexus ES300 |
| ## 50 | Lexus SC300 |
| ## 51 | Lincoln Continental |
| ## 52 | Lincoln Town_Car |
| ## 53 | Mazda 323 |
| ## 54 | Mazda Protege |
| ## 55 | Mazda 626 |
| ## 56 | Mazda MPV |
| ## 57 | Mazda RX-7 |
| ## 58 | Mercedes-Benz 190E |
| ## 59 | Mercedes-Benz 300E |
| ## 60 | Mercury Capri |
| ## 61 | Mercury Cougar |
| ## 62 | Mitsubishi Mirage |
| ## 63 | Mitsubishi Diamante |
| ## 64 | Nissan Sentra |
| ## 65 | Nissan Altima |
| ## 66 | Nissan Quest |
| ## 67 | Nissan Maxima |
| ## 68 | Oldsmobile Achieva |
| ## 69 | Oldsmobile Cutlass_Ciera |
| ## 70 | Oldsmobile Silhouette |
| ## 71 | Oldsmobile Eighty-Eight |
| ## 72 | Plymouth Laser |
| ## 73 | Pontiac LeMans |
| ## 74 | Pontiac Sunbird |
| ## 75 | Pontiac Firebird |
| ## 76 | Pontiac Grand_Prix |
| ## 77 | Pontiac Bonneville |
| ## 78 | Saab 900 |
| ## 79 | Saturn SL |

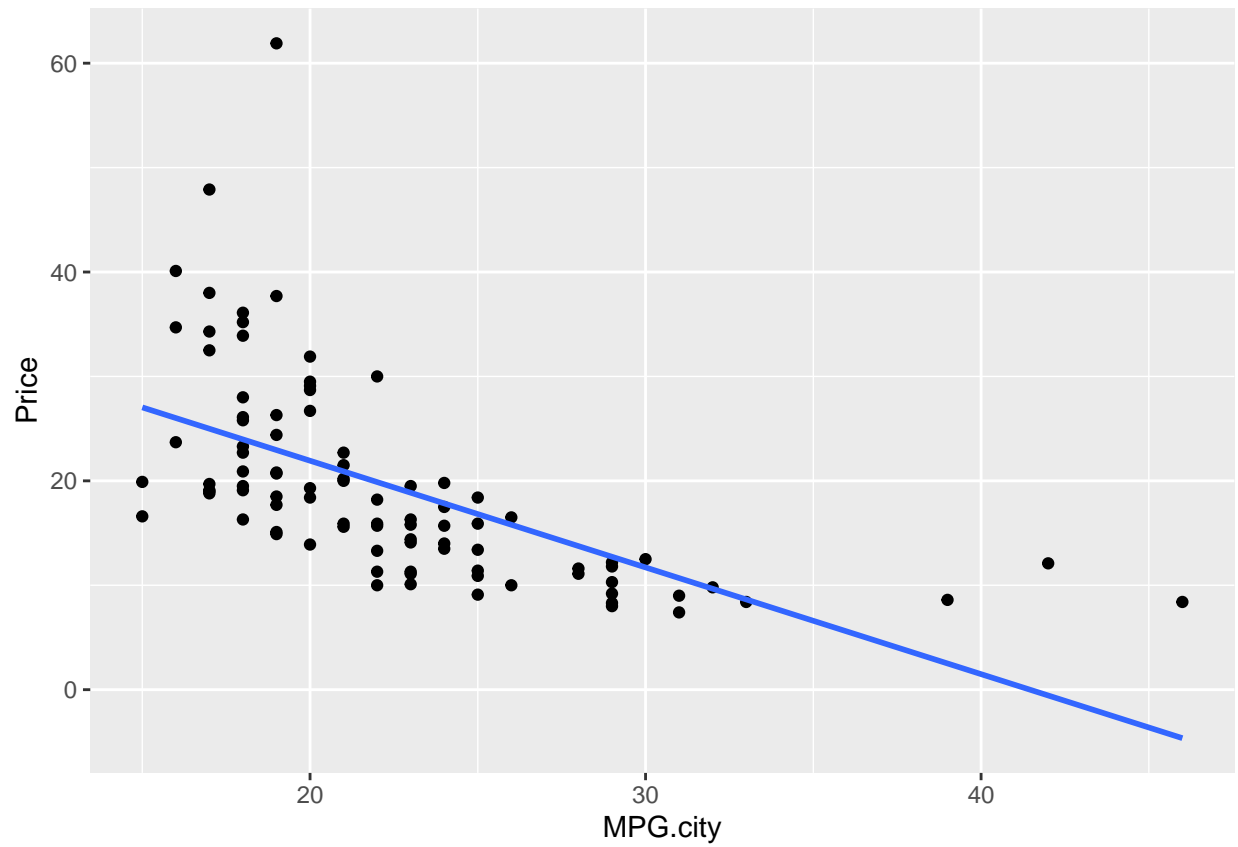
```
## 80          Subaru Justy
## 81          Subaru Loyale
## 82          Subaru Legacy
## 83          Suzuki Swift
## 84          Toyota Tercel
## 85          Toyota Celica
## 86          Toyota Camry
## 87          Toyota Previa
## 88          Volkswagen Fox
## 89          Volkswagen Eurovan
## 90          Volkswagen Passat
## 91          Volkswagen Corrado
## 92          Volvo 240
## 93          Volvo 850
```

```
for (var in columns93)
{
  print(var);
  p <- ggplot(cars93, aes_string(x = var, y = "Price")) +
  geom_point() +
  geom_smooth(method = "lm", se = FALSE);
  print(p);
  correlation <- cor(cars93$Price, cars93[[var]]);
  print(correlation)
}
```

```
## [1] "MPG.city"
```

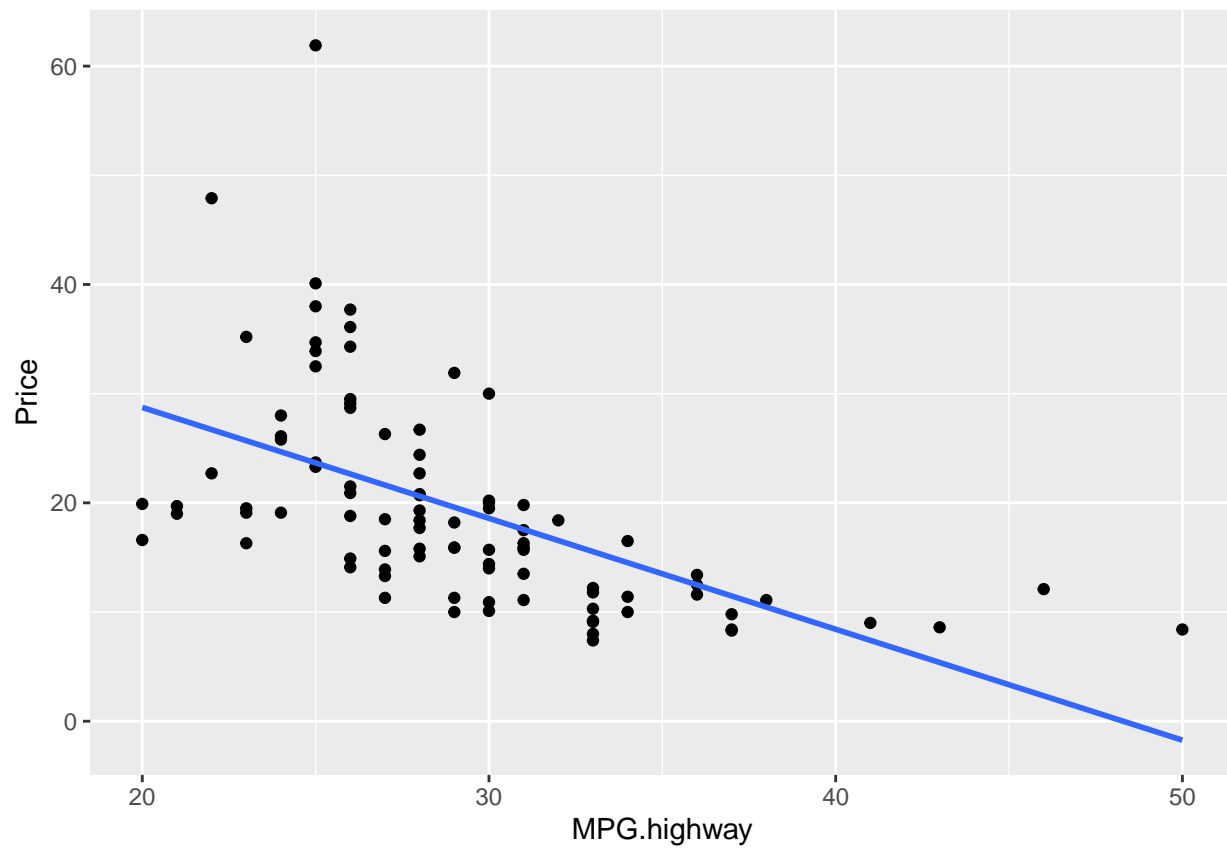
```
## Warning: 'aes_string()' was deprecated in ggplot2 3.0.0.
## i Please use tidy evaluation idioms with 'aes()'.
## i See also 'vignette("ggplot2-in-packages")' for more information.
## This warning is displayed once every 8 hours.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was
## generated.
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```



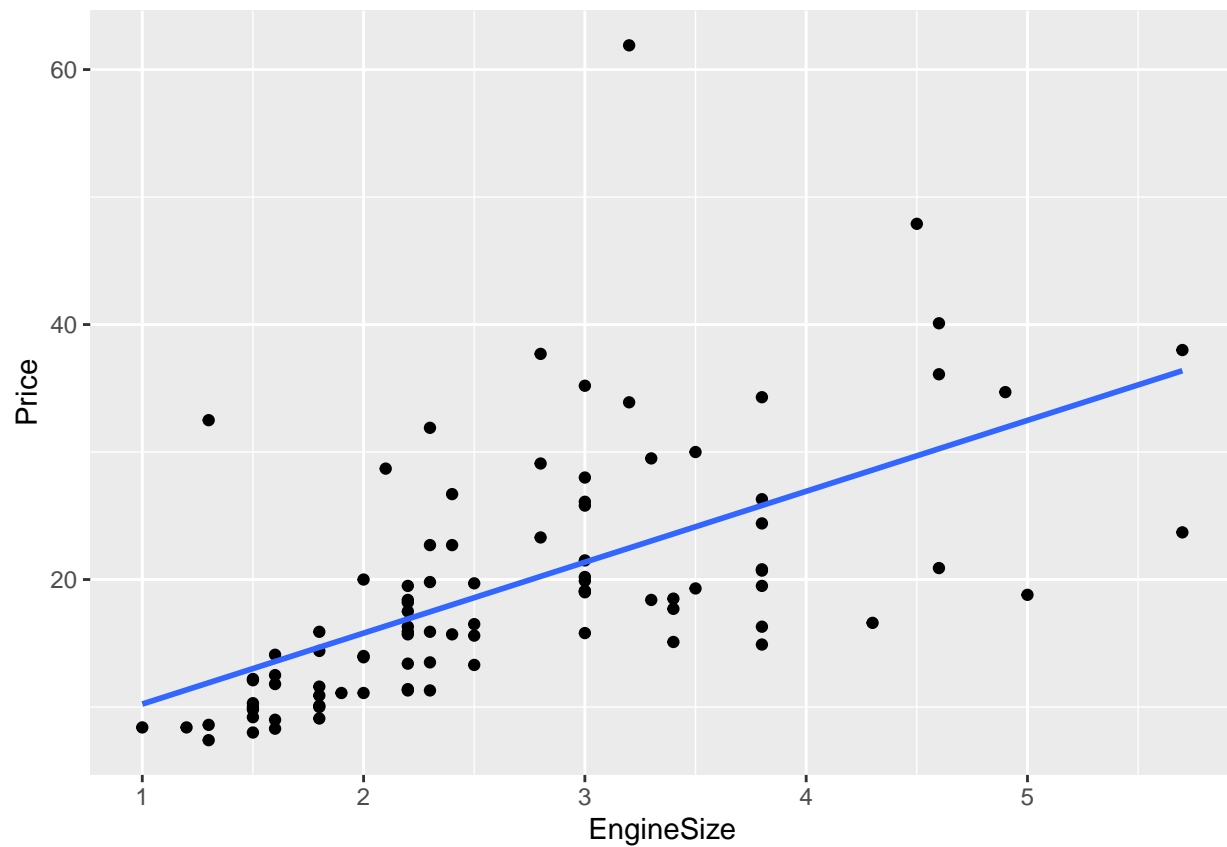
```
## [1] -0.5945622
## [1] "MPG.highway"

## 'geom_smooth()' using formula = 'y ~ x'
```



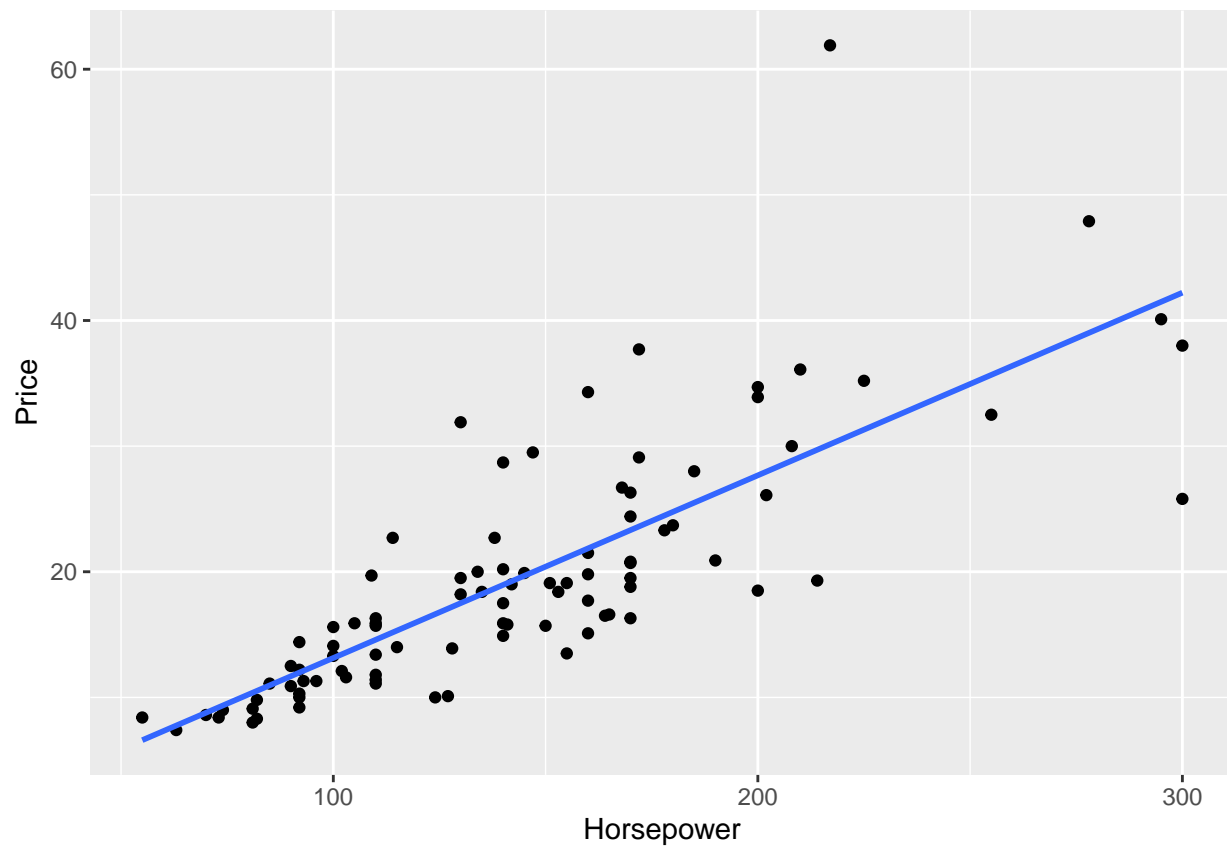
```
## [1] -0.5606804  
## [1] "EngineSize"
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```

```
## [1] 0.5974254  
## [1] "Horsepower"
```

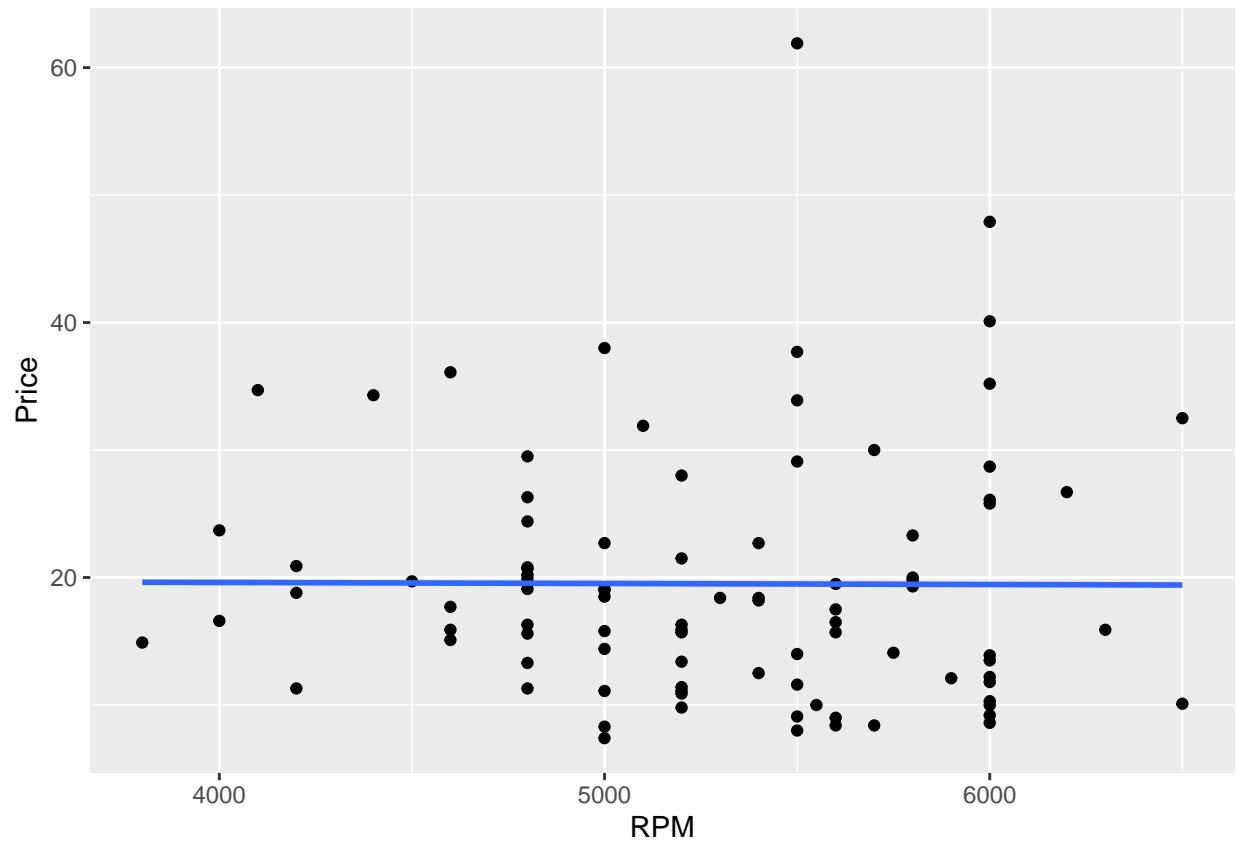
```
## 'geom_smooth()' using formula = 'y ~ x'
```



```
## [1] 0.7882176
```

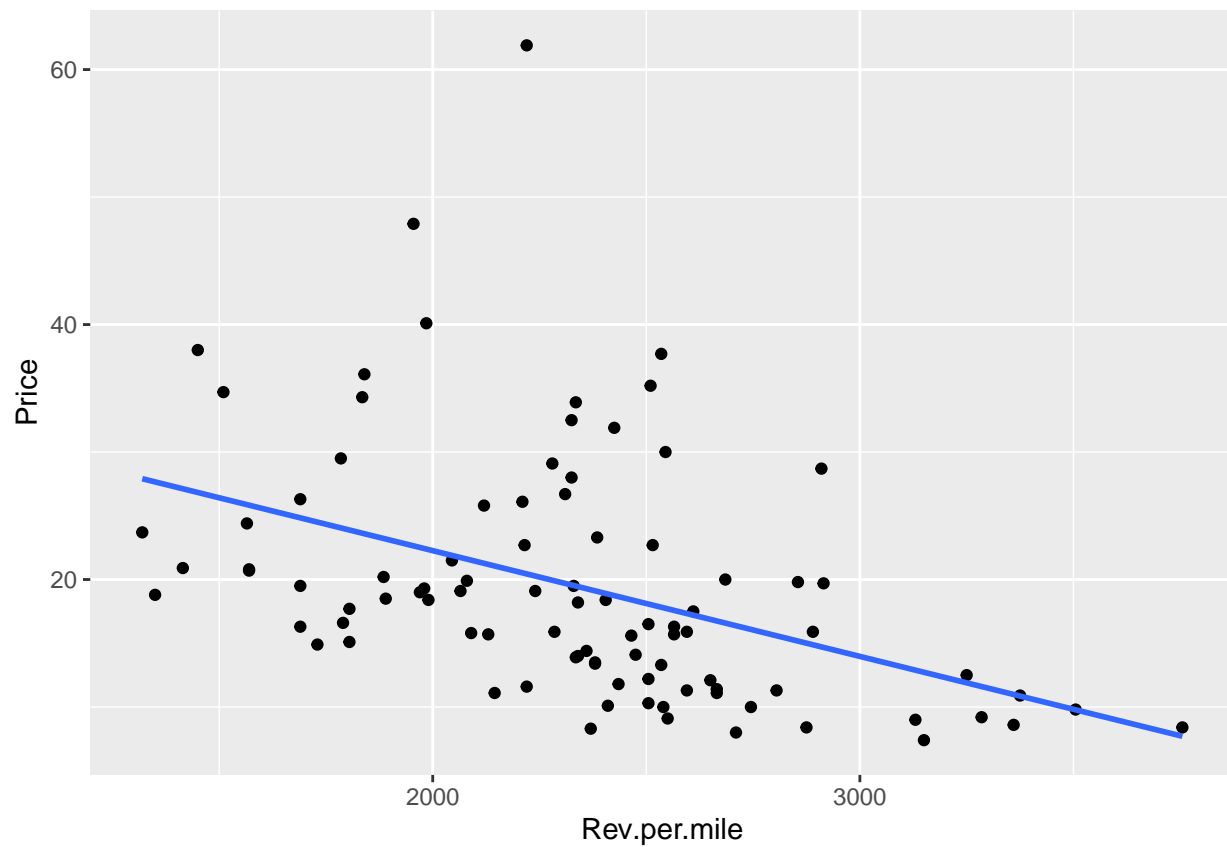
```
## [1] "RPM"
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```



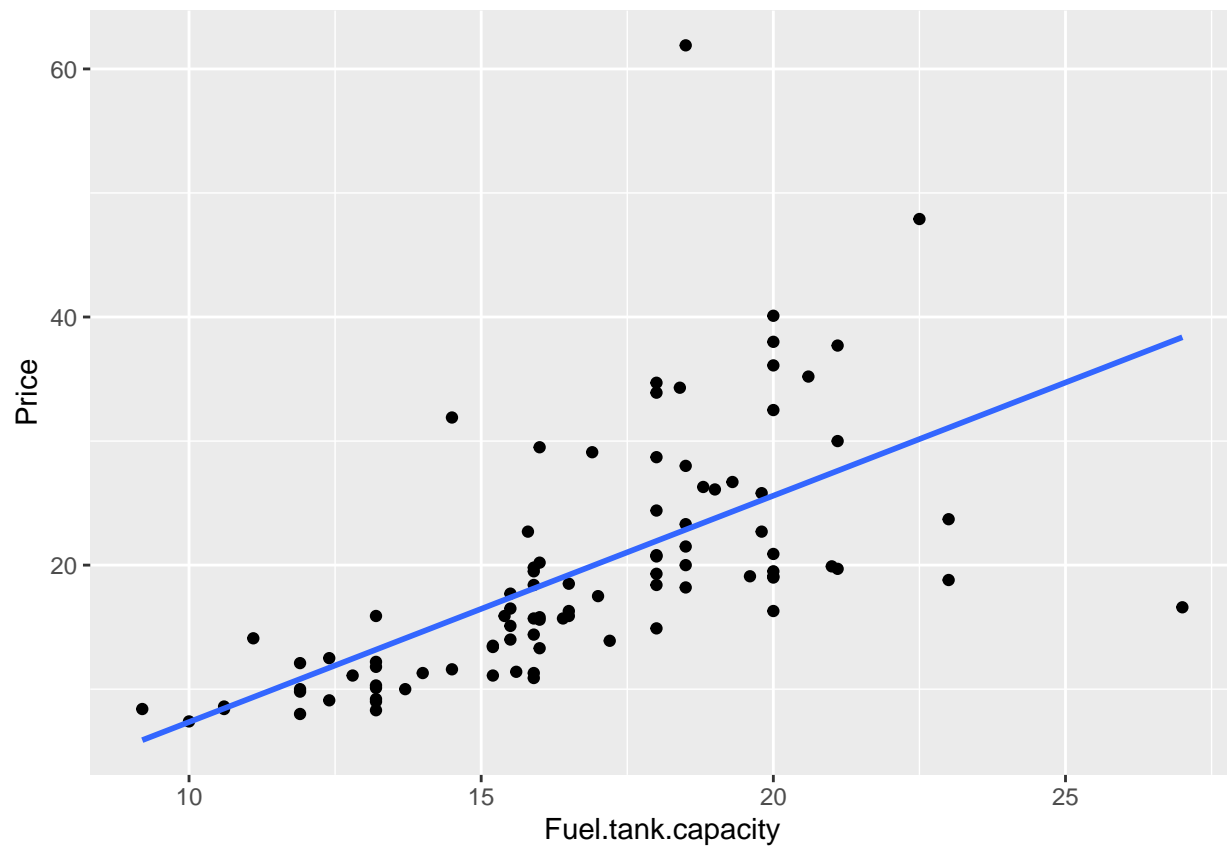
```
## [1] -0.004954931  
## [1] "Rev.per.mile"
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```



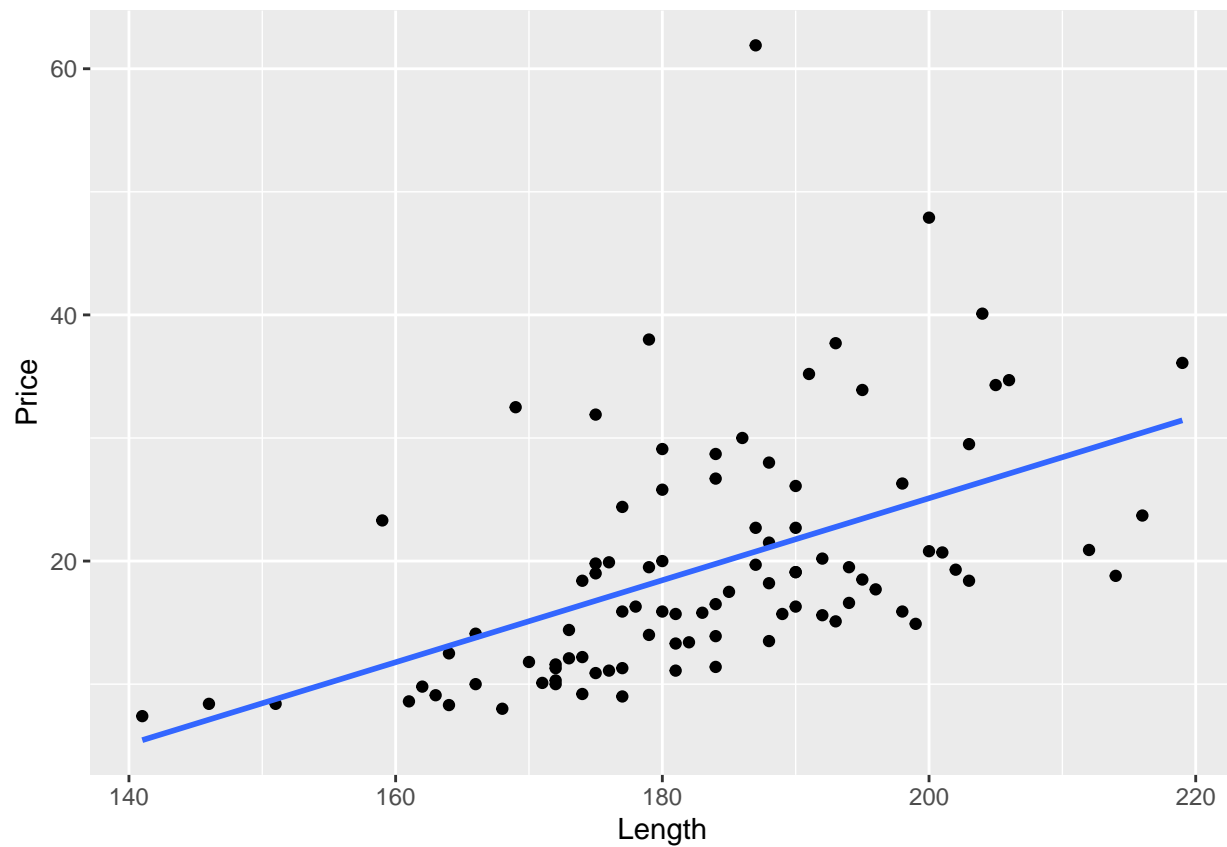
```
## [1] -0.4263951
## [1] "Fuel.tank.capacity"

## 'geom_smooth()' using formula = 'y ~ x'
```



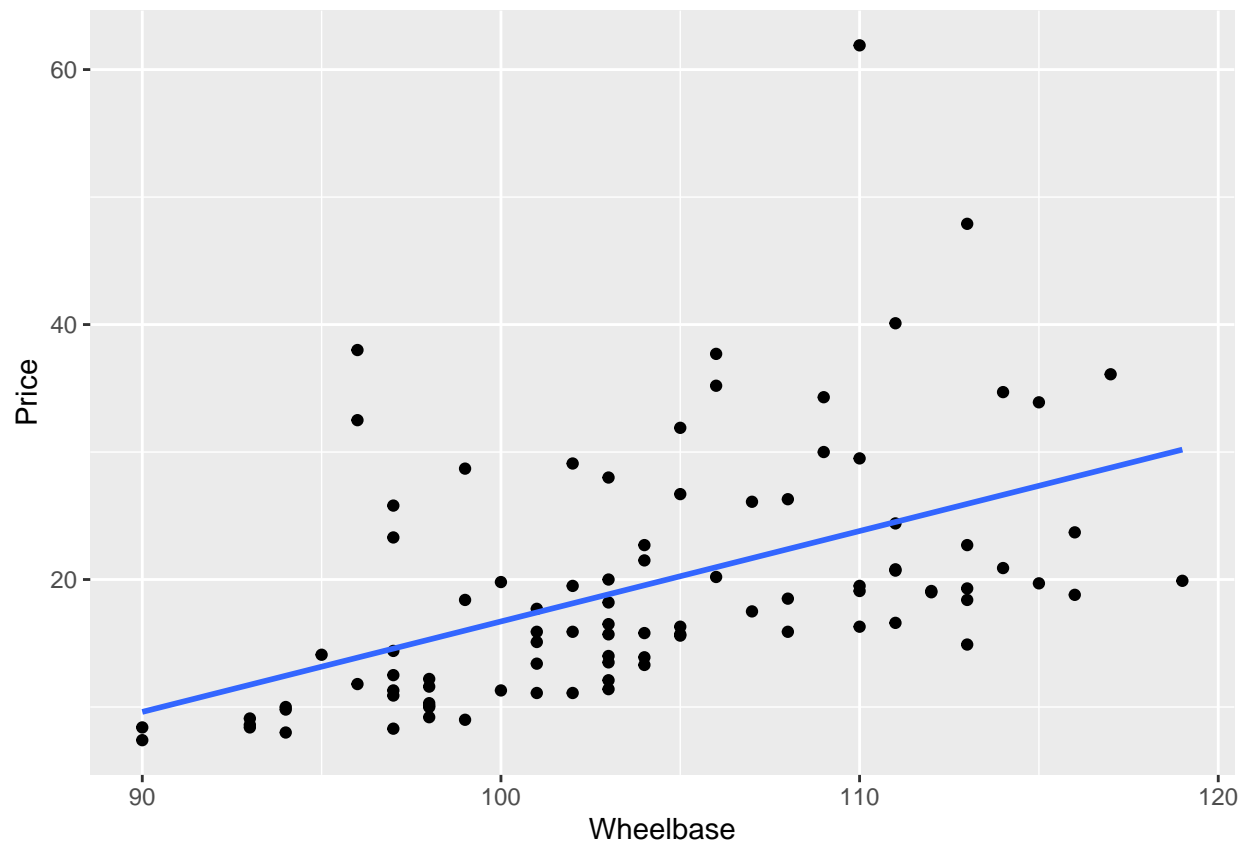
```
## [1] 0.61948  
## [1] "Length"
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```



```
## [1] 0.5036284  
## [1] "Wheelbase"
```

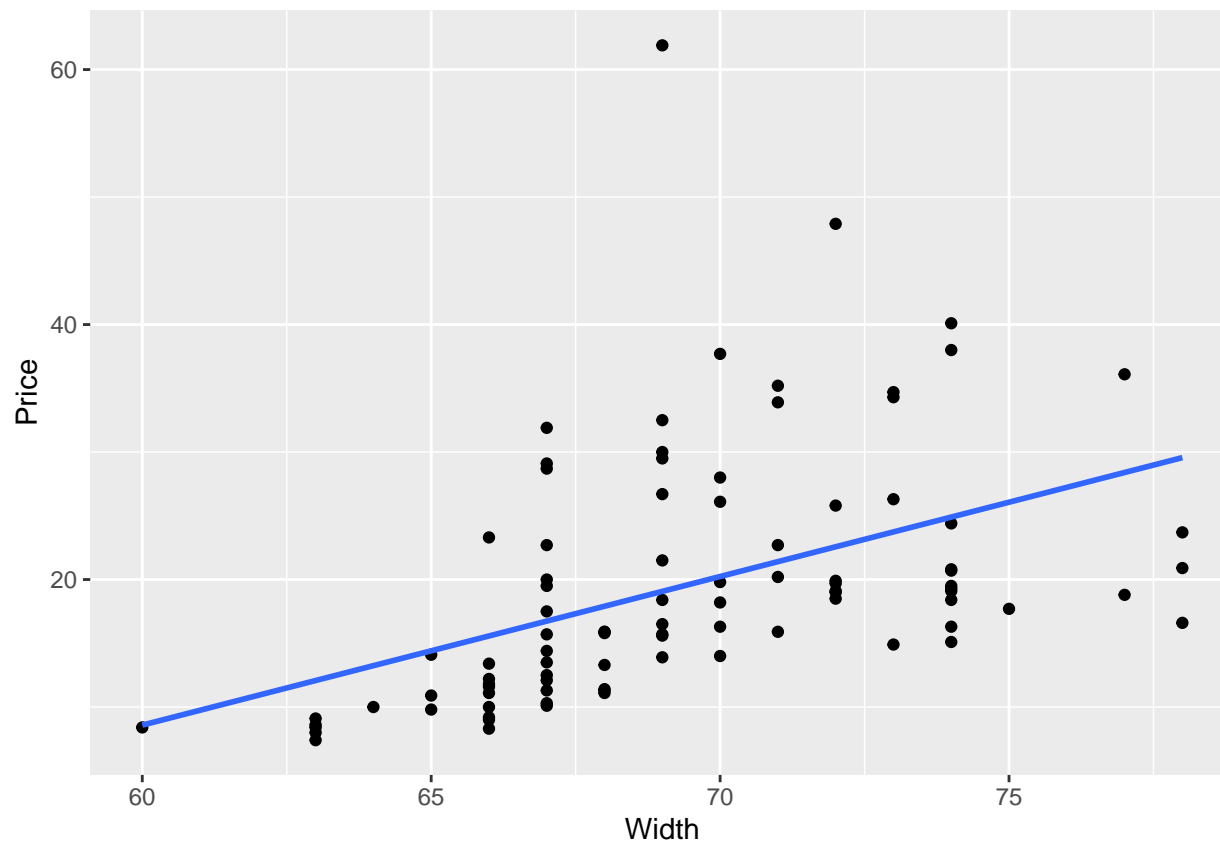
```
## 'geom_smooth()' using formula = 'y ~ x'
```



```
## [1] 0.5008642
```

```
## [1] "Width"
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```

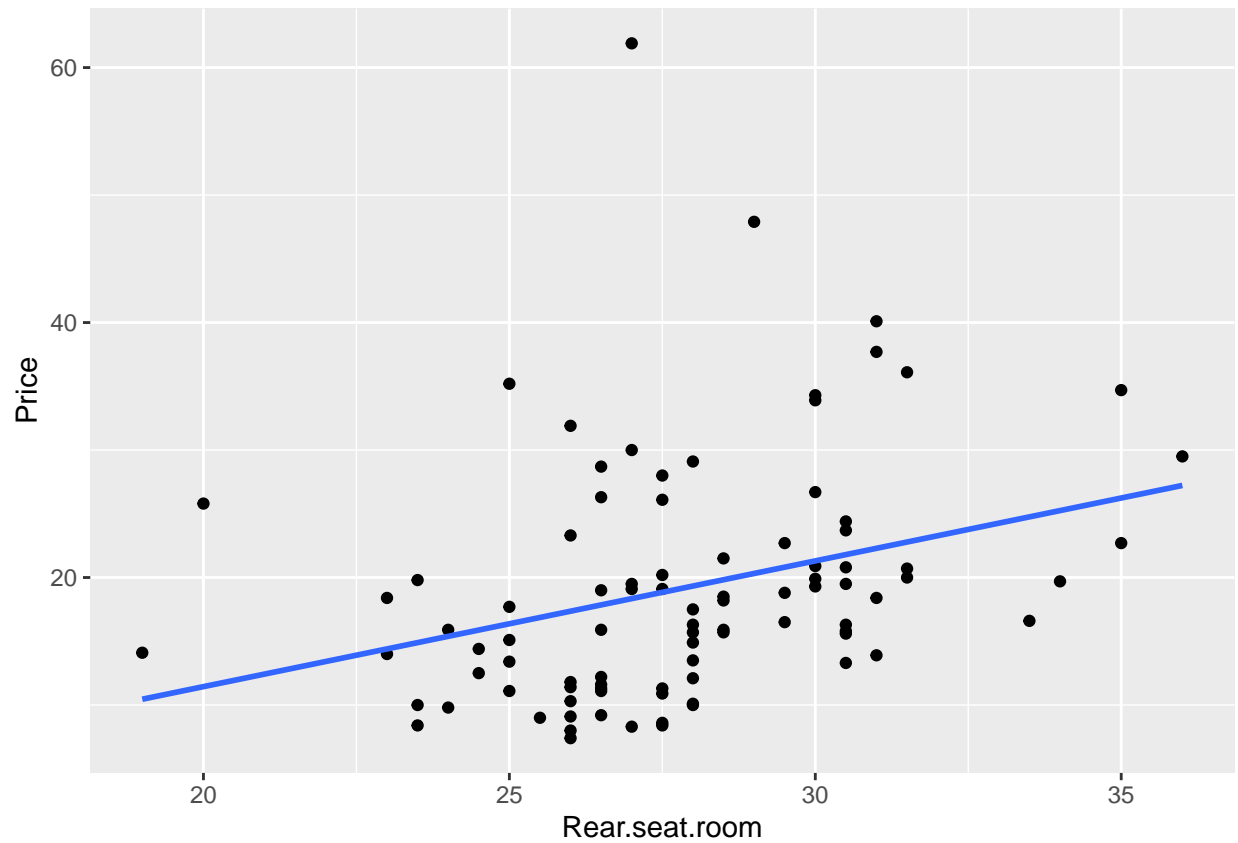


```
## [1] 0.4560279
## [1] "Rear.seat.room"

## 'geom_smooth()' using formula = 'y ~ x'

## Warning: Removed 2 rows containing non-finite outside the scale range
## ('stat_smooth()').

## Warning: Removed 2 rows containing missing values or values outside the scale range
## ('geom_point()').
```

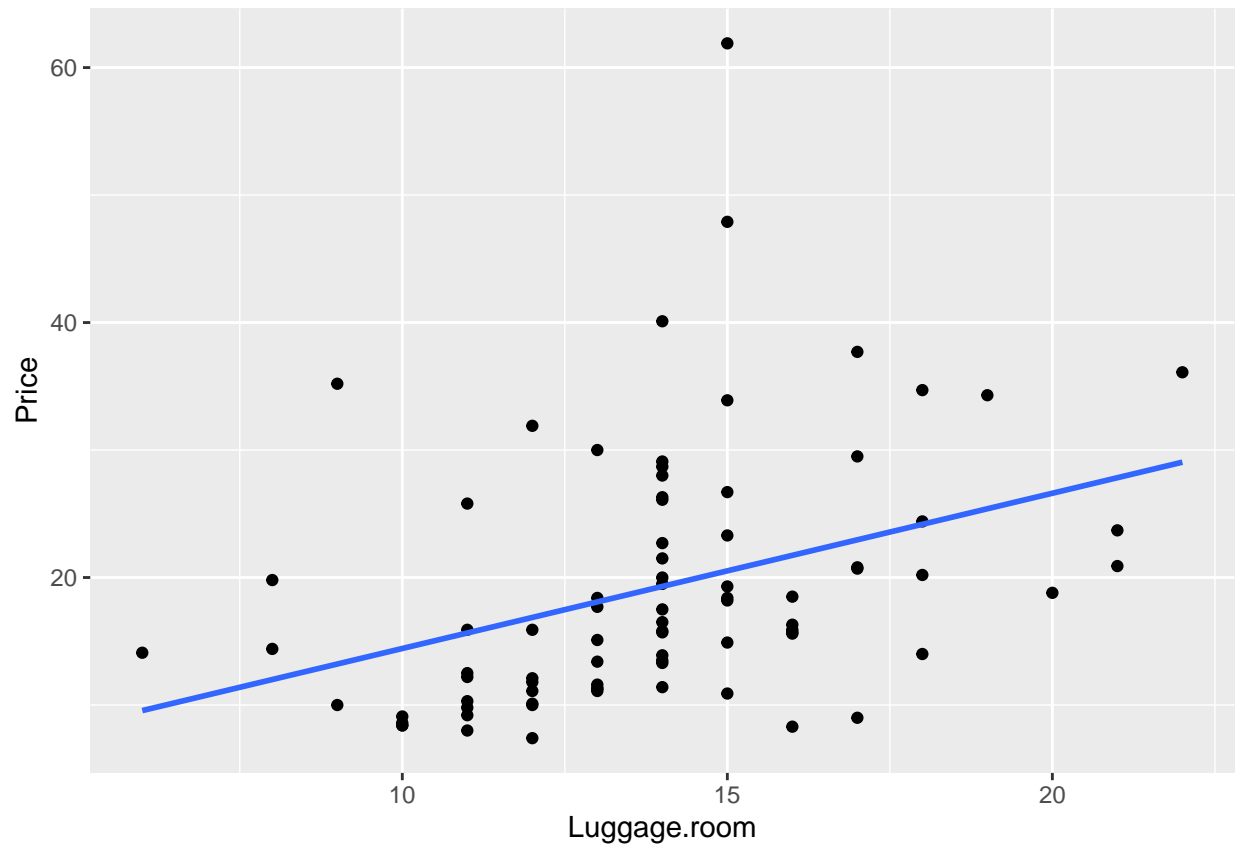



```
## [1] NA
## [1] "Luggage.room"

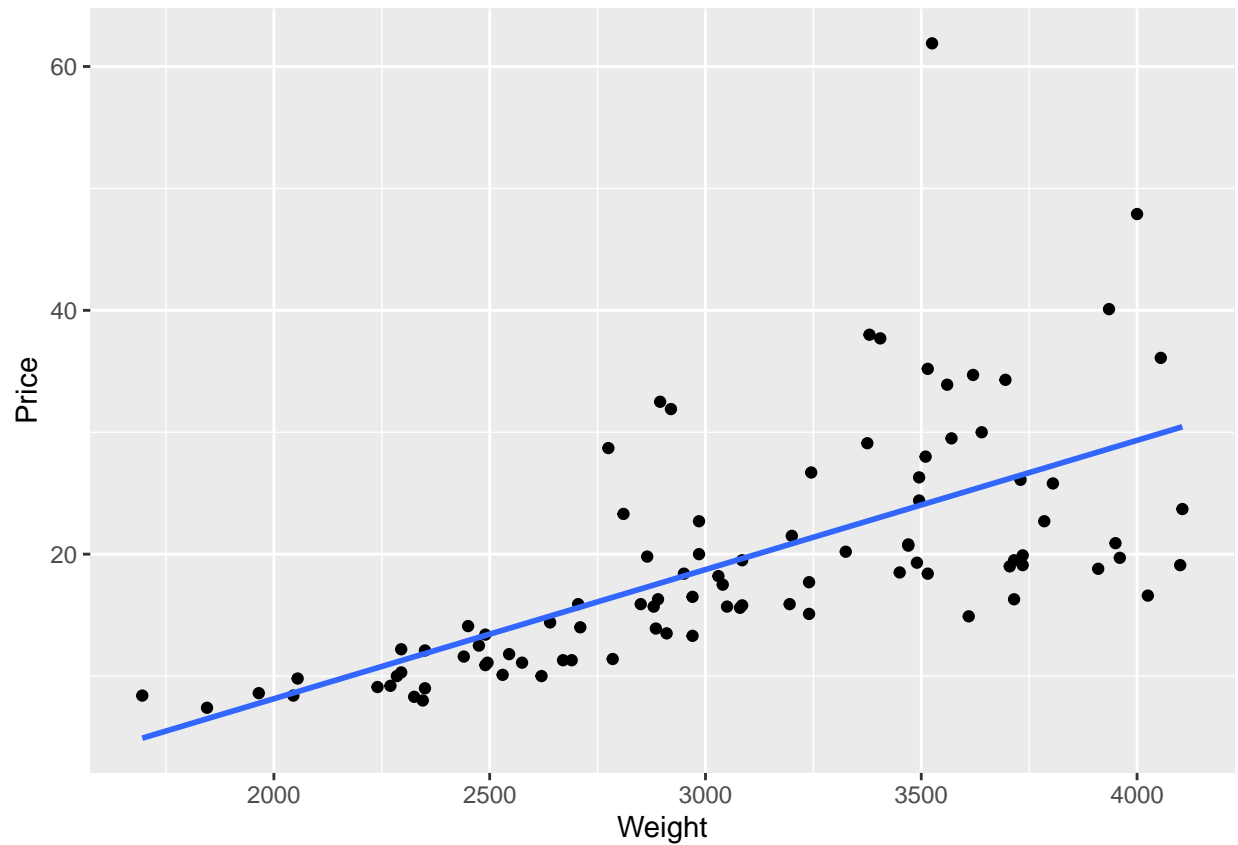
## 'geom_smooth()' using formula = 'y ~ x'

## Warning: Removed 11 rows containing non-finite outside the scale range
## ('stat_smooth()').

## Warning: Removed 11 rows containing missing values or values outside the scale range
## ('geom_point()').
```



```
## [1] NA  
## [1] "Weight"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```



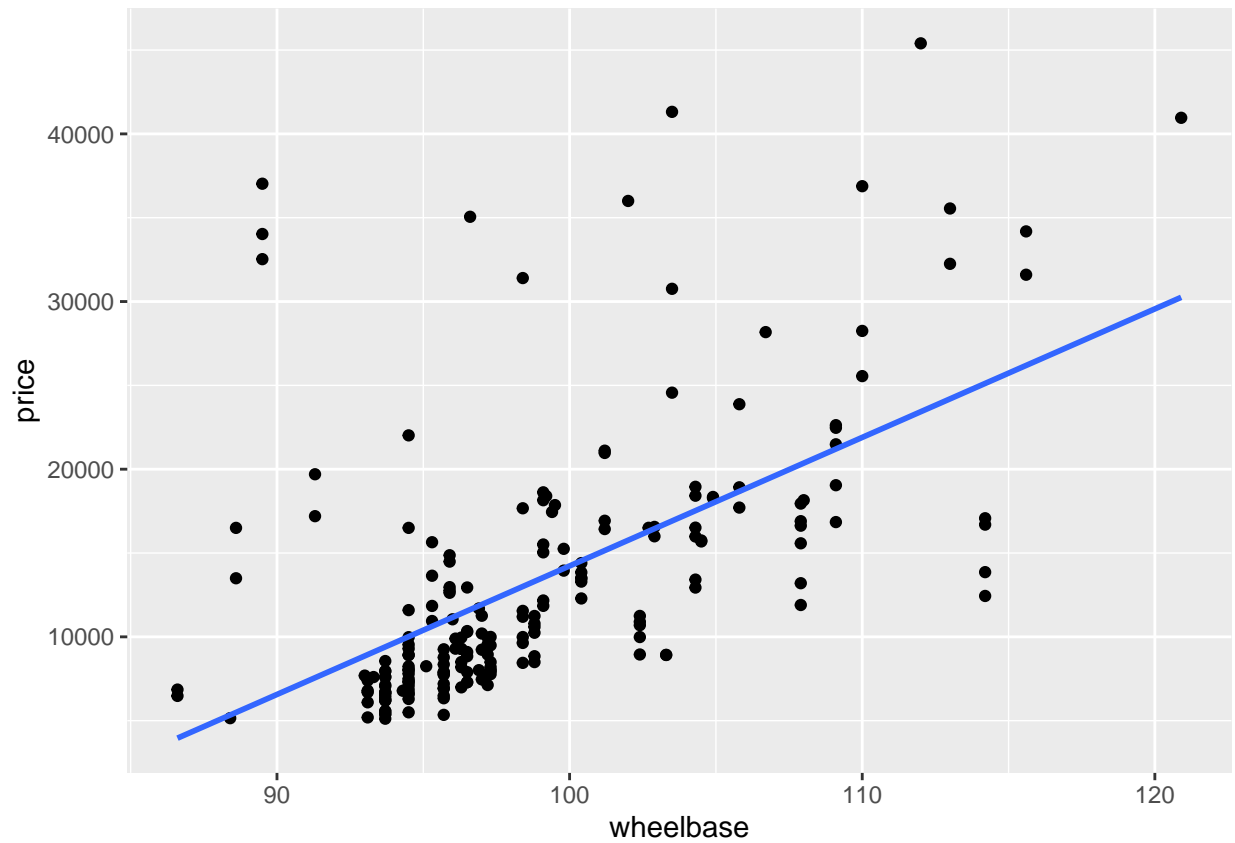
```
## [1] 0.647179
```

```
columns_assign <- c("wheelbase", "carlength", "carwidth", "carheight", "curbweight", "enginesize", "horsepower")
```

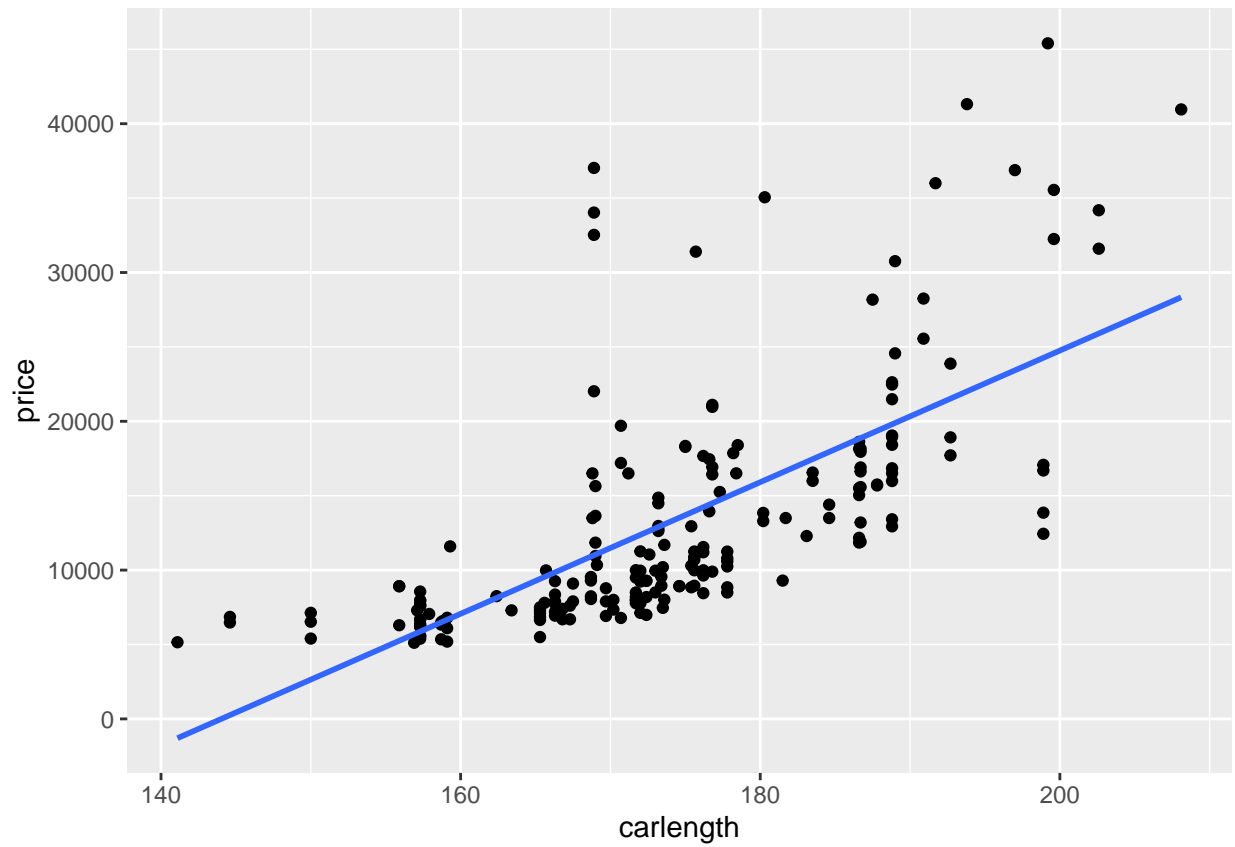
```
for (var in columns_assign)
{
  print(var);
  p <- ggplot(cars_assign, aes_string(x = var, y = "price")) +
    geom_point() +
    geom_smooth(method = "lm", se = FALSE);
  print(p);
  correlation <- cor(cars_assign$price, cars_assign[[var]]);
  print(correlation)
}
```

```
## [1] "wheelbase"
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```

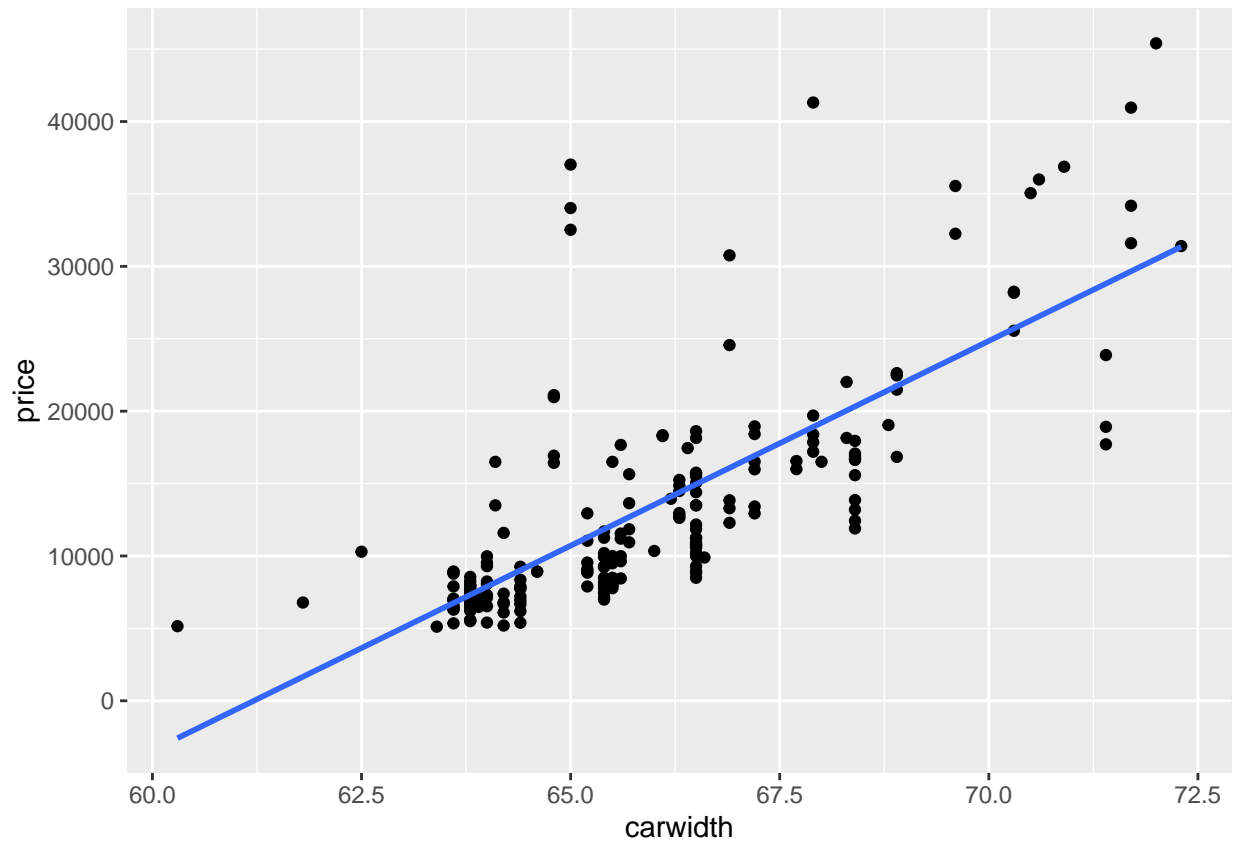


```
## [1] 0.5778156  
## [1] "carlength"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```



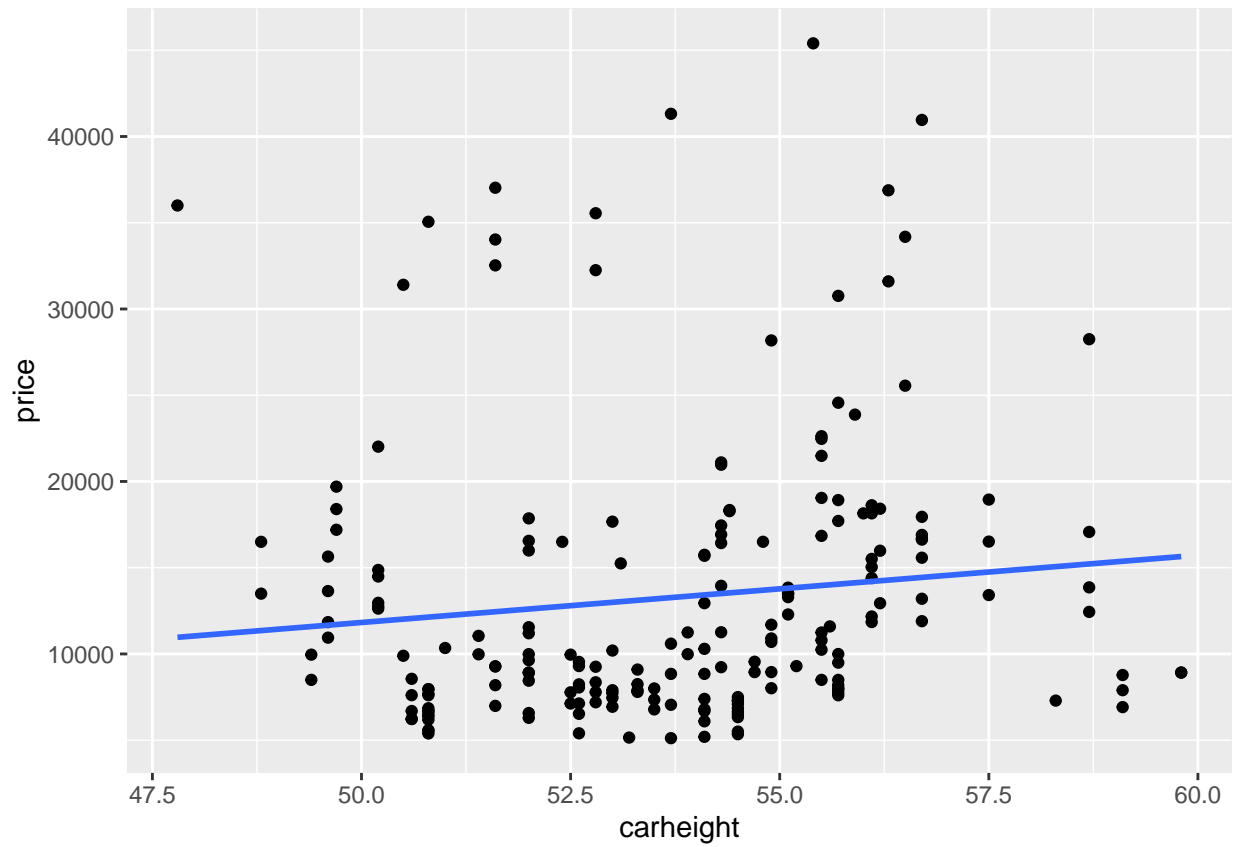
```
## [1] 0.68292  
## [1] "carwidth"
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```



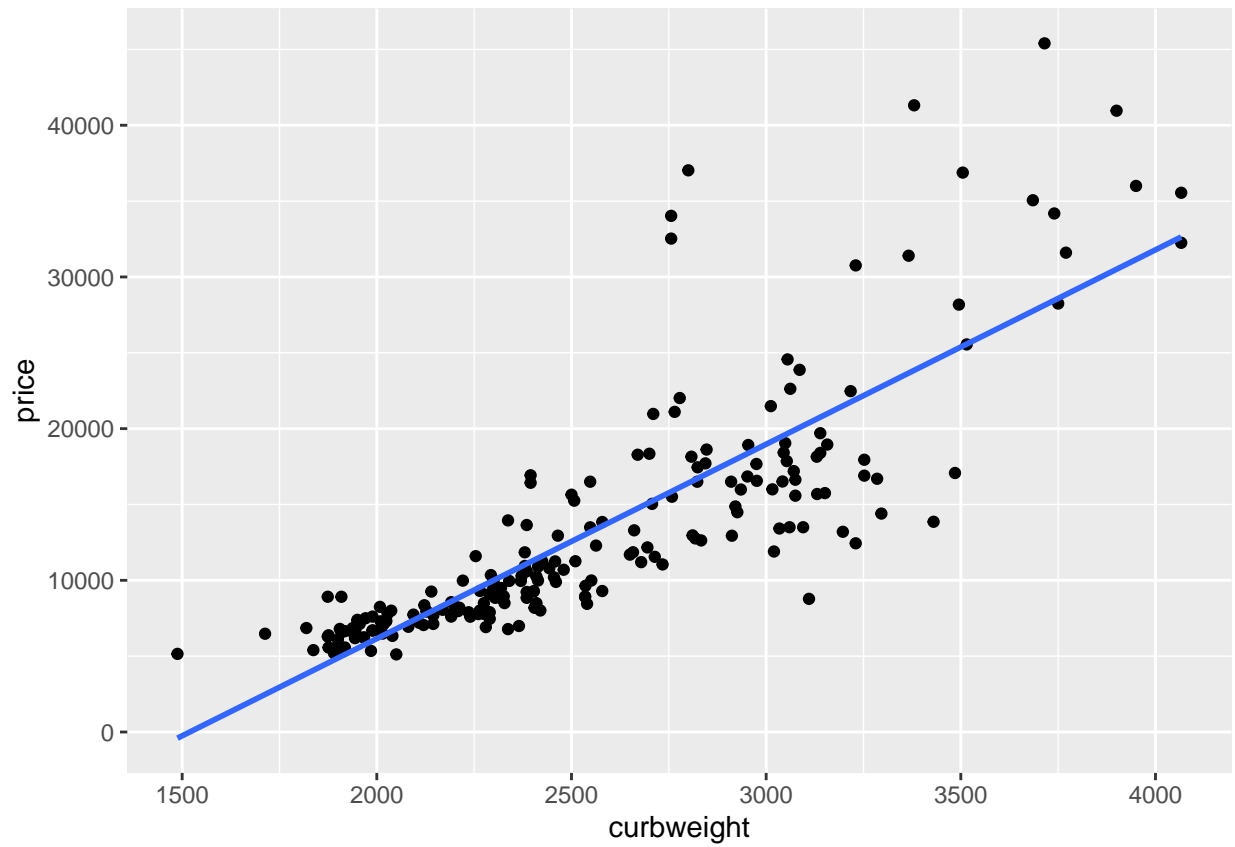
```
## [1] 0.7593253  
## [1] "carheight"
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```



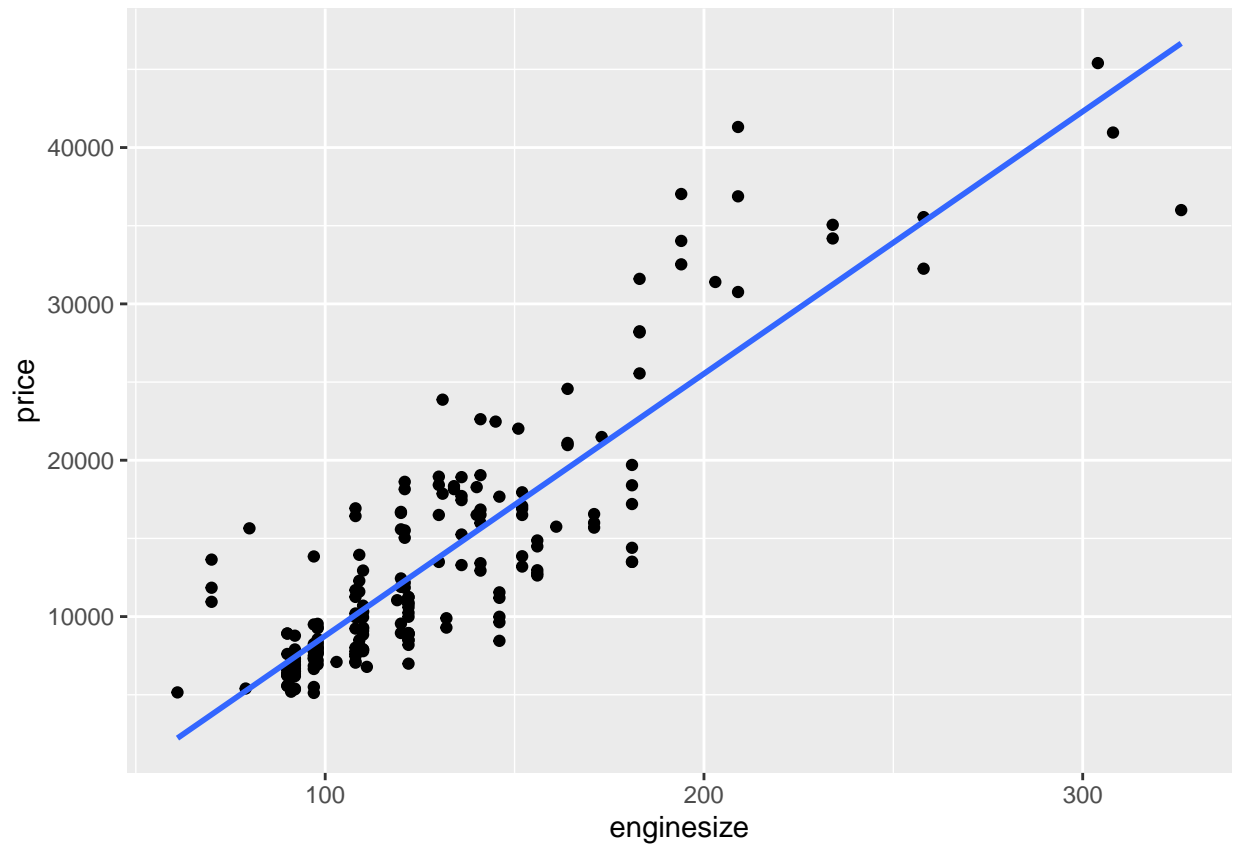
```
## [1] 0.1193362
## [1] "curbweight"

## 'geom_smooth()' using formula = 'y ~ x'
```



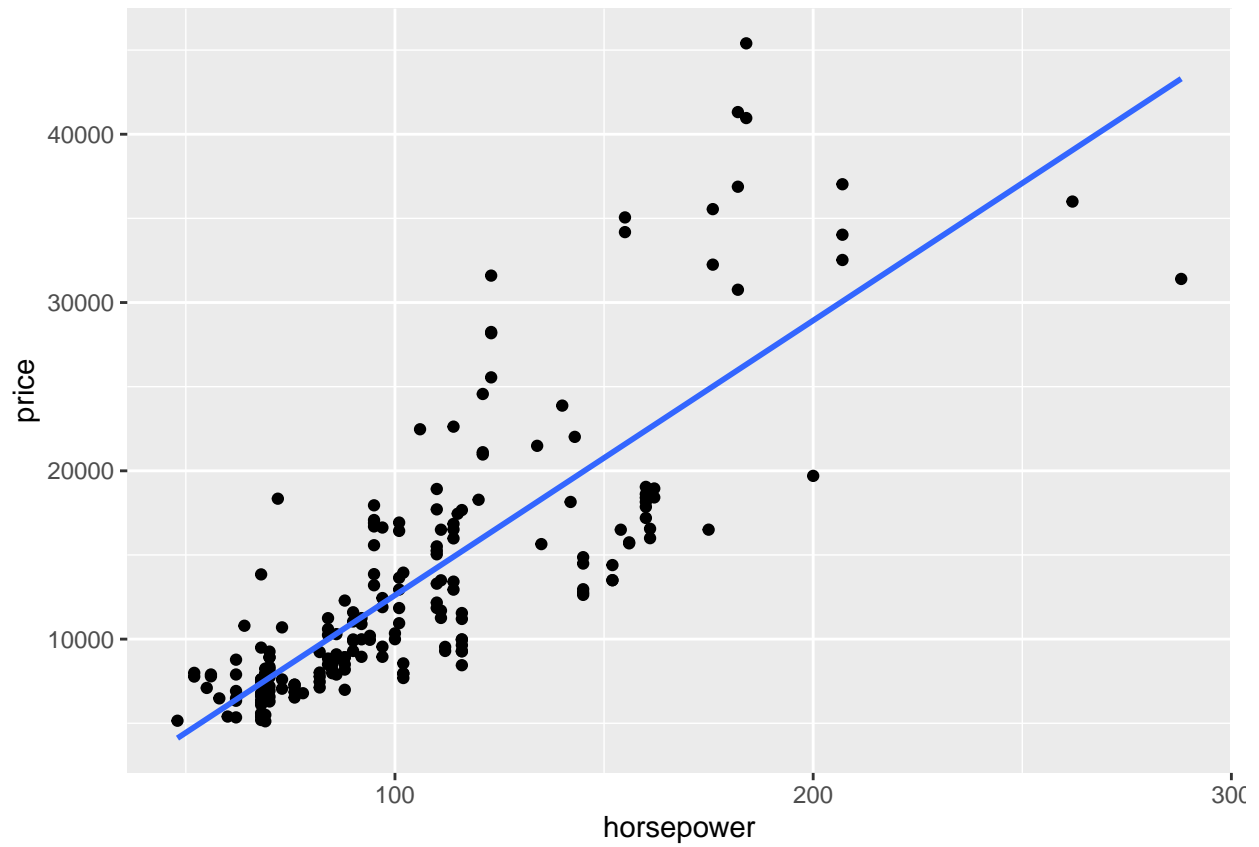
```
## [1] 0.8353049  
## [1] "enginesize"
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```

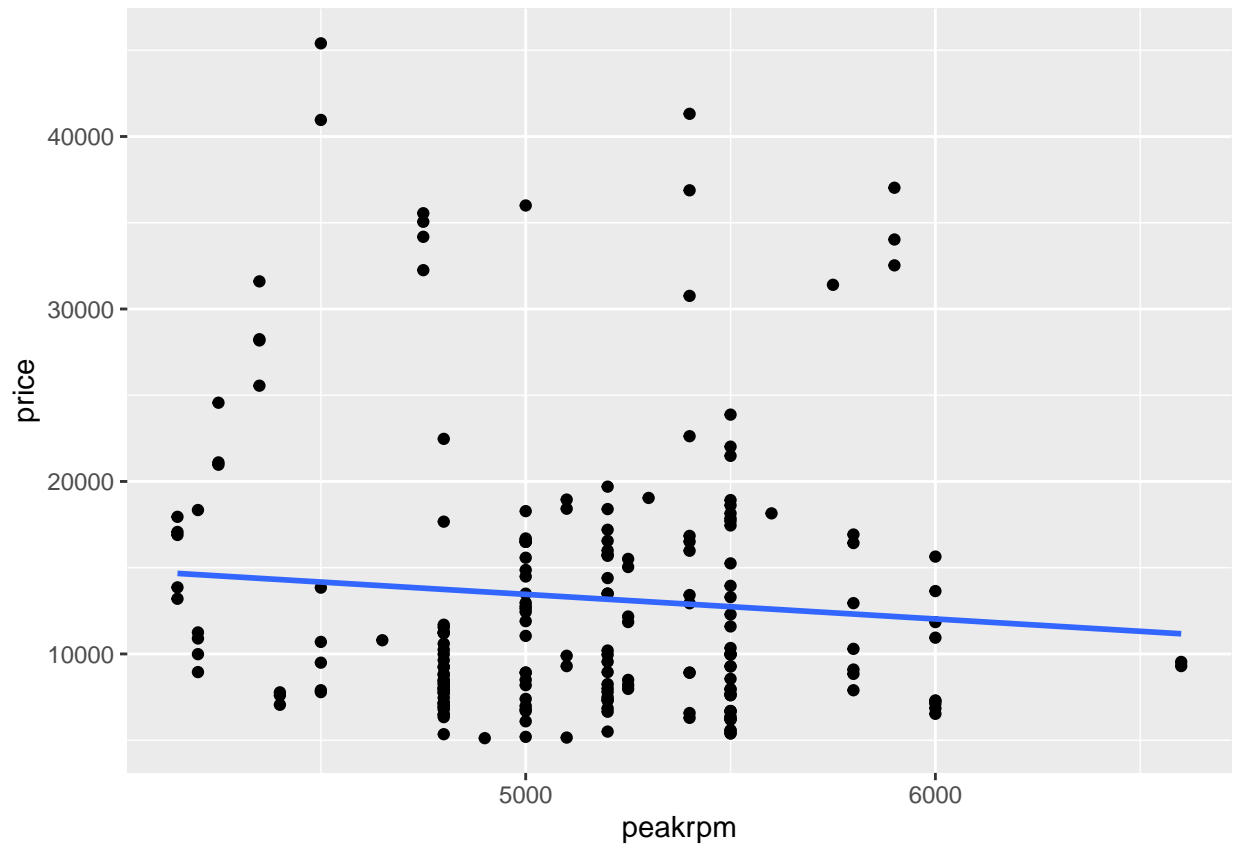
```
## [1] 0.8741448  
## [1] "horsepower"
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```



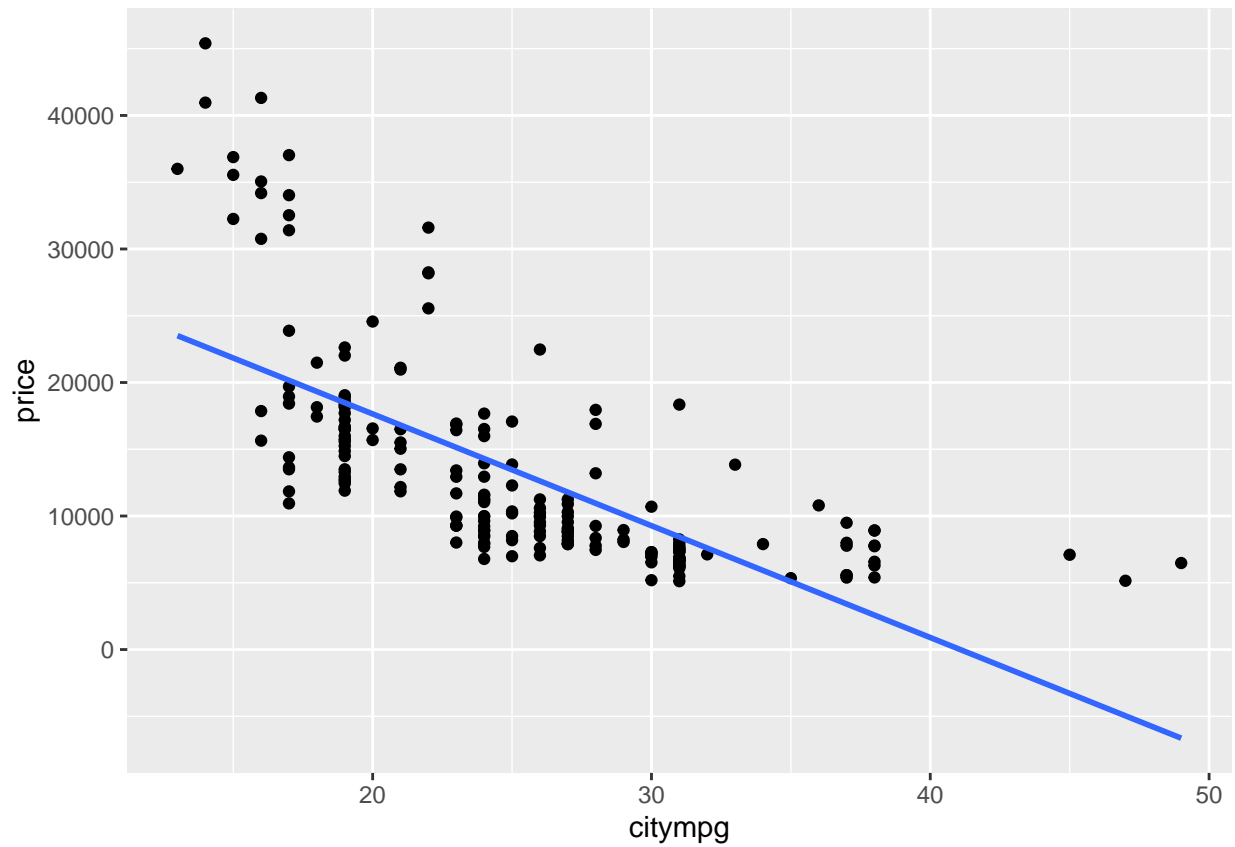
```
## [1] 0.8081388  
## [1] "peakrpm"
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```



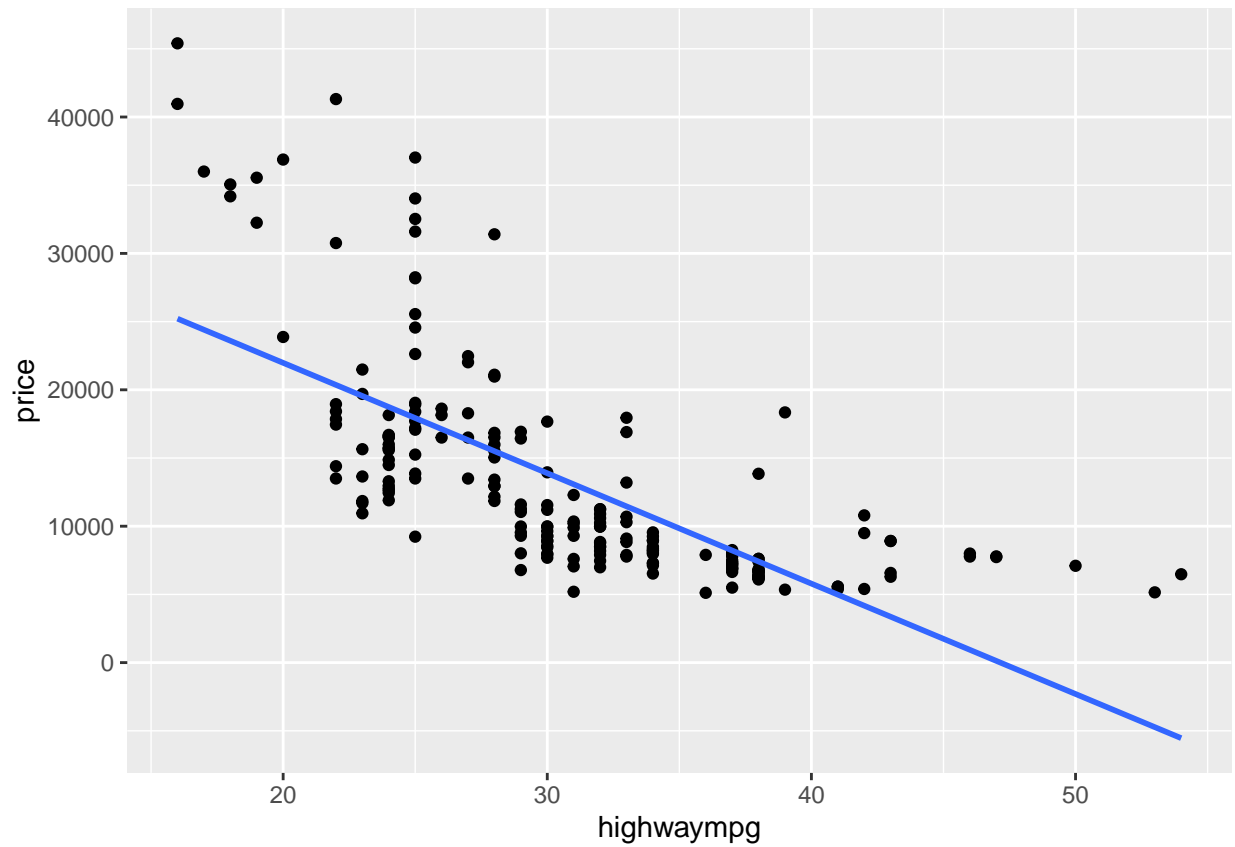
```
## [1] -0.08526715  
## [1] "citympg"
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```



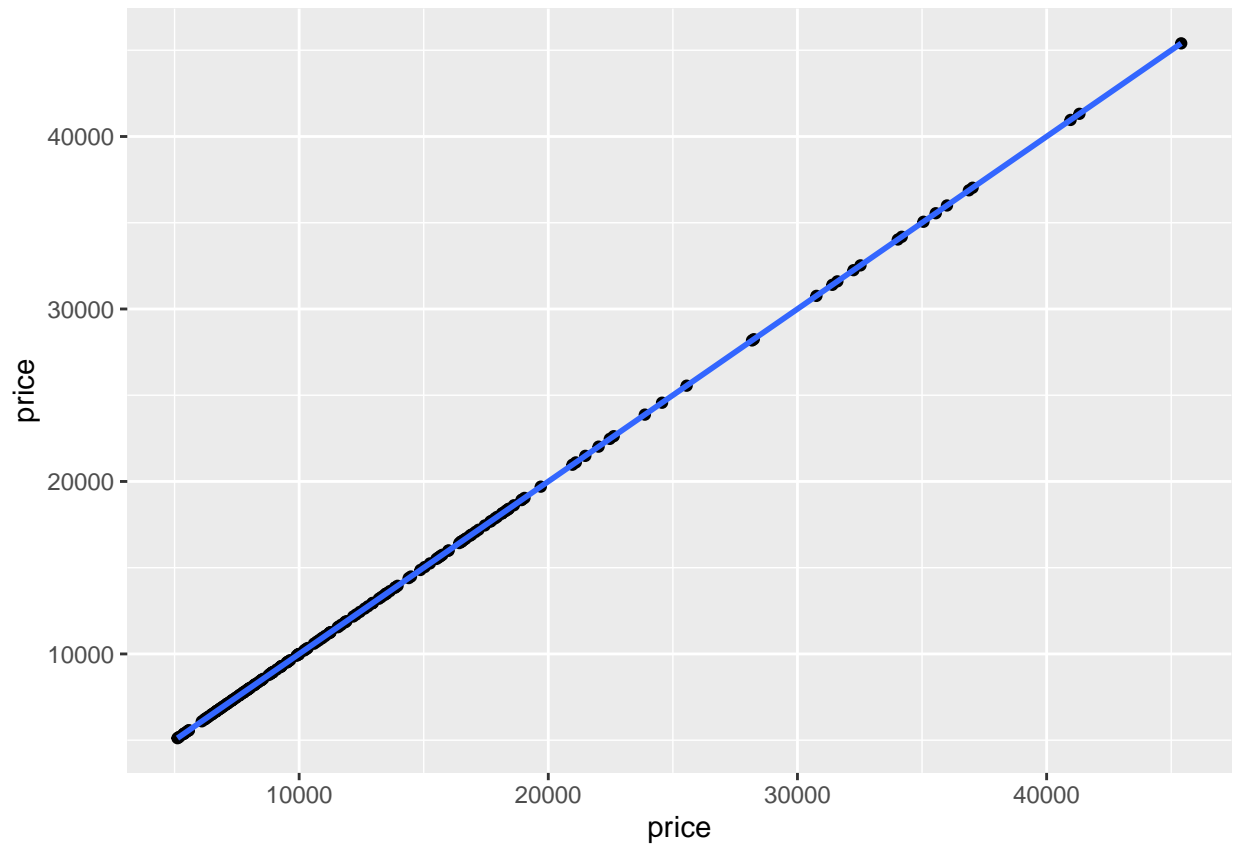
```
## [1] -0.6857513
## [1] "highwaympg"

## 'geom_smooth()' using formula = 'y ~ x'
```



```
## [1] -0.6975991  
## [1] "price"
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```

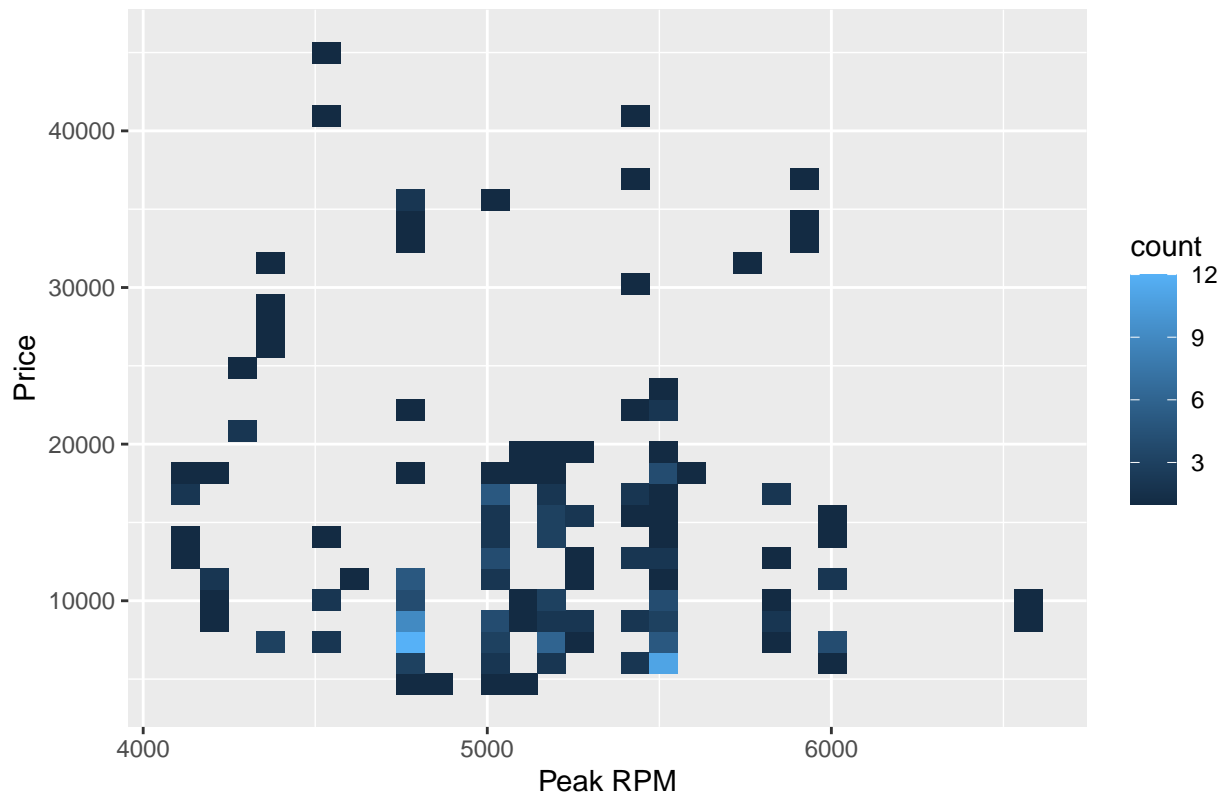


```
## [1] 1
```

Heatmap visualization

```
ggplot(cars_assign, aes(x = peakrpm, y = price)) +  
  geom_bin2d() +  
  labs(x = "Peak RPM", y = "Price", title = "Heatmap of Price vs. Peak RPM")
```

Heatmap of Price vs. Peak RPM



Cars93 linear regression model

```
model93 <- lm(Price ~ MPG.city + EngineSize + Horsepower + Fuel.tank.capacity + Weight, data = cars93)
summary(model93)
```

```
##
## Call:
## lm(formula = Price ~ MPG.city + EngineSize + Horsepower + Fuel.tank.capacity +
##     Weight, data = cars93)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -15.824  -2.713   -0.456    1.444   31.936
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -3.509880  11.026931  -0.318   0.751
## MPG.city       -0.078651   0.215438  -0.365   0.716
## EngineSize     -0.797961   1.187025  -0.672   0.503
## Horsepower      0.128456   0.019018   6.754 1.55e-09 ***
## Fuel.tank.capacity 0.040706   0.445052   0.091   0.927
## Weight         0.002523   0.003136   0.805   0.423
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 6.015 on 87 degrees of freedom
```

```
## Multiple R-squared:  0.6333, Adjusted R-squared:  0.6122
## F-statistic: 30.05 on 5 and 87 DF,  p-value: < 2.2e-16
```

cars_assignment linear regression model

```
model_assign <- lm(price ~ carwidth + curbweight + enginesize + horsepower + citympg, data = cars_assign)
summary(model_assign)
```

```
##
## Call:
## lm(formula = price ~ carwidth + curbweight + enginesize + horsepower +
##     citympg, data = cars_assign)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -8184.4 -1715.3   -91.6  1339.2 14273.7
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -47388.322  13175.253  -3.597  0.000406 ***
## carwidth      574.857    225.058   2.554  0.011390 *
## curbweight     2.378     1.324   1.795  0.074101 .
## enginesize    82.921     13.342   6.215  2.95e-09 ***
## horsepower    52.703     13.280   3.968  0.000101 ***
## citympg       27.400     70.724   0.387  0.698860
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3432 on 199 degrees of freedom
## Multiple R-squared:  0.8199, Adjusted R-squared:  0.8154
## F-statistic: 181.2 on 5 and 199 DF,  p-value: < 2.2e-16
```

mpgcity+highway horsepower enginesize weight

cars_assign - Correlation between pairs of variables except price

```
for (var in c("carwidth", "curbweight", "enginesize", "horsepower", "citympg")) {
  for (var1 in c("carwidth", "curbweight", "enginesize", "horsepower", "citympg")) {

    if (var != var1) {
      print(paste("x =", var, ", y =", var1))

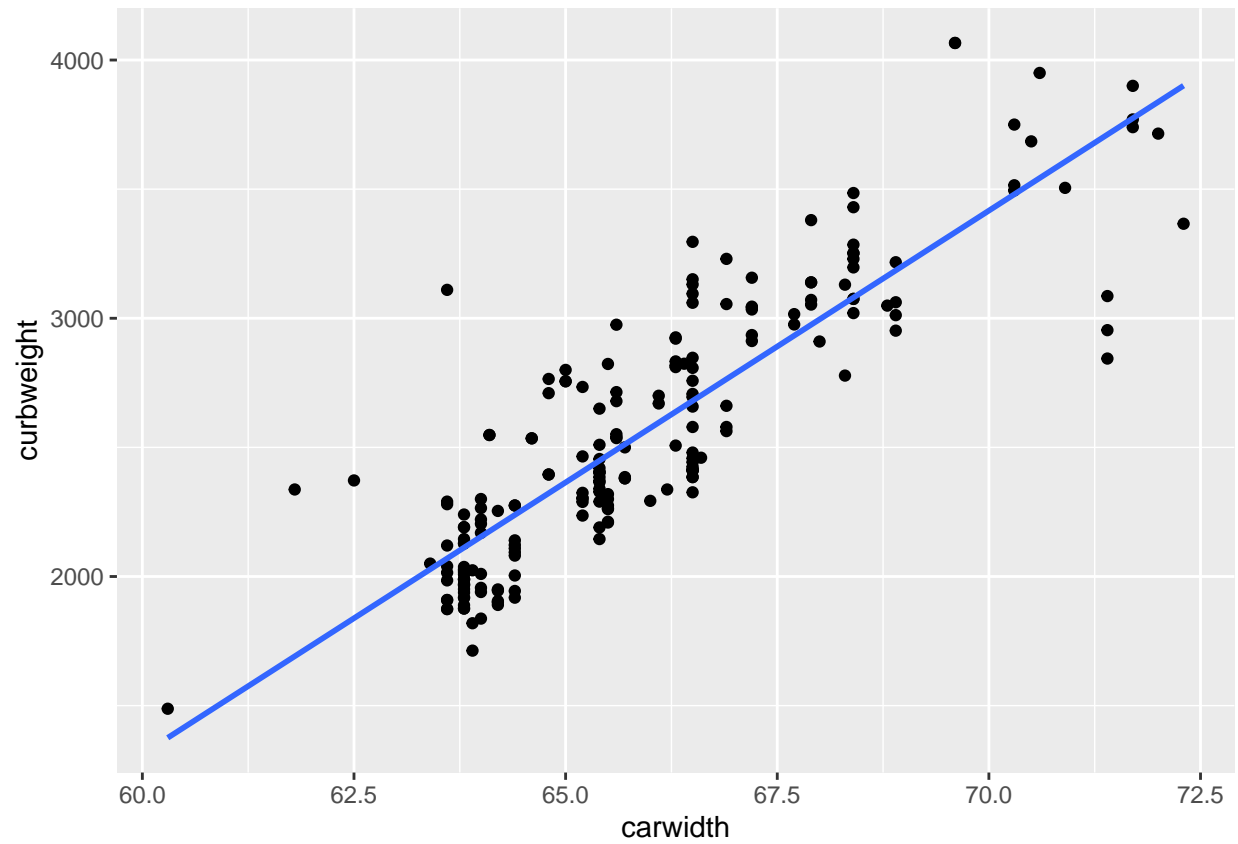
      p <- ggplot(cars_assign, aes_string(x = var, y = var1)) +
        geom_point() +
        geom_smooth(method = "lm", se = FALSE)
      print(p)

      correlation <- cor(cars_assign[[var1]], cars_assign[[var]])
      print(paste("Correlation between", var, "and", var1, ":", correlation))
    }
  }
}
```

```
## [1] "x = carwidth , y = curbweight"
```



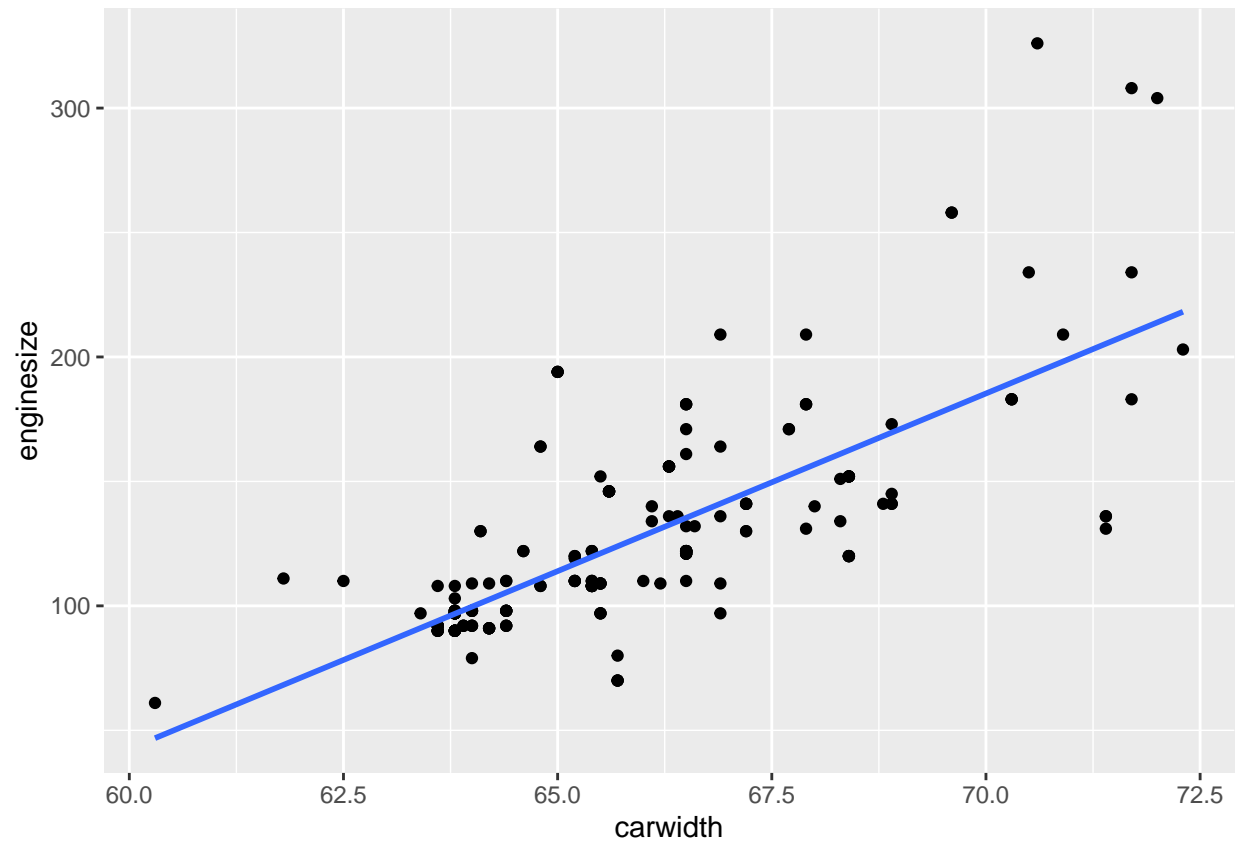
```
## 'geom_smooth()' using formula = 'y ~ x'
```



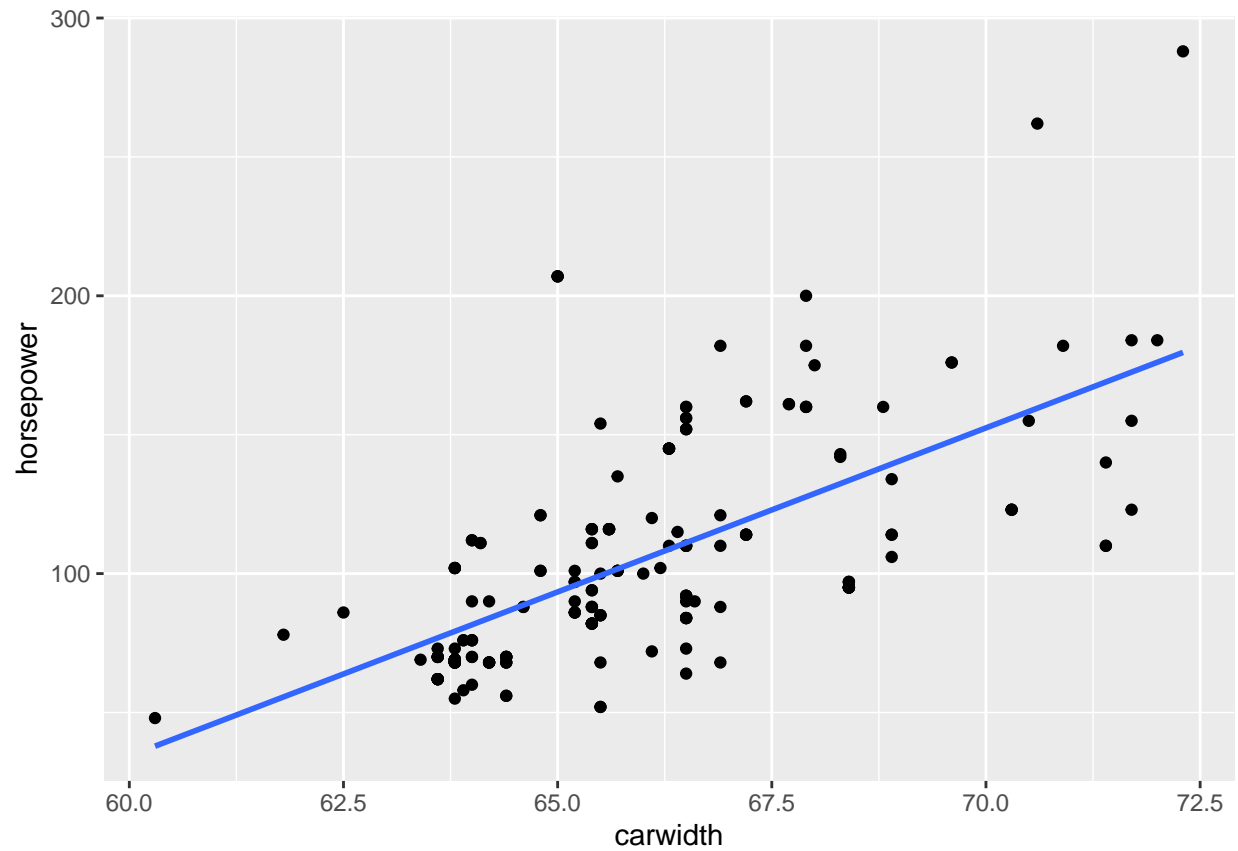
```
## [1] "Correlation between carwidth and curbweight : 0.867032464679123"
```

```
## [1] "x = carwidth , y = enginesize"
```

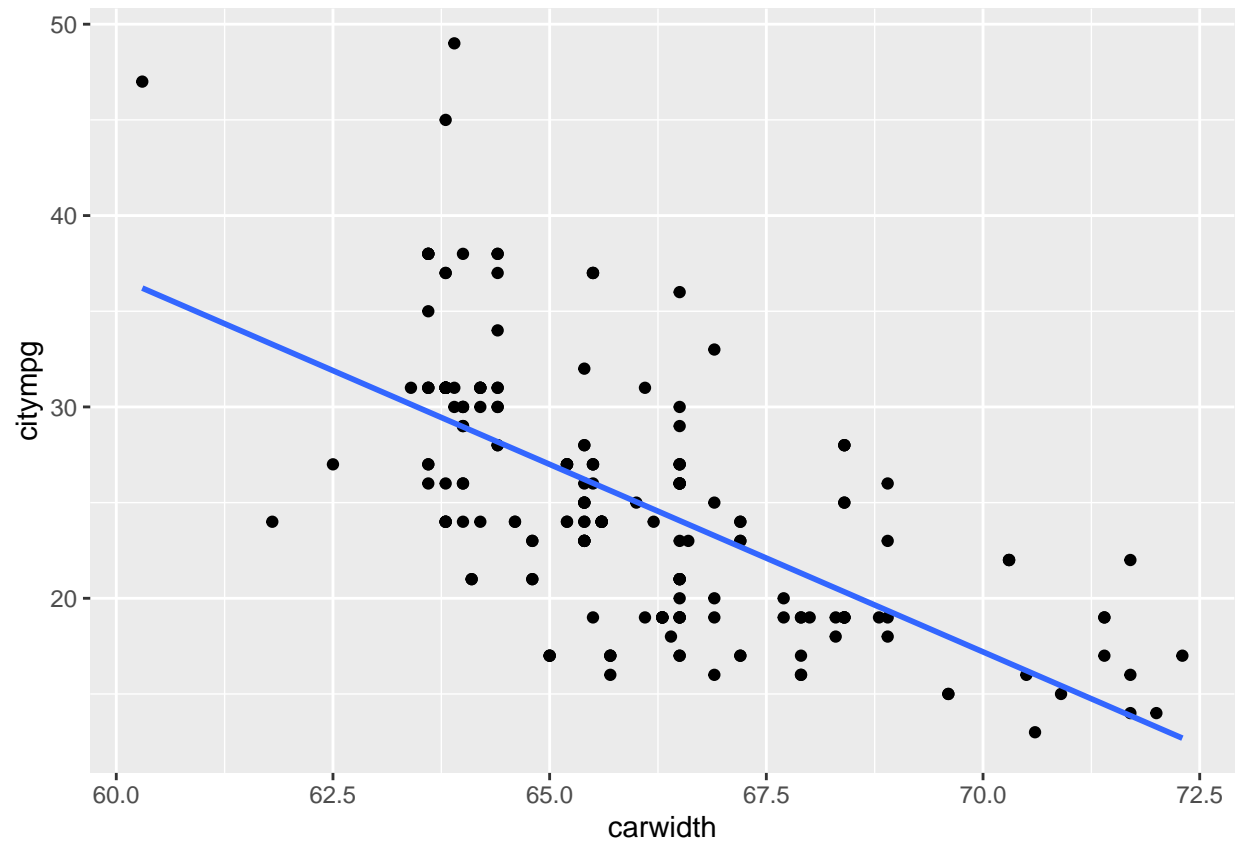
```
## 'geom_smooth()' using formula = 'y ~ x'
```



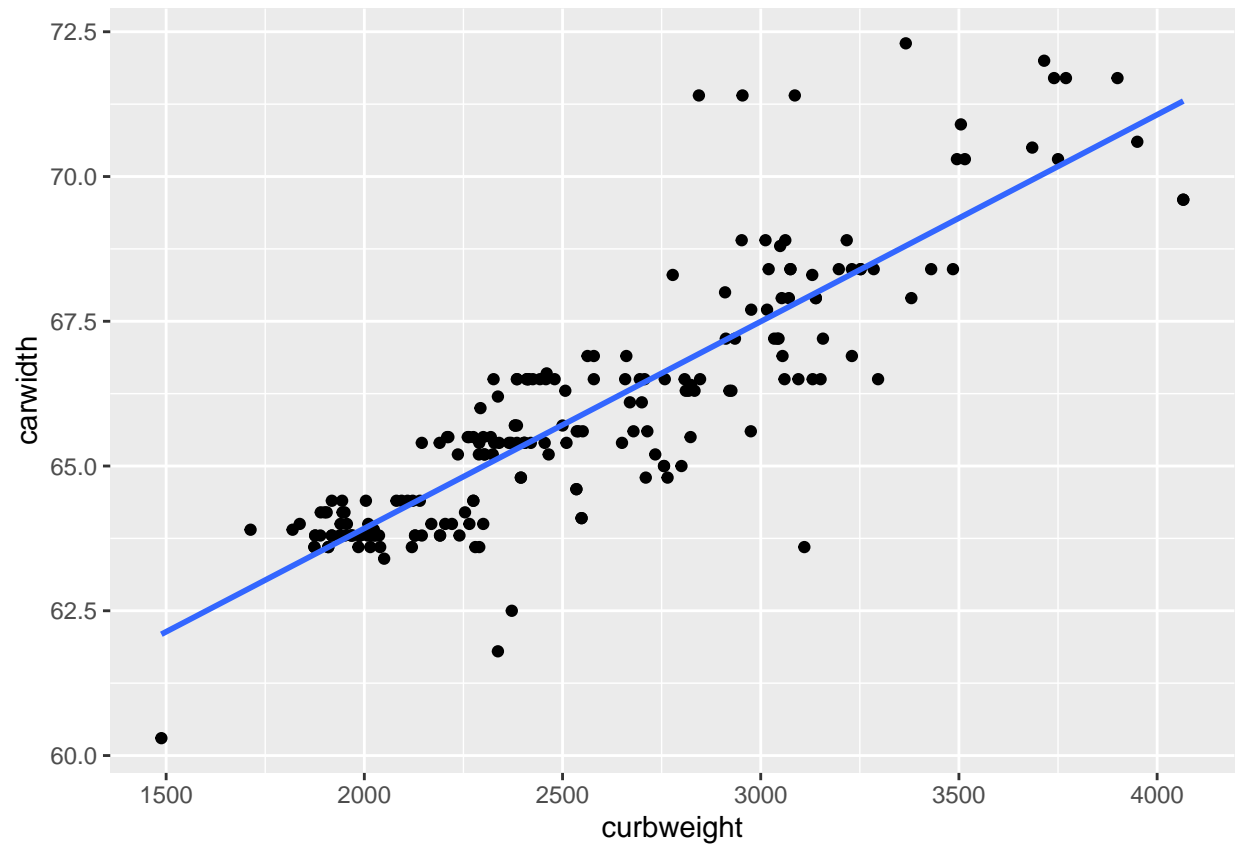
```
## [1] "Correlation between carwidth and enginesize : 0.735433404779637"  
## [1] "x = carwidth , y = horsepower"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```



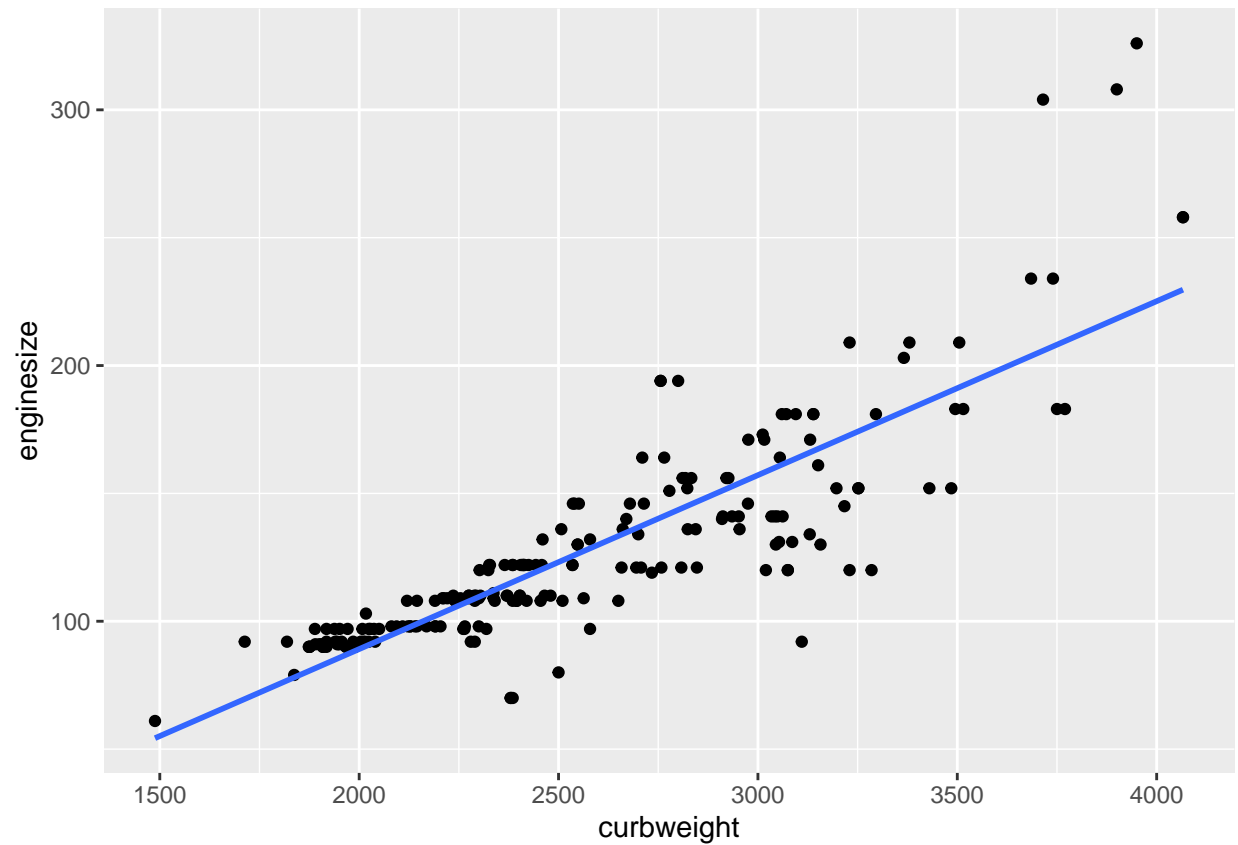
```
## [1] "Correlation between carwidth and horsepower : 0.640732075572292"  
## [1] "x = carwidth , y = citympg"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```



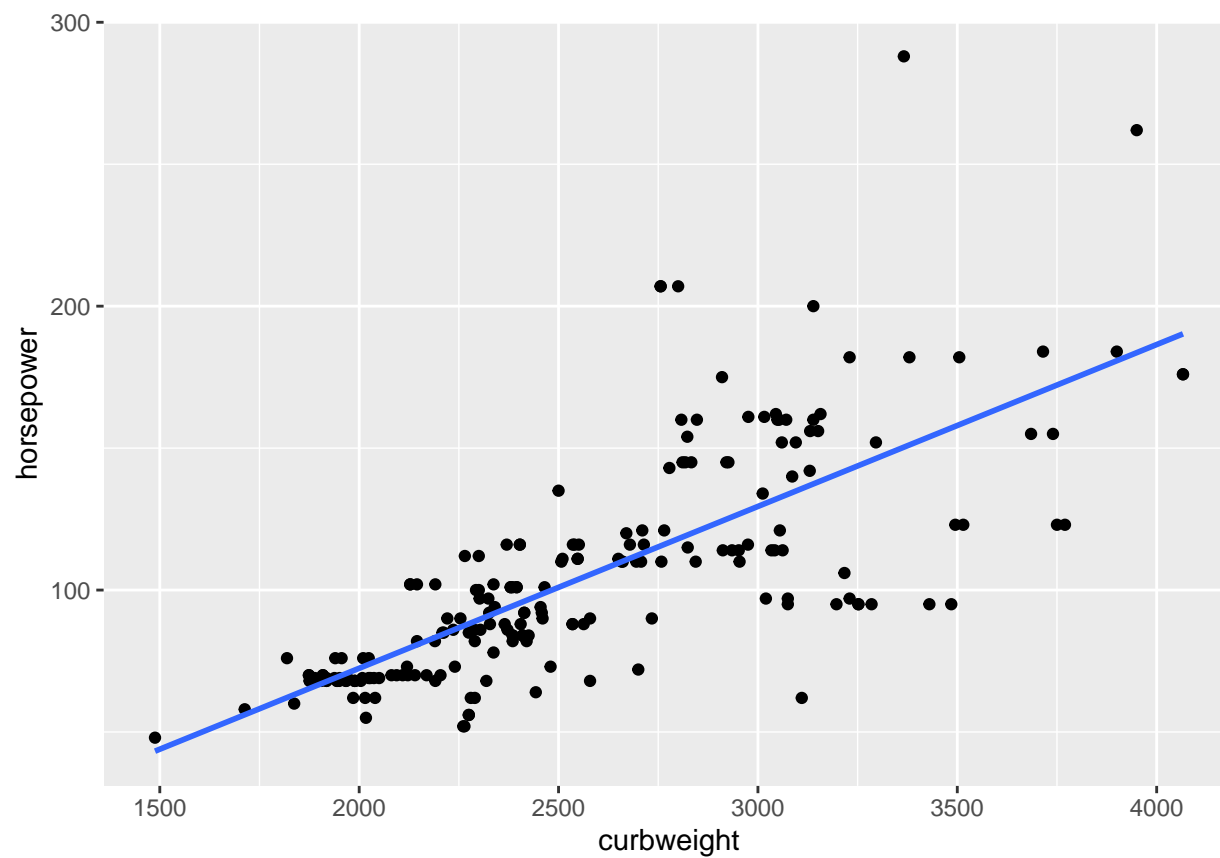
```
## [1] "Correlation between carwidth and citympg : -0.642704340710898"  
## [1] "x = curbweight , y = carwidth"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```



```
## [1] "Correlation between curbweight and carwidth : 0.867032464679123"  
## [1] "x = curbweight , y = enginesize"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```

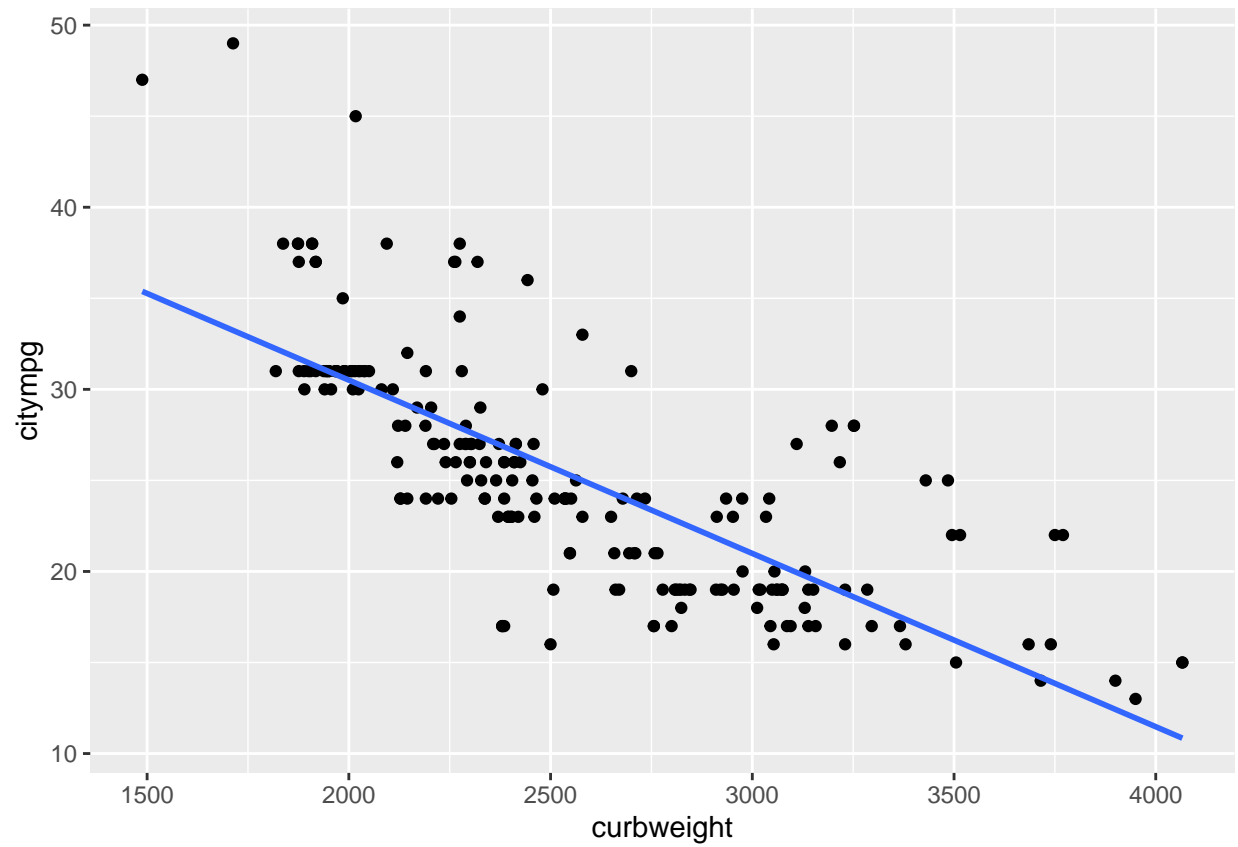


```
## [1] "Correlation between curbweight and enginesize : 0.850594073426277"  
## [1] "x = curbweight , y = horsepower"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```

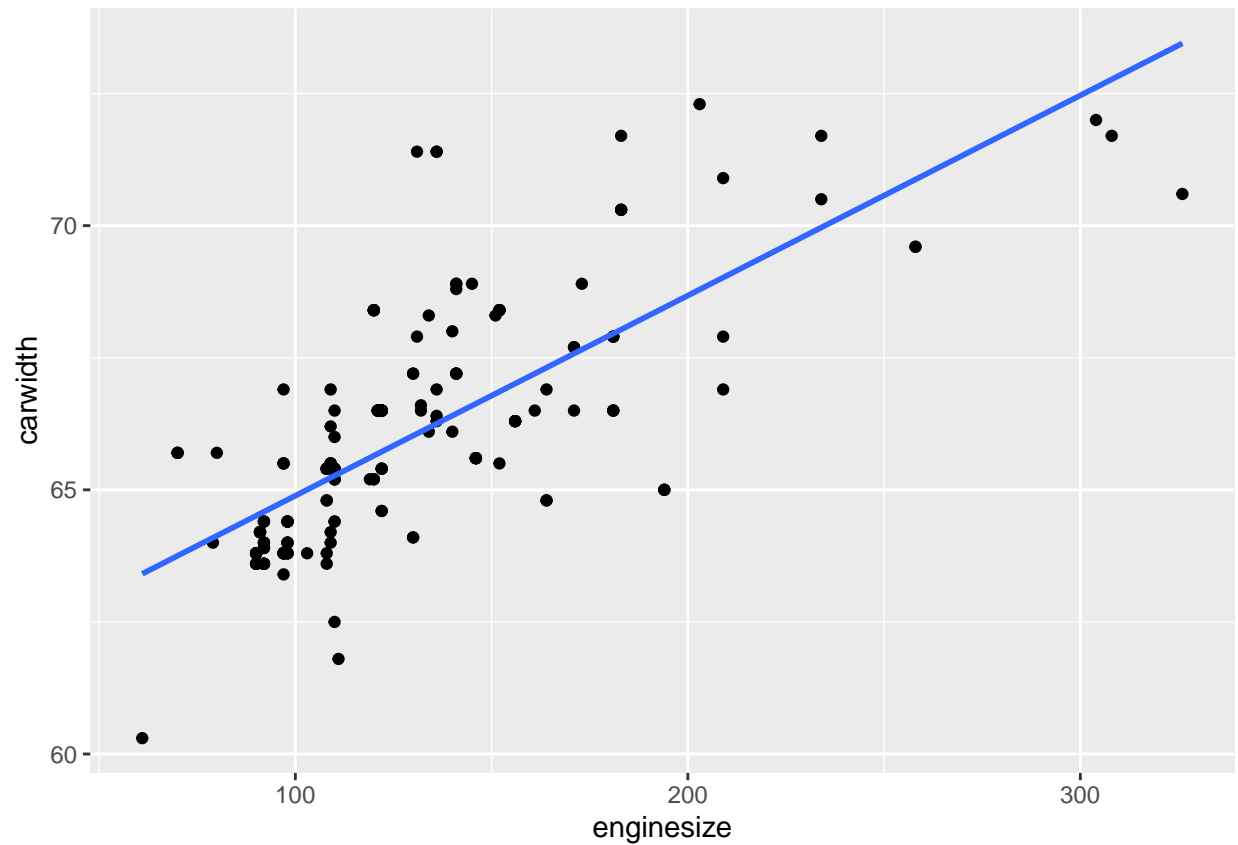


```
## [1] "Correlation between curbweight and horsepower : 0.750739251443481"  
## [1] "x = curbweight , y = citympg"
```

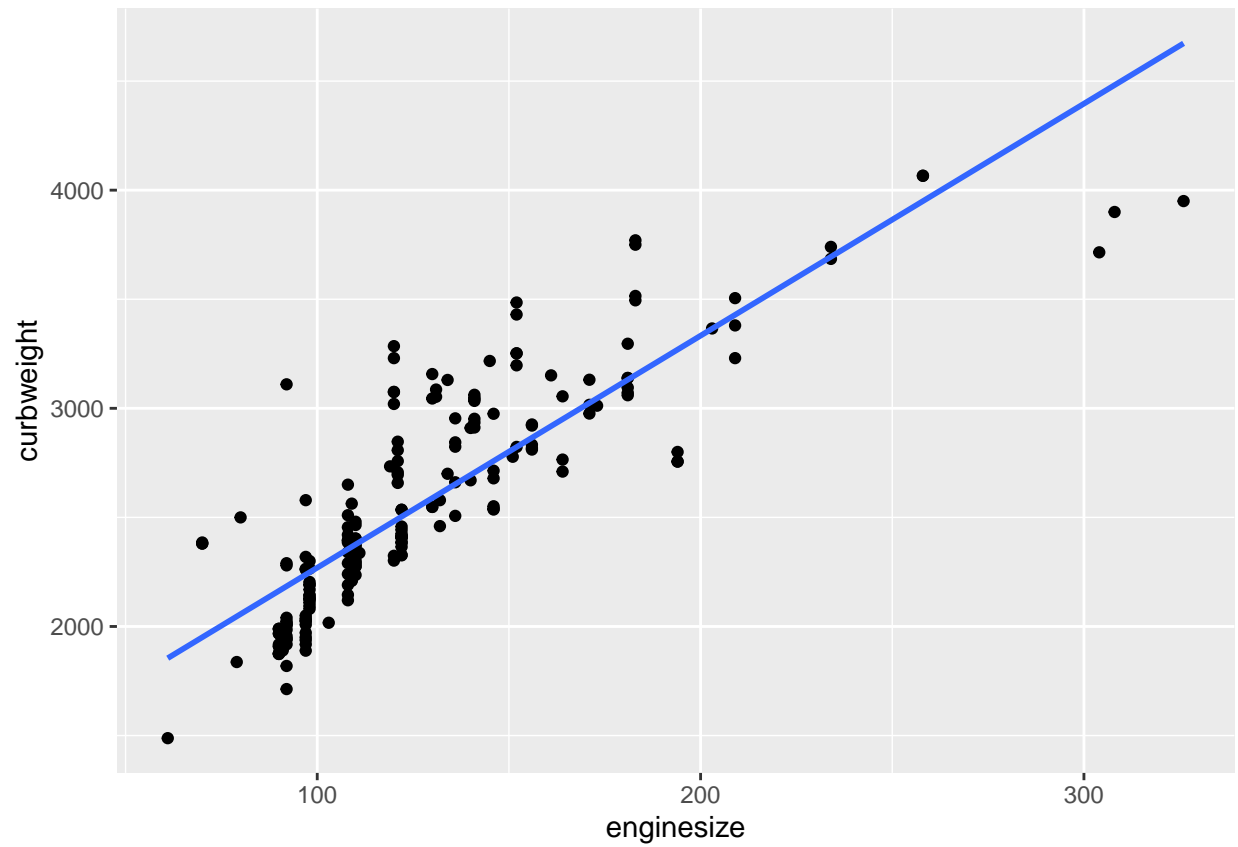
```
## 'geom_smooth()' using formula = 'y ~ x'
```



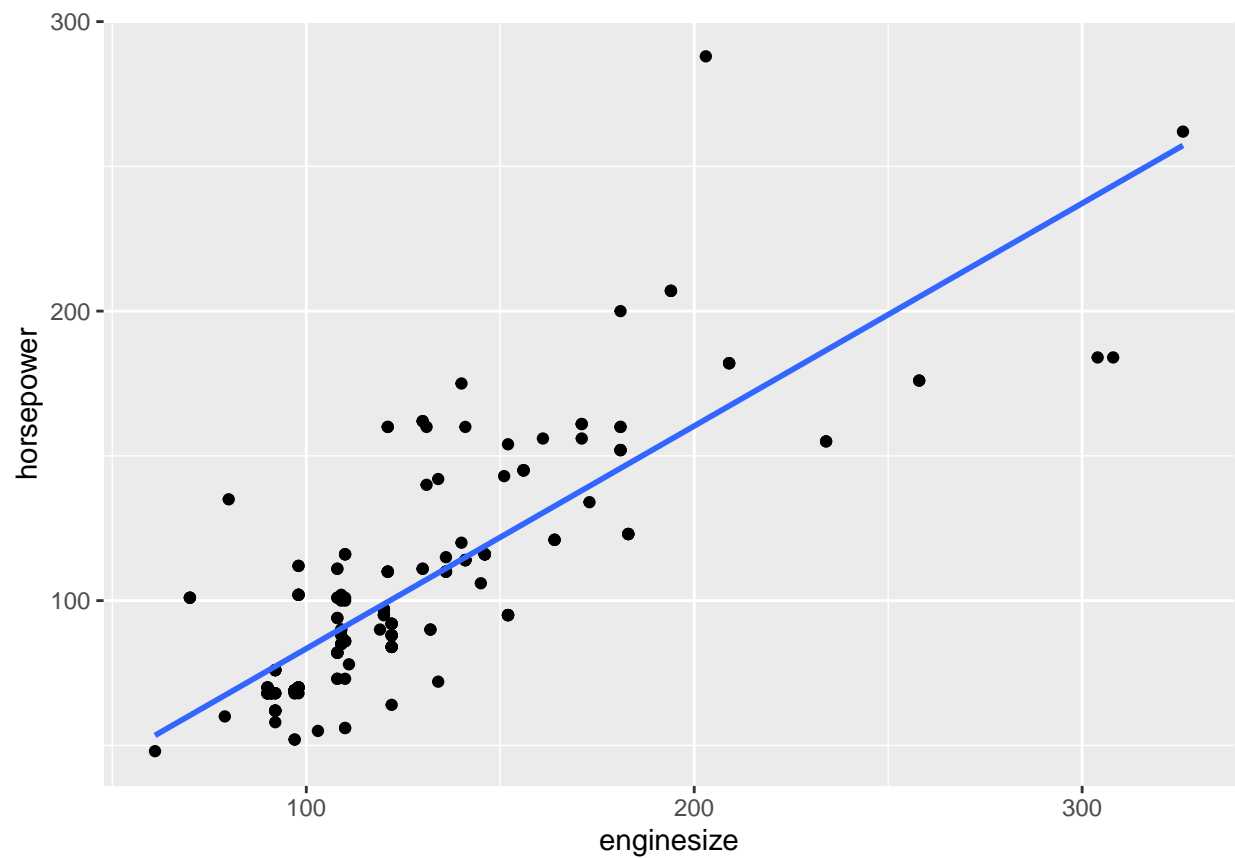
```
## [1] "Correlation between curbweight and citympg : -0.757413784505601"  
## [1] "x = enginesize , y = carwidth"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```

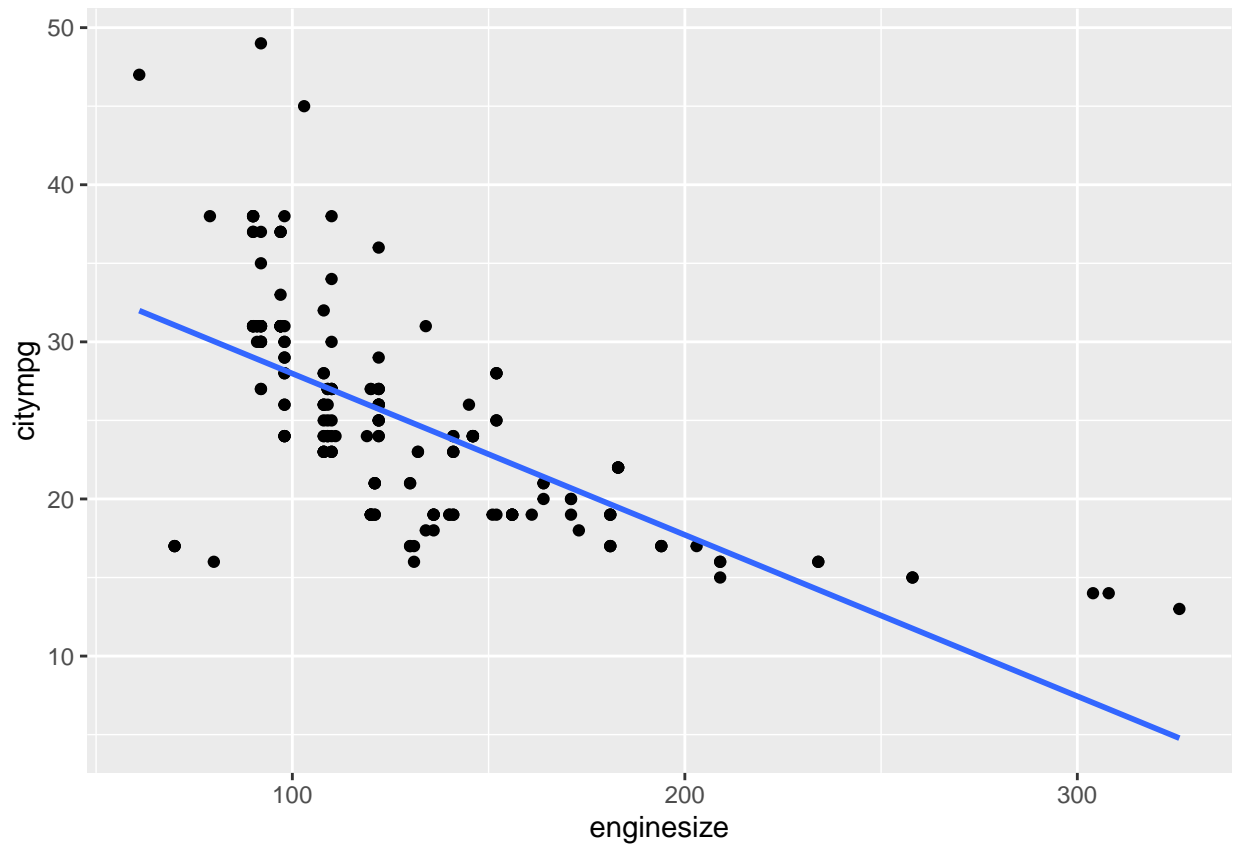
```
## [1] "Correlation between enginesize and carwidth : 0.735433404779637"  
## [1] "x = enginesize , y = curbweight"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```



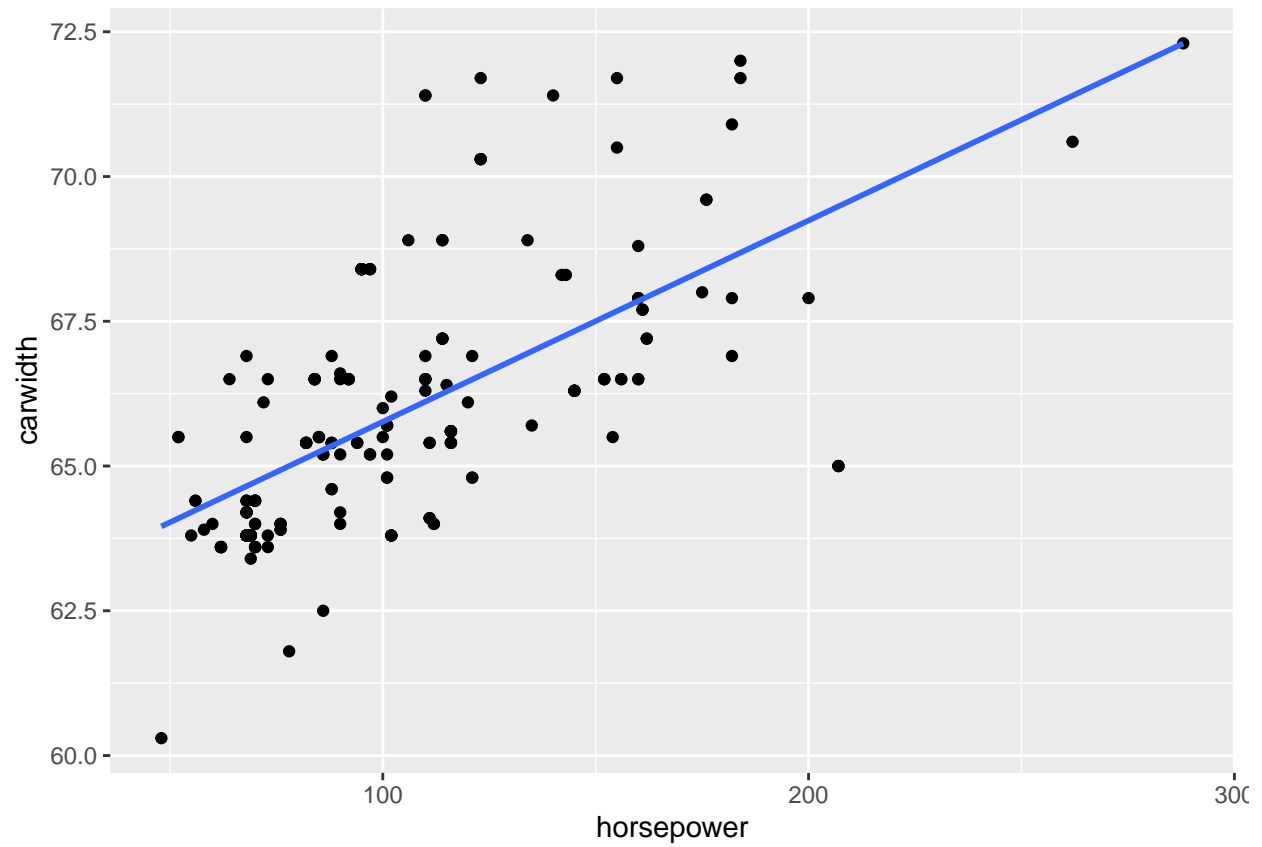
```
## [1] "Correlation between enginesize and curbweight : 0.850594073426277"  
## [1] "x = enginesize , y = horsepower"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```



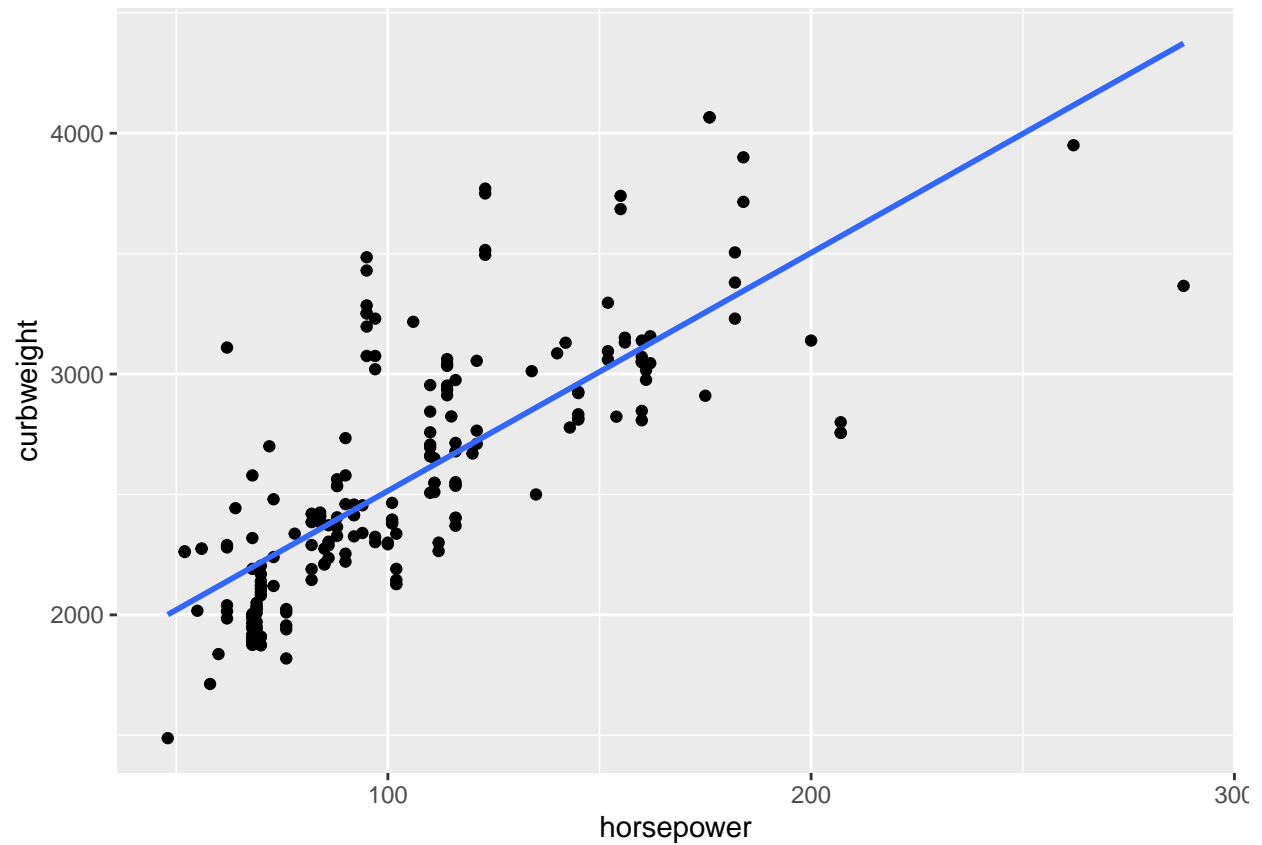
```
## [1] "Correlation between enginesize and horsepower : 0.80976865453773"  
## [1] "x = enginesize , y = citympg"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```



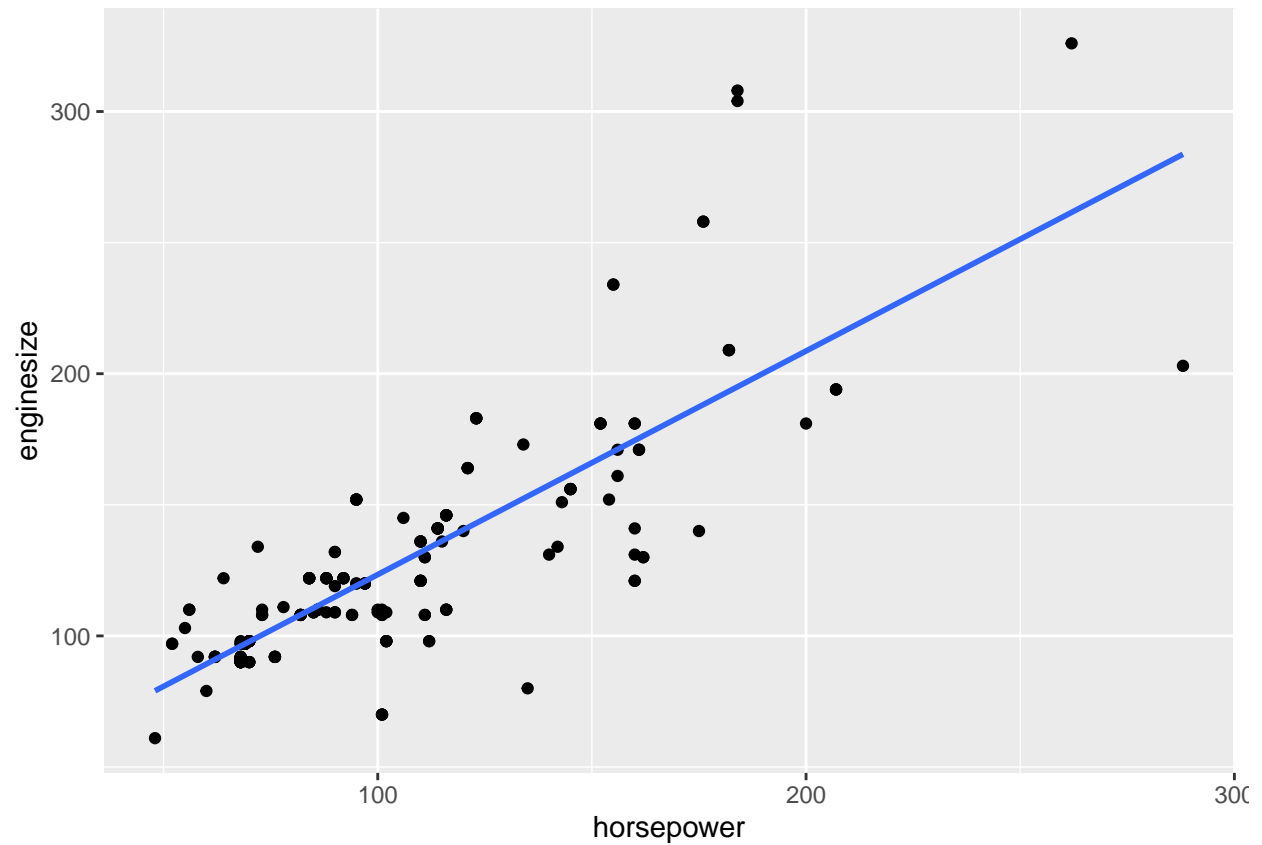
```
## [1] "Correlation between enginesize and citympg : -0.65365791631142"  
## [1] "x = horsepower , y = carwidth"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```



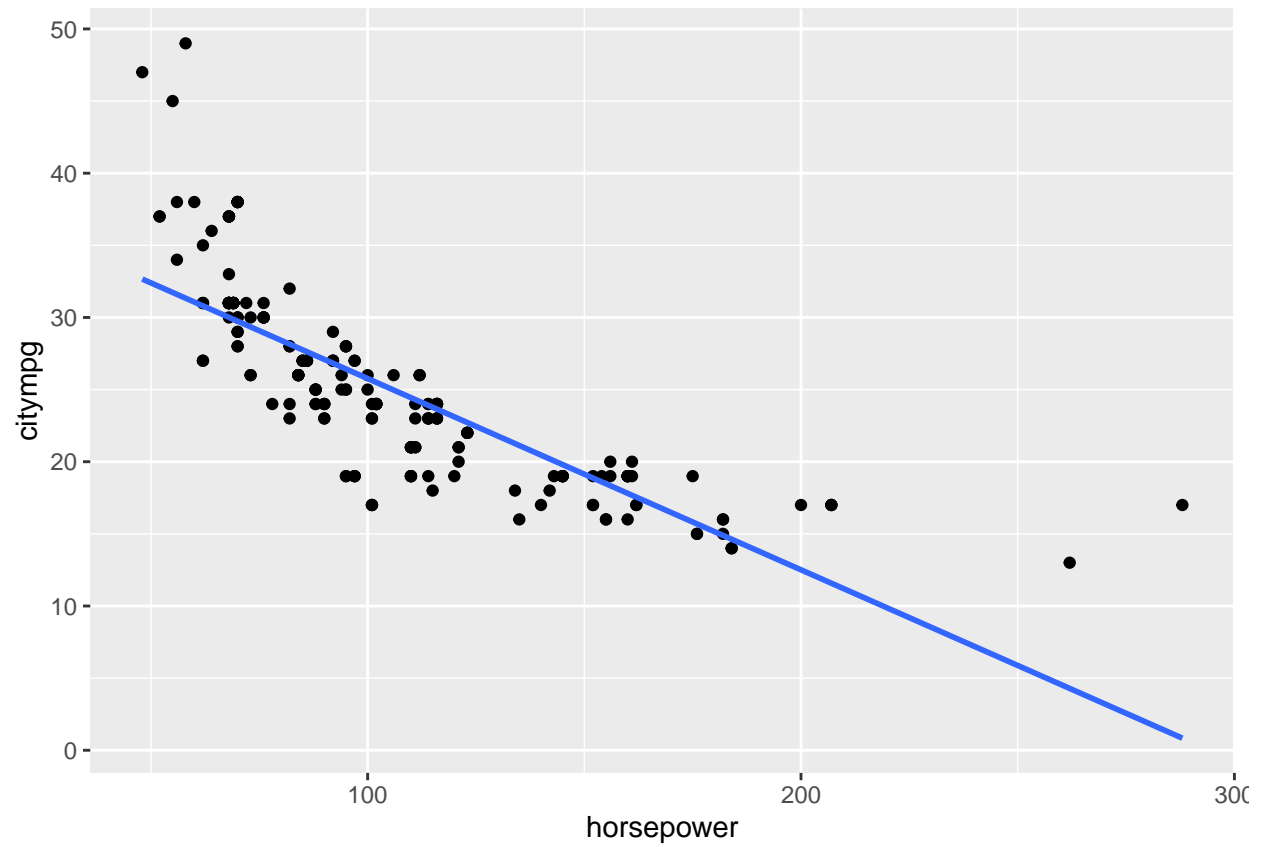
```
## [1] "Correlation between horsepower and carwidth : 0.640732075572292"  
## [1] "x = horsepower , y = curbweight"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```



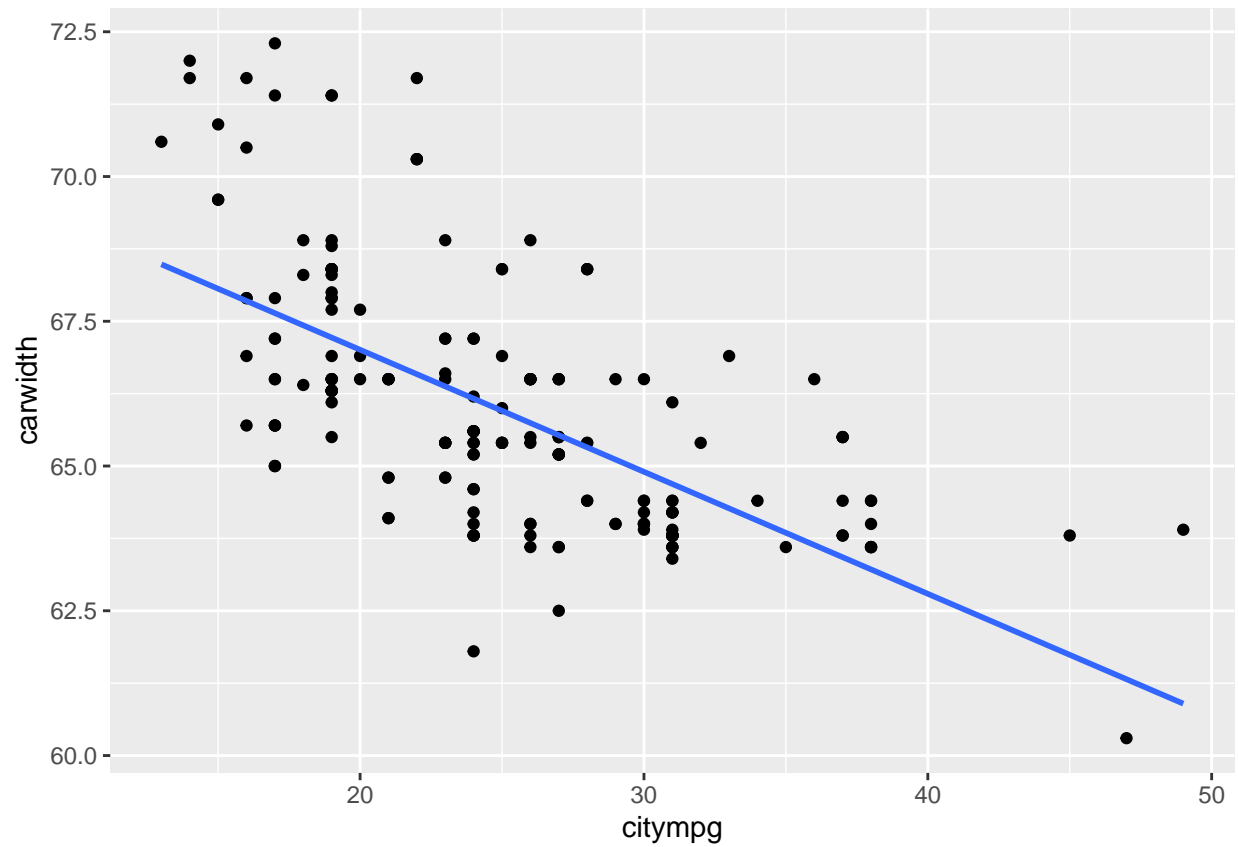
```
## [1] "Correlation between horsepower and curbweight : 0.750739251443481"  
## [1] "x = horsepower , y = enginesize"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```



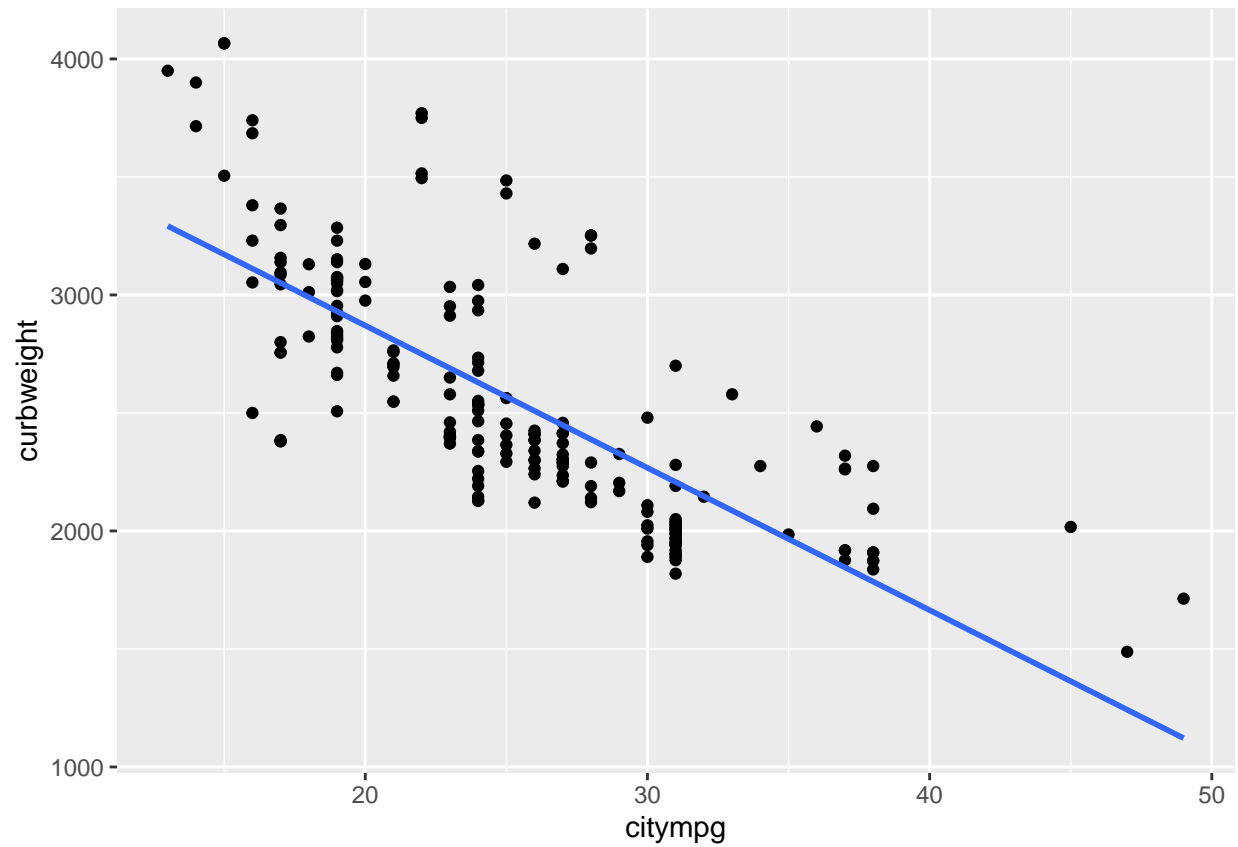
```
## [1] "Correlation between horsepower and enginesize : 0.80976865453773"  
## [1] "x = horsepower , y = citympg"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```



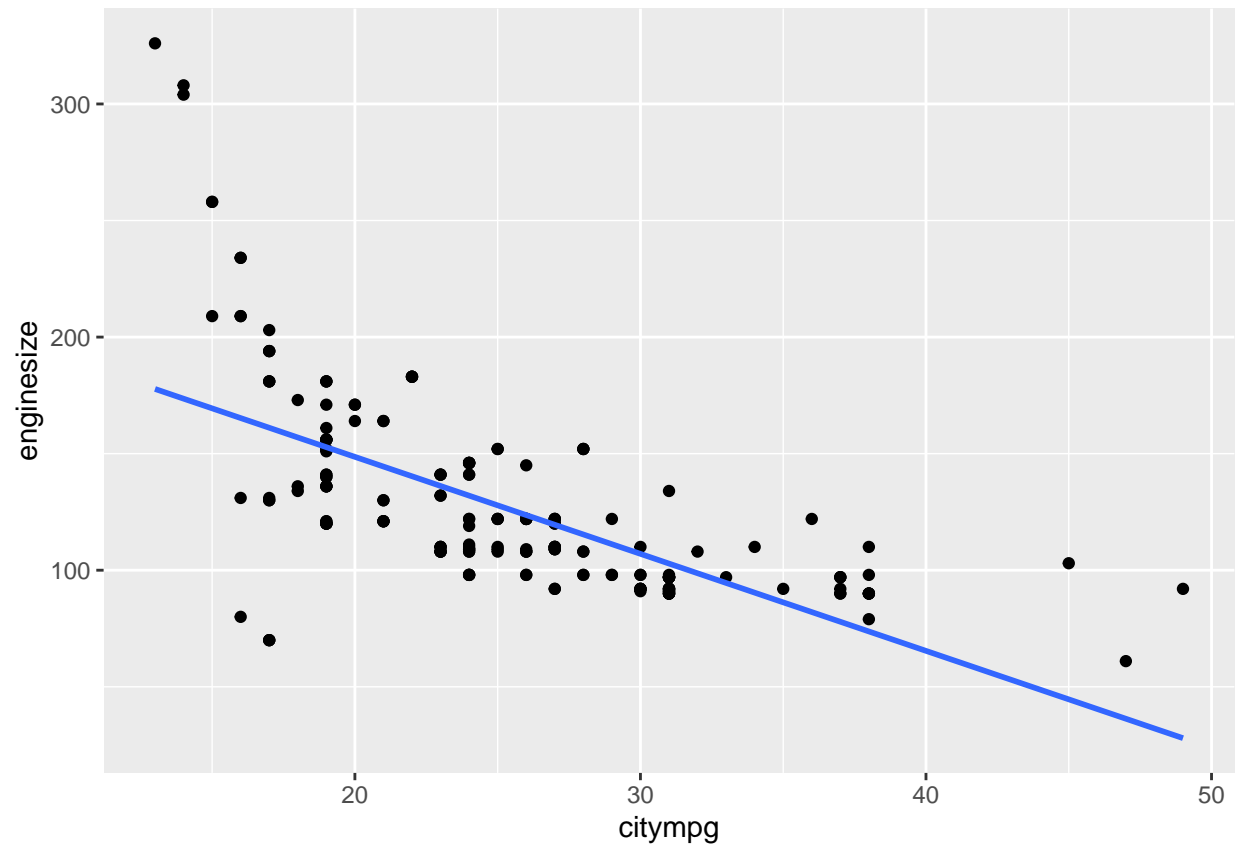
```
## [1] "Correlation between horsepower and citympg : -0.80145617566627"  
## [1] "x = citympg , y = carwidth"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```

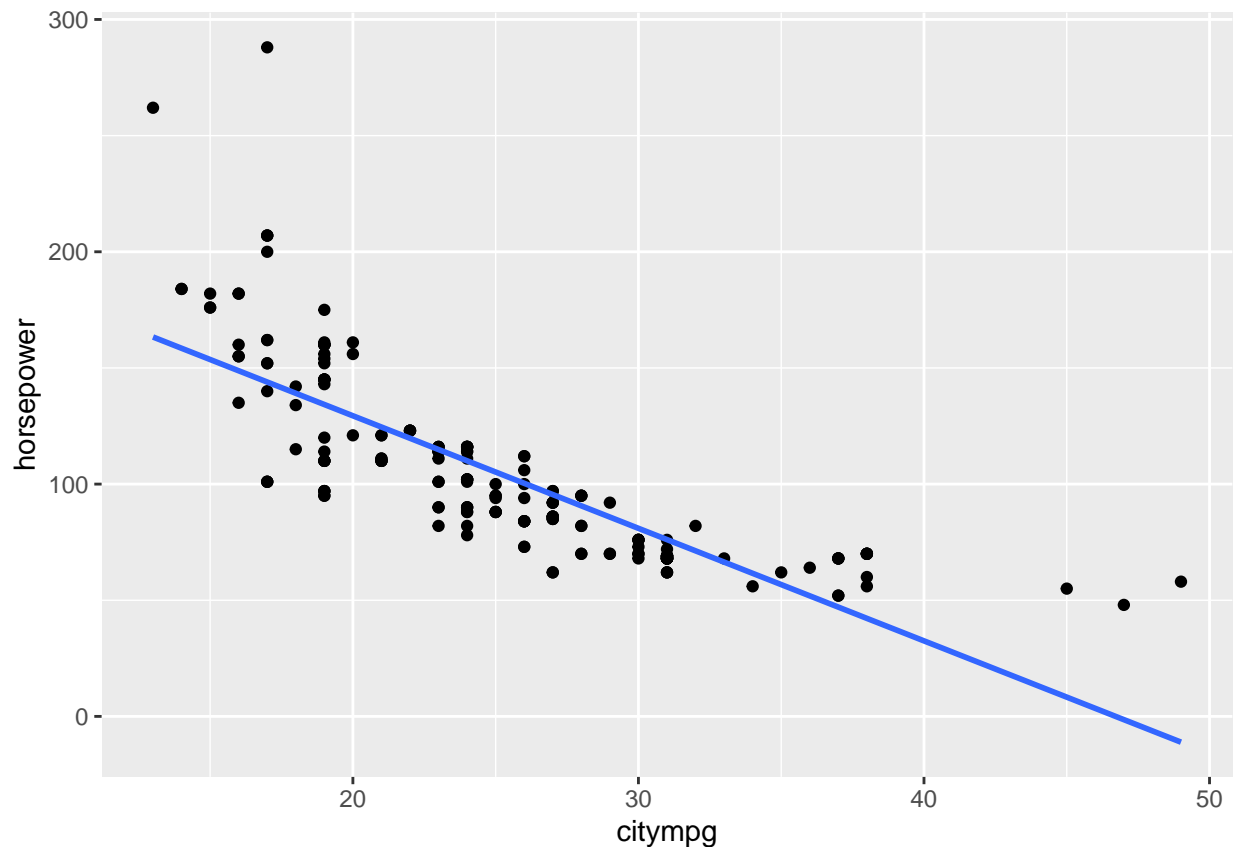
```
## [1] "Correlation between citympg and carwidth : -0.642704340710898"  
## [1] "x = citympg , y = curbweight"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```



```
## [1] "Correlation between citympg and curbweight : -0.757413784505601"  
## [1] "x = citympg , y = enginesize"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```



```
## [1] "Correlation between citympg and enginesize : -0.65365791631142"  
## [1] "x = citympg , y = horsepower"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```



```
## [1] "Correlation between citympg and horsepower : -0.80145617566627"
```

cars93 - Correlation between pairs of variables except price

```
for (var in c("MPG.city", "EngineSize", "Horsepower", "Fuel.tank.capacity", "Weight")) {
  for (var1 in c("MPG.city", "EngineSize", "Horsepower", "Fuel.tank.capacity", "Weight")) {

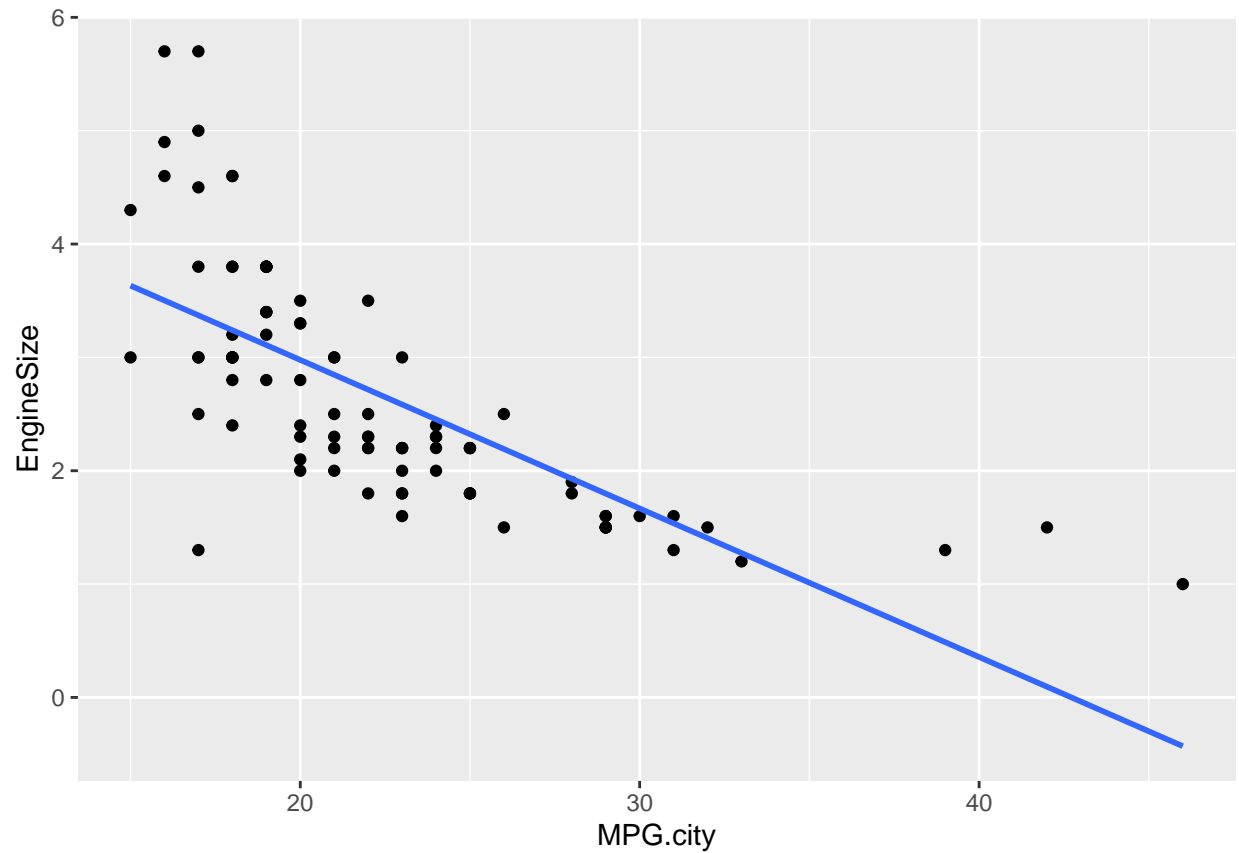
    if (var != var1) {
      print(paste("x =", var, ", y =", var1))

      p <- ggplot(cars93, aes_string(x = var, y = var1)) +
        geom_point() +
        geom_smooth(method = "lm", se = FALSE)
      print(p)

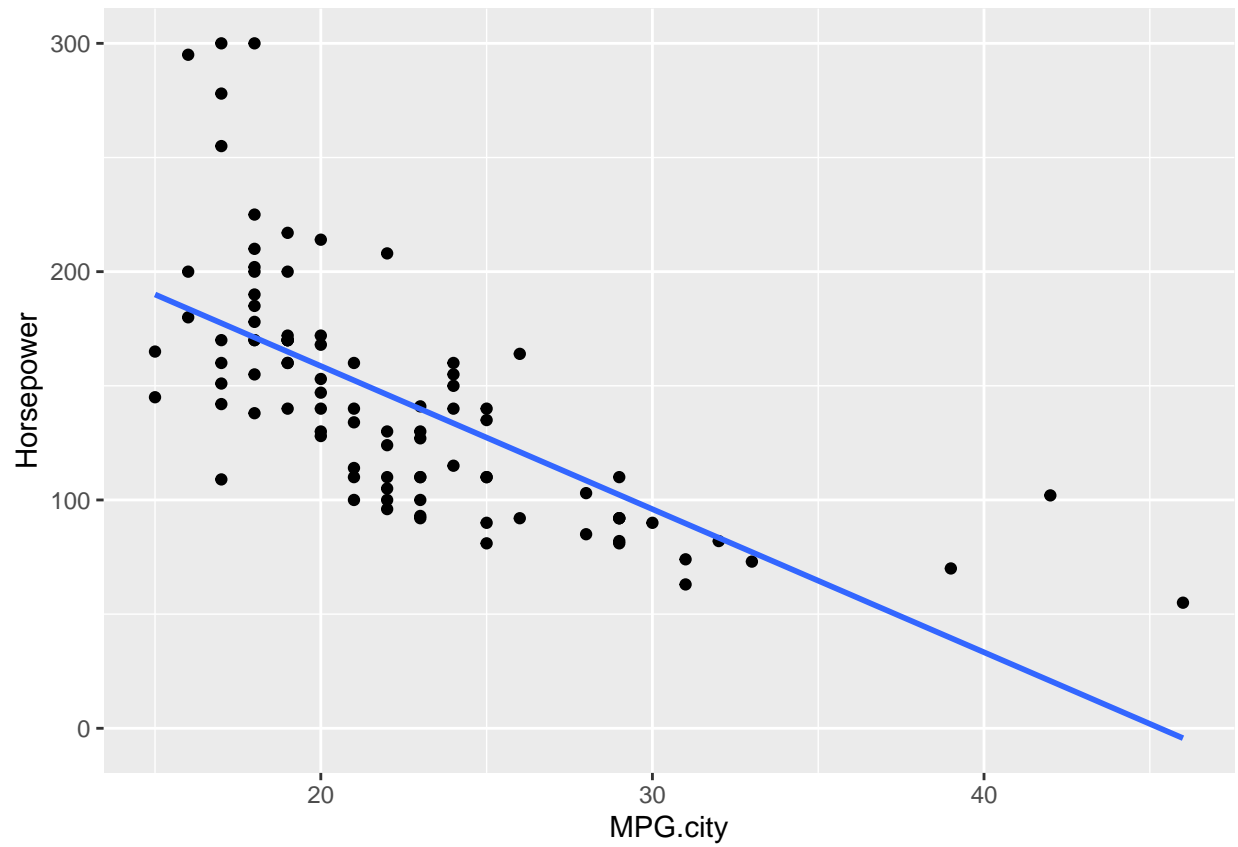
      correlation <- cor(cars93[[var1]], cars93[[var]])
      print(paste("Correlation between", var, "and", var1, ":", correlation))
    }
  }
}
```

```
## [1] "x = MPG.city , y = EngineSize"
```

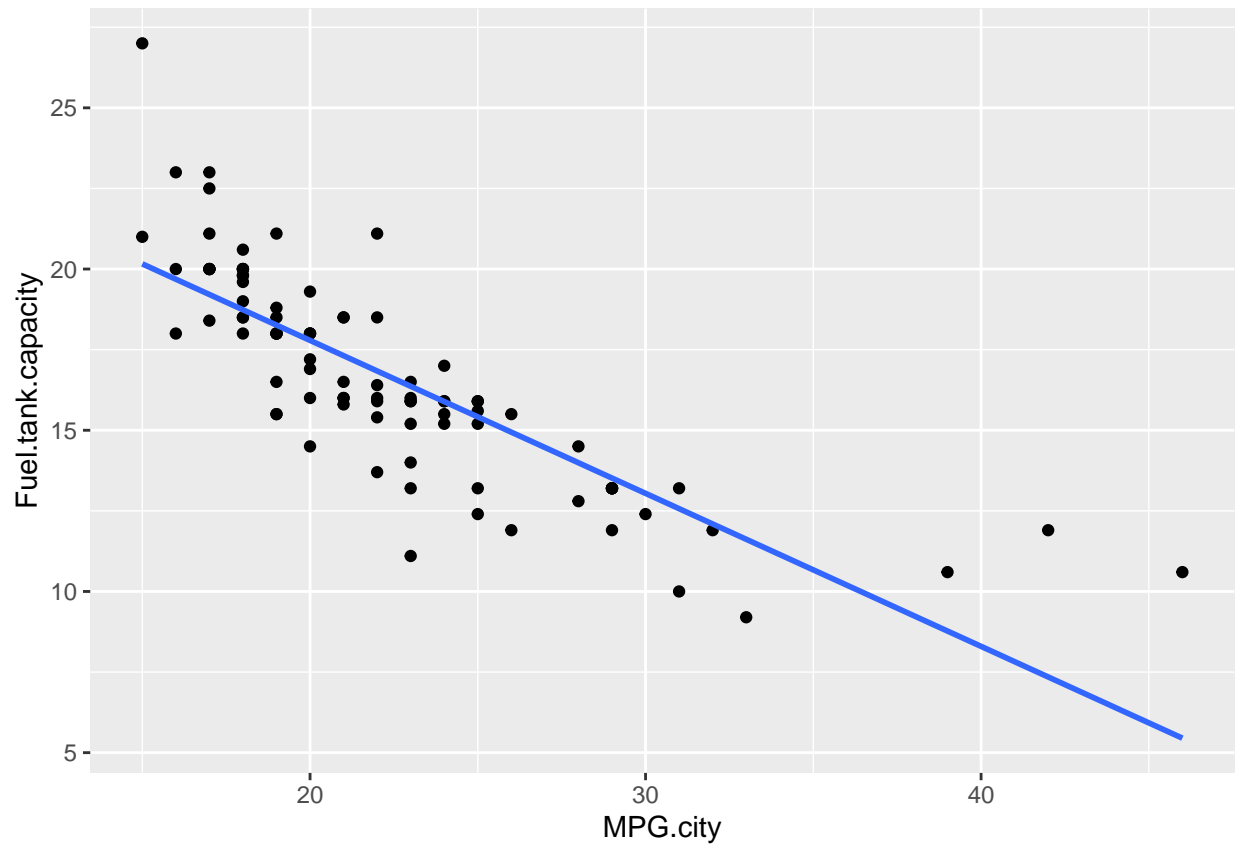
```
## 'geom_smooth()' using formula = 'y ~ x'
```



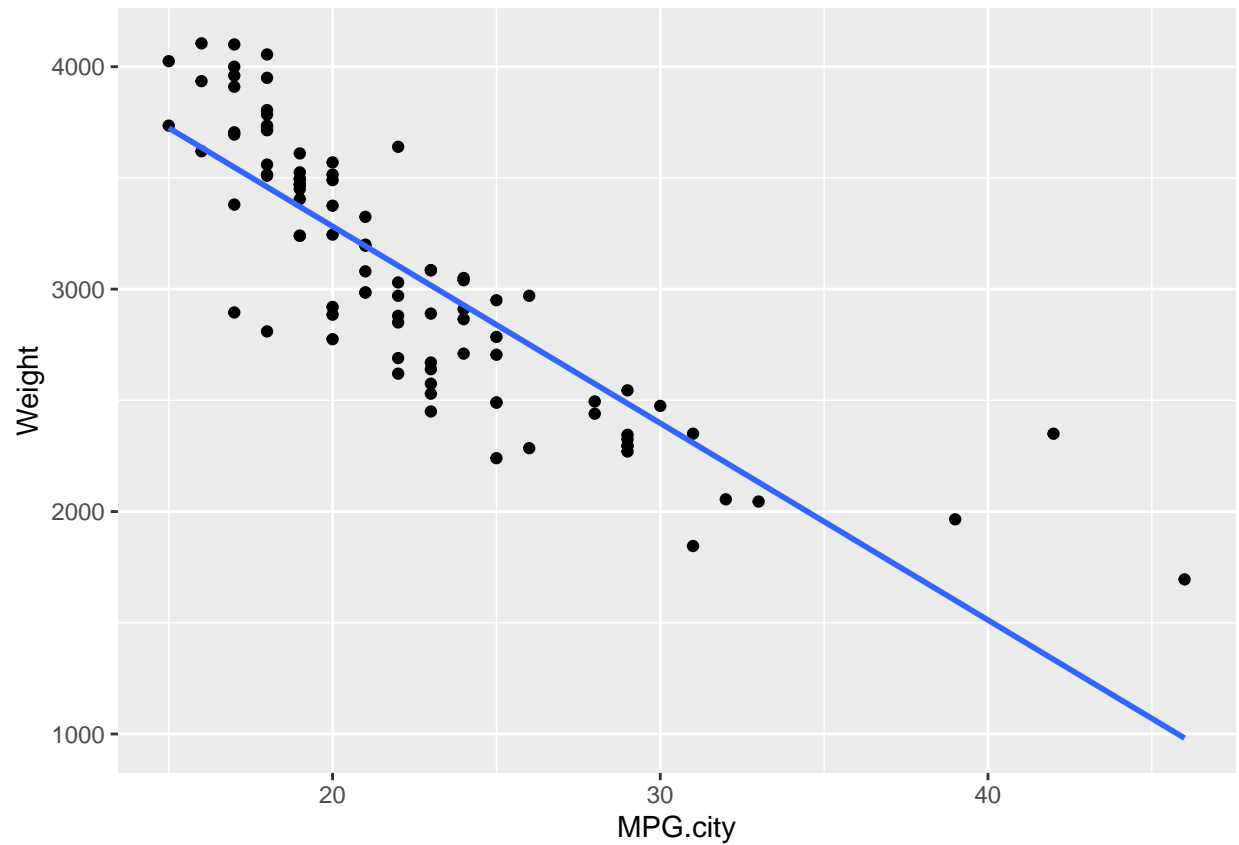
```
## [1] "Correlation between MPG.city and EngineSize : -0.710003161620369"  
## [1] "x = MPG.city , y = Horsepower"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```



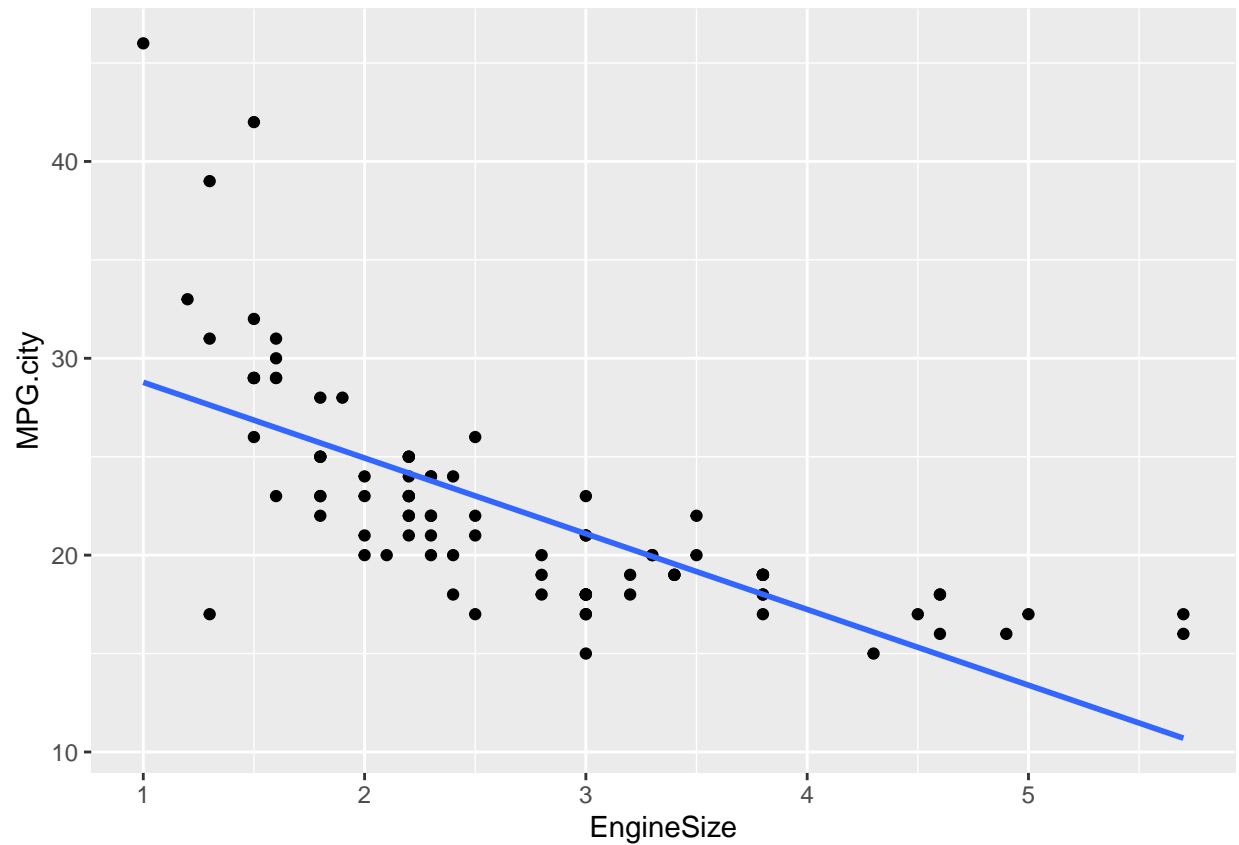
```
## [1] "Correlation between MPG.city and Horsepower : -0.672636150795724"  
## [1] "x = MPG.city , y = Fuel.tank.capacity"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```



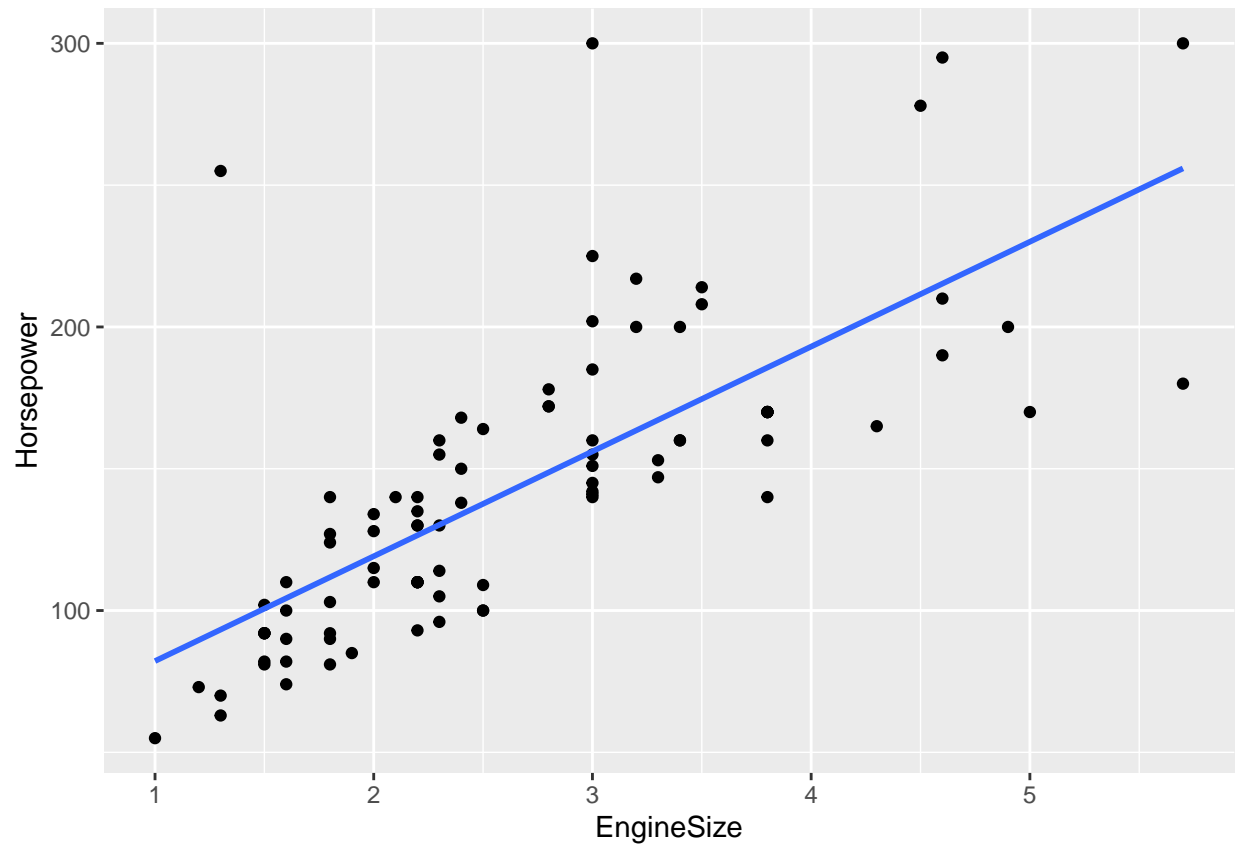
```
## [1] "Correlation between MPG.city and Fuel.tank.capacity : -0.813144416992242"  
## [1] "x = MPG.city , y = Weight"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```



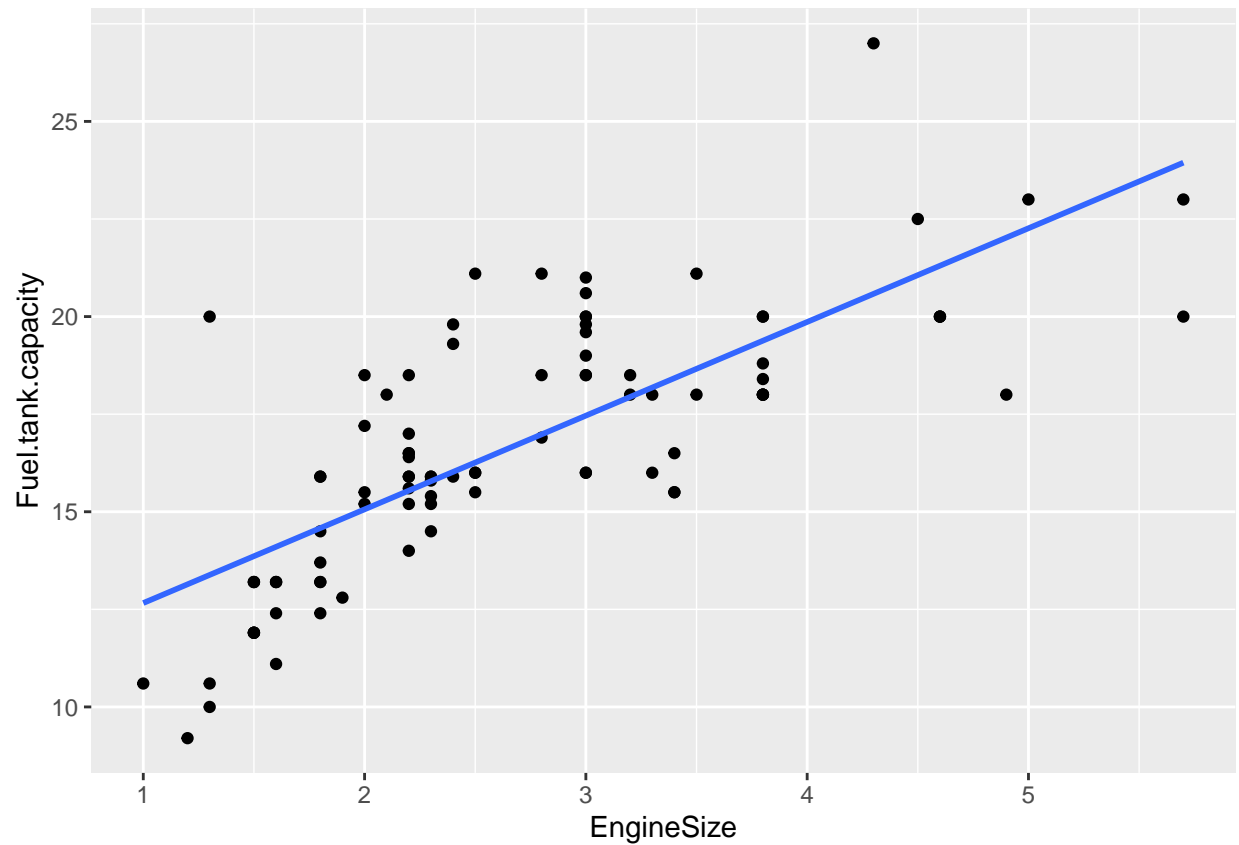
```
## [1] "Correlation between MPG.city and Weight : -0.84313854799682"  
## [1] "x = EngineSize , y = MPG.city"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```

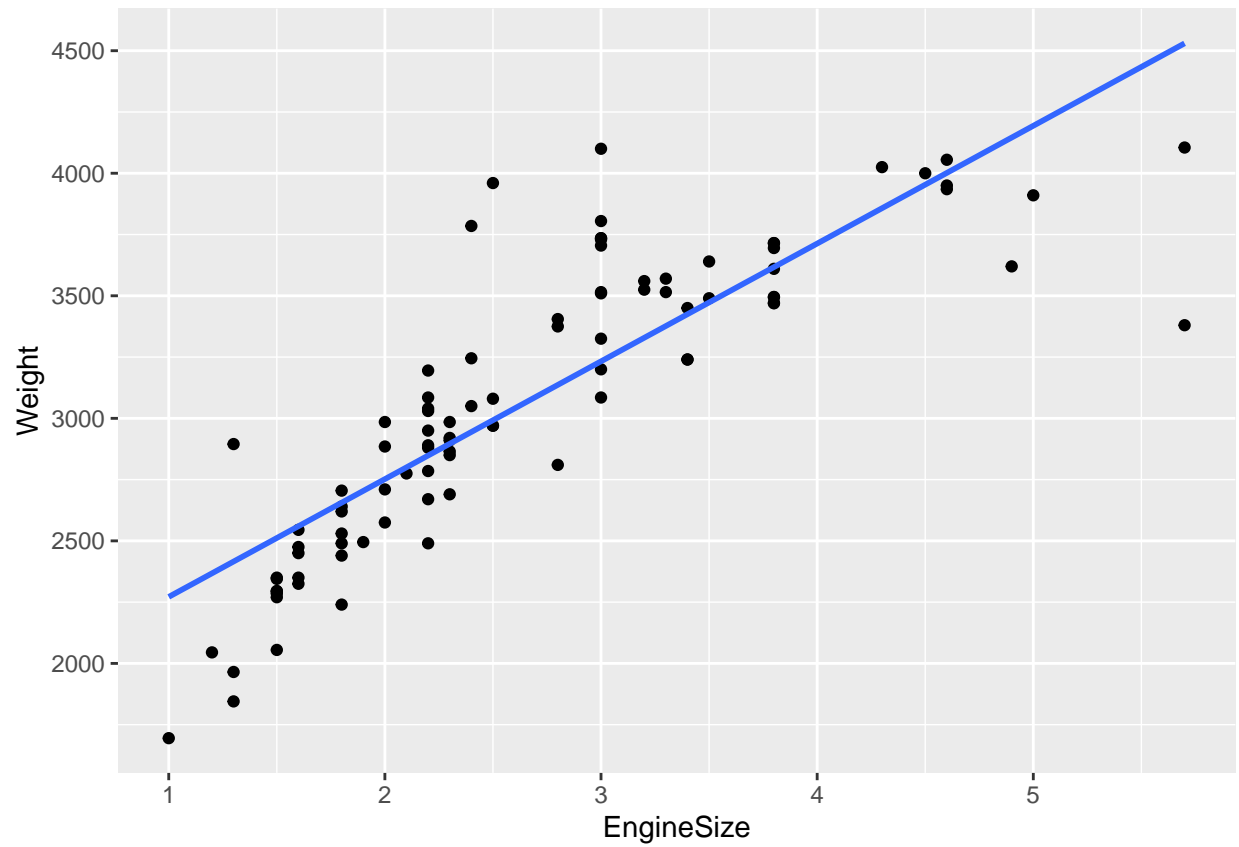
```
## [1] "Correlation between EngineSize and MPG.city : -0.710003161620369"  
## [1] "x = EngineSize , y = Horsepower"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```



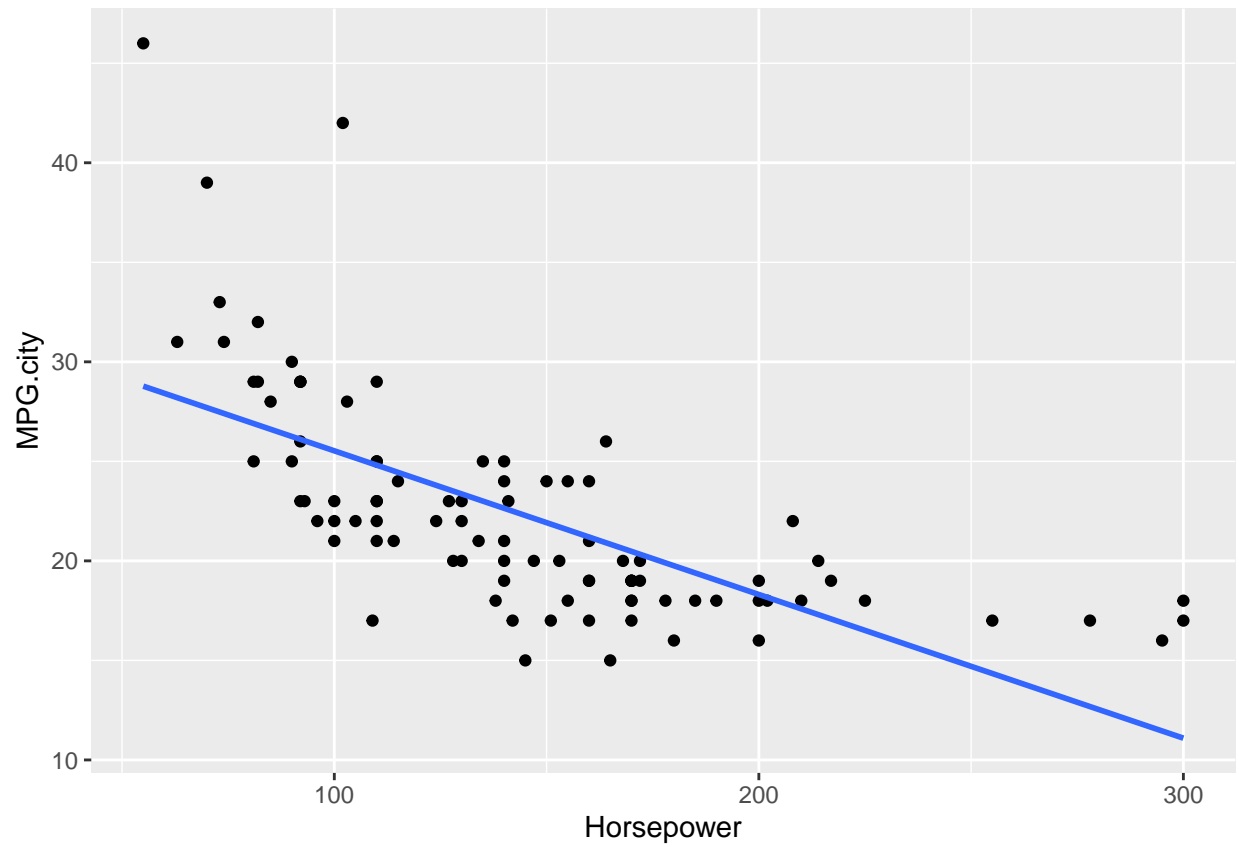
```
## [1] "Correlation between EngineSize and Horsepower : 0.732119730417387"  
## [1] "x = EngineSize , y = Fuel.tank.capacity"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```



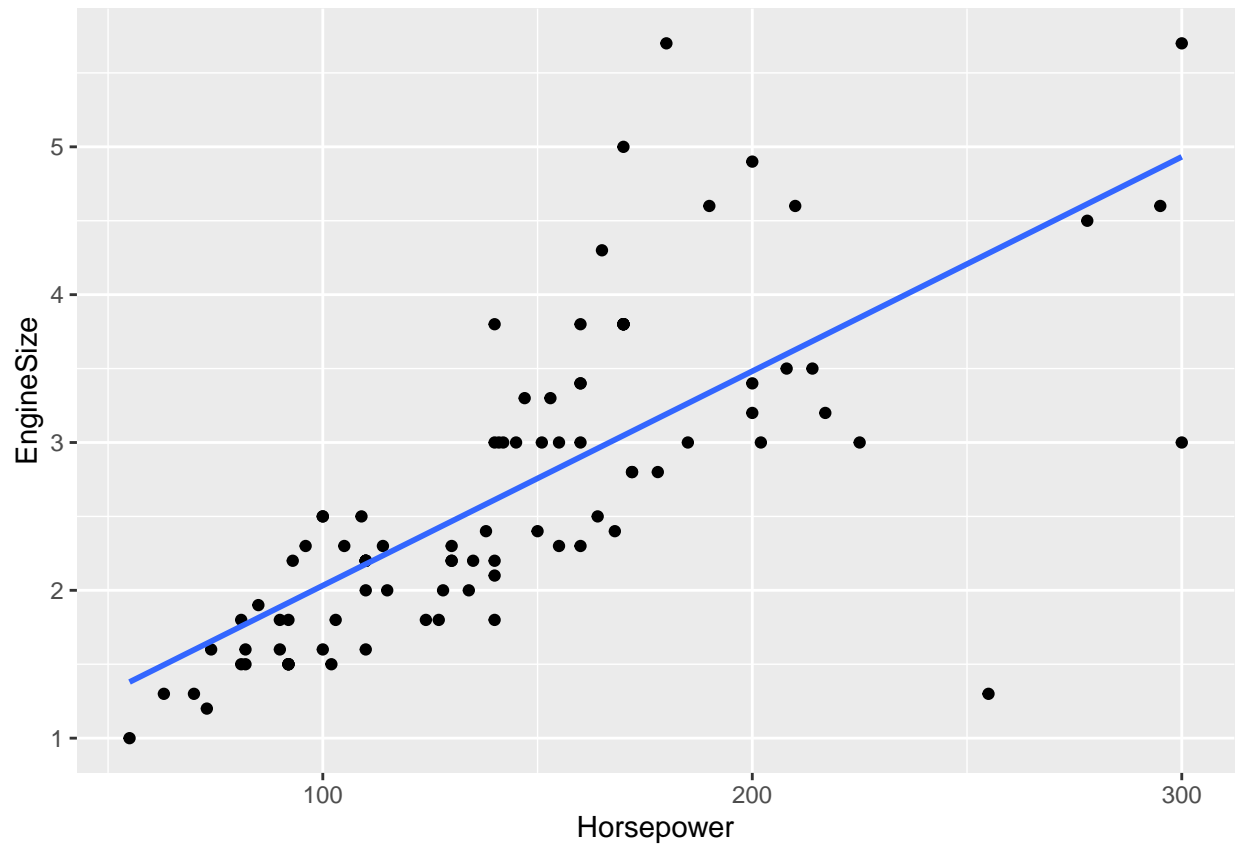
```
## [1] "Correlation between EngineSize and Fuel.tank.capacity : 0.759306224903401"  
## [1] "x = EngineSize , y = Weight"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```



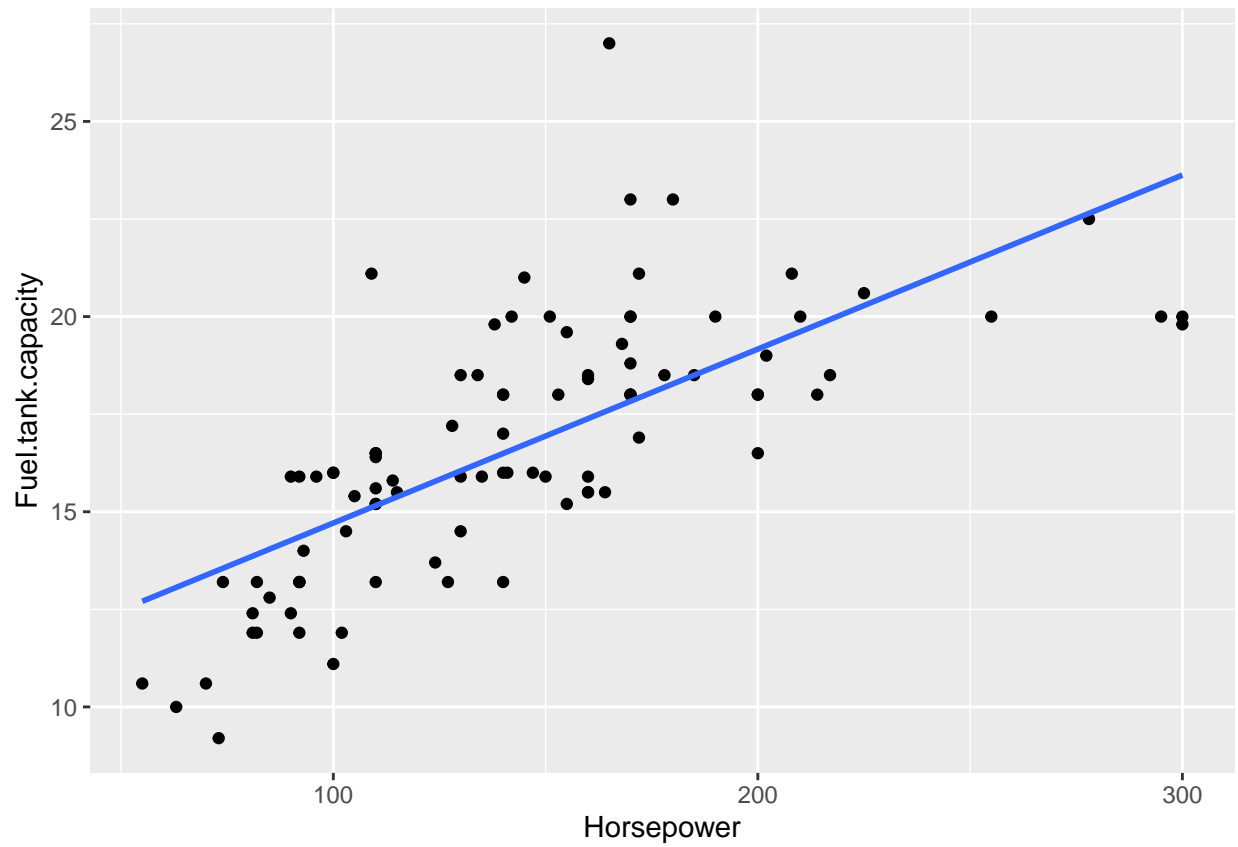
```
## [1] "Correlation between EngineSize and Weight : 0.845075335037262"  
## [1] "x = Horsepower , y = MPG.city"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```



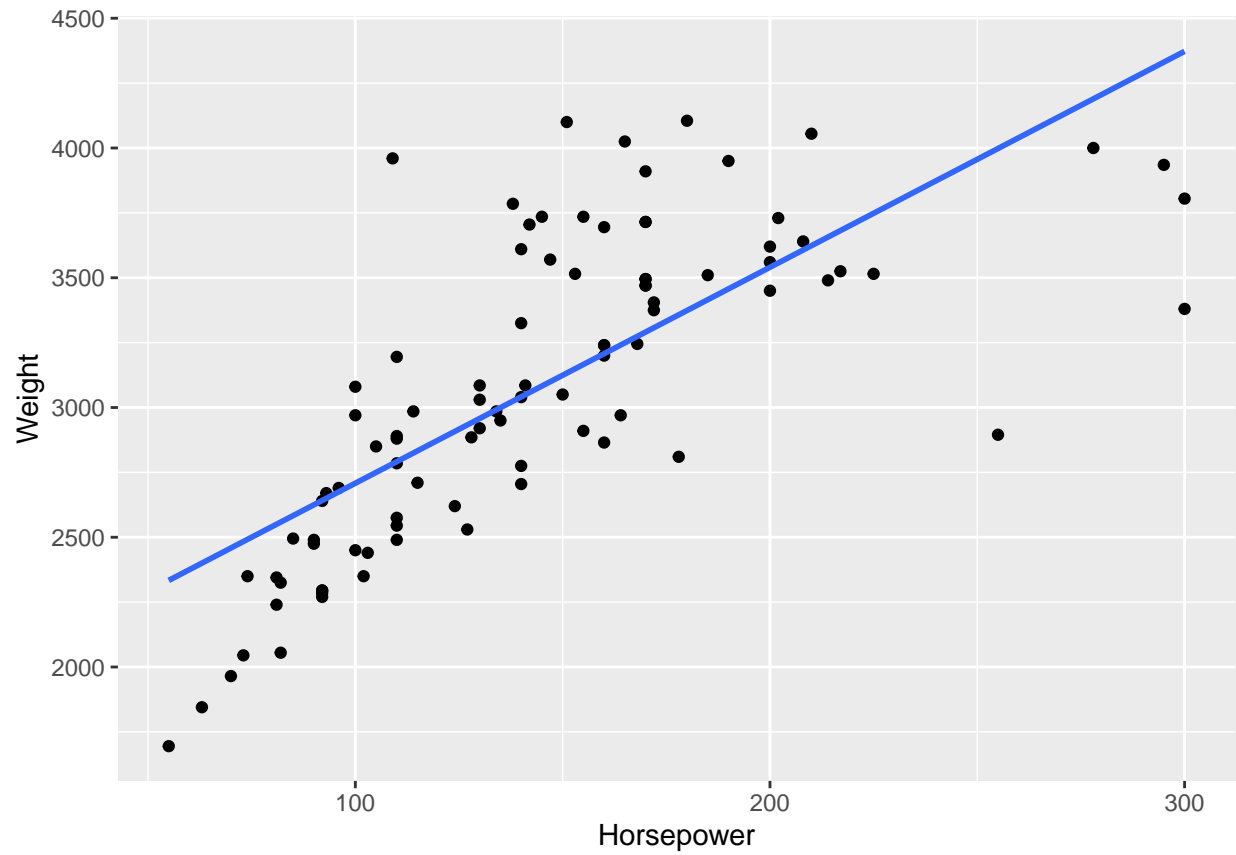
```
## [1] "Correlation between Horsepower and MPG.city : -0.672636150795724"  
## [1] "x = Horsepower , y = EngineSize"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```



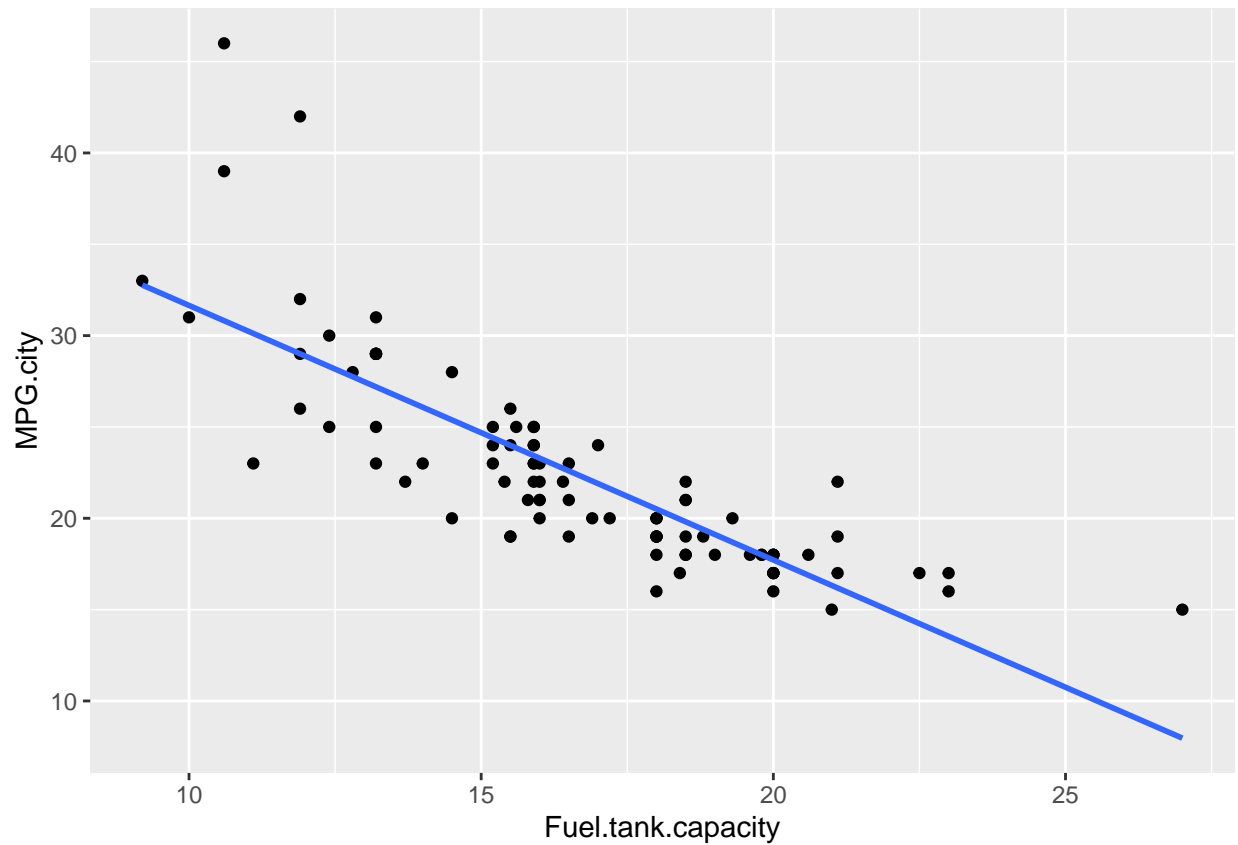
```
## [1] "Correlation between Horsepower and EngineSize : 0.732119730417387"  
## [1] "x = Horsepower , y = Fuel.tank.capacity"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```



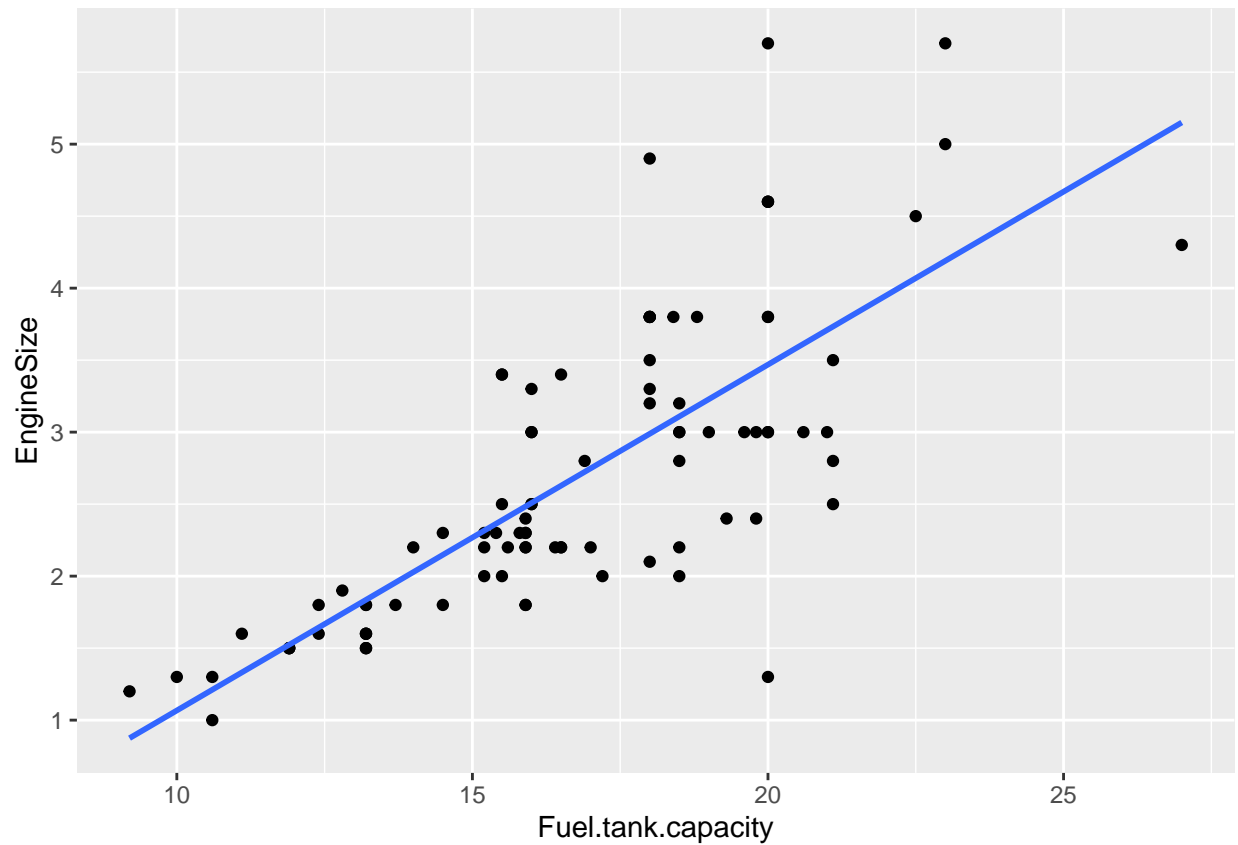
```
## [1] "Correlation between Horsepower and Fuel.tank.capacity : 0.711790317259719"  
## [1] "x = Horsepower , y = Weight"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```



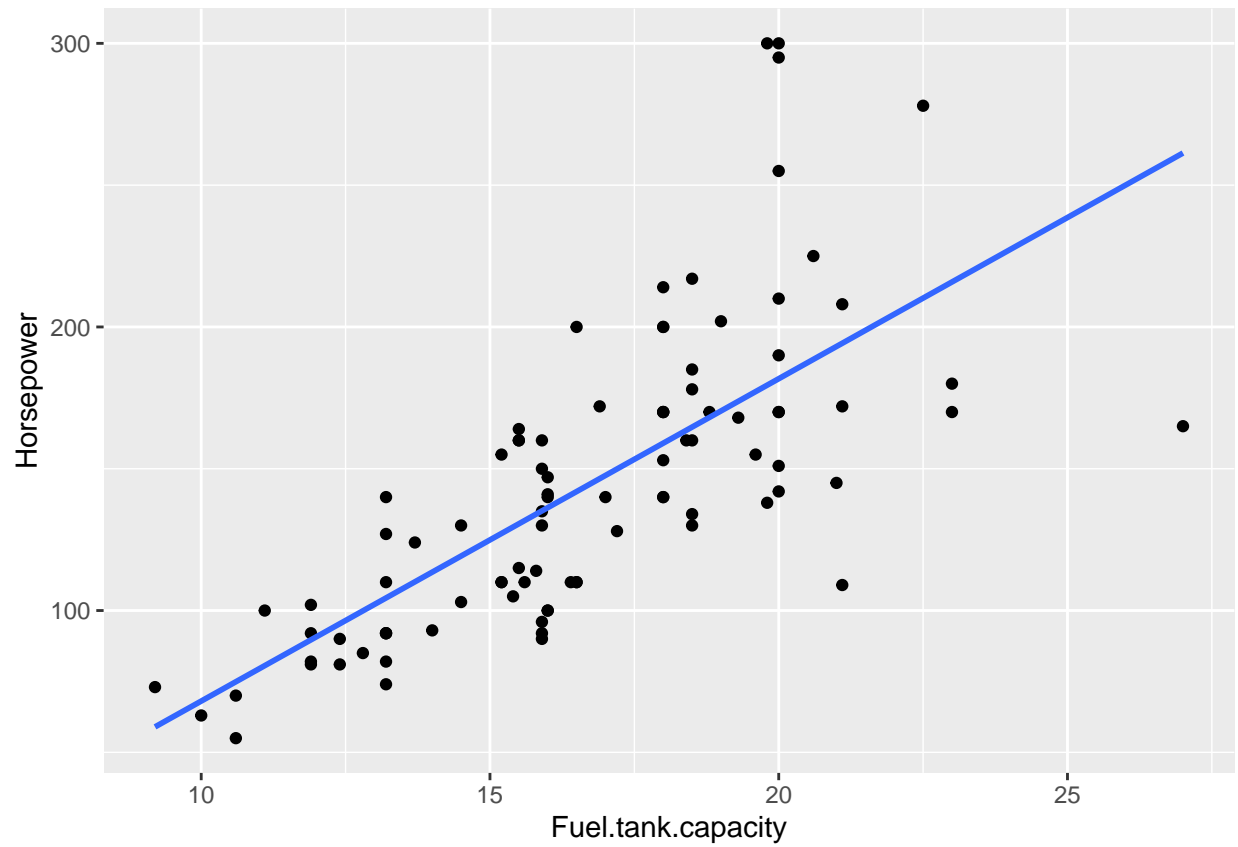
```
## [1] "Correlation between Horsepower and Weight : 0.738797515540372"  
## [1] "x = Fuel.tank.capacity , y = MPG.city"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```

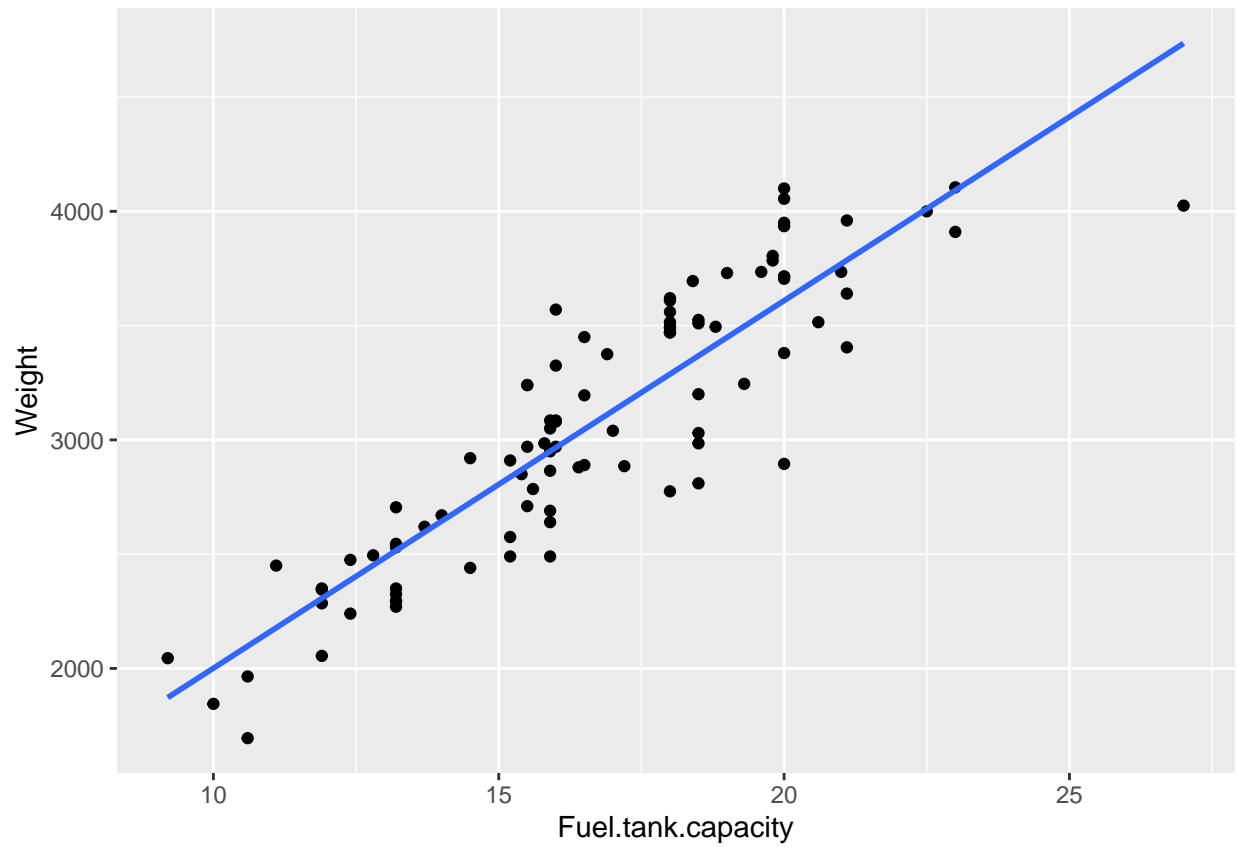
```
## [1] "Correlation between Fuel.tank.capacity and MPG.city : -0.813144416992242"  
## [1] "x = Fuel.tank.capacity , y = EngineSize"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```



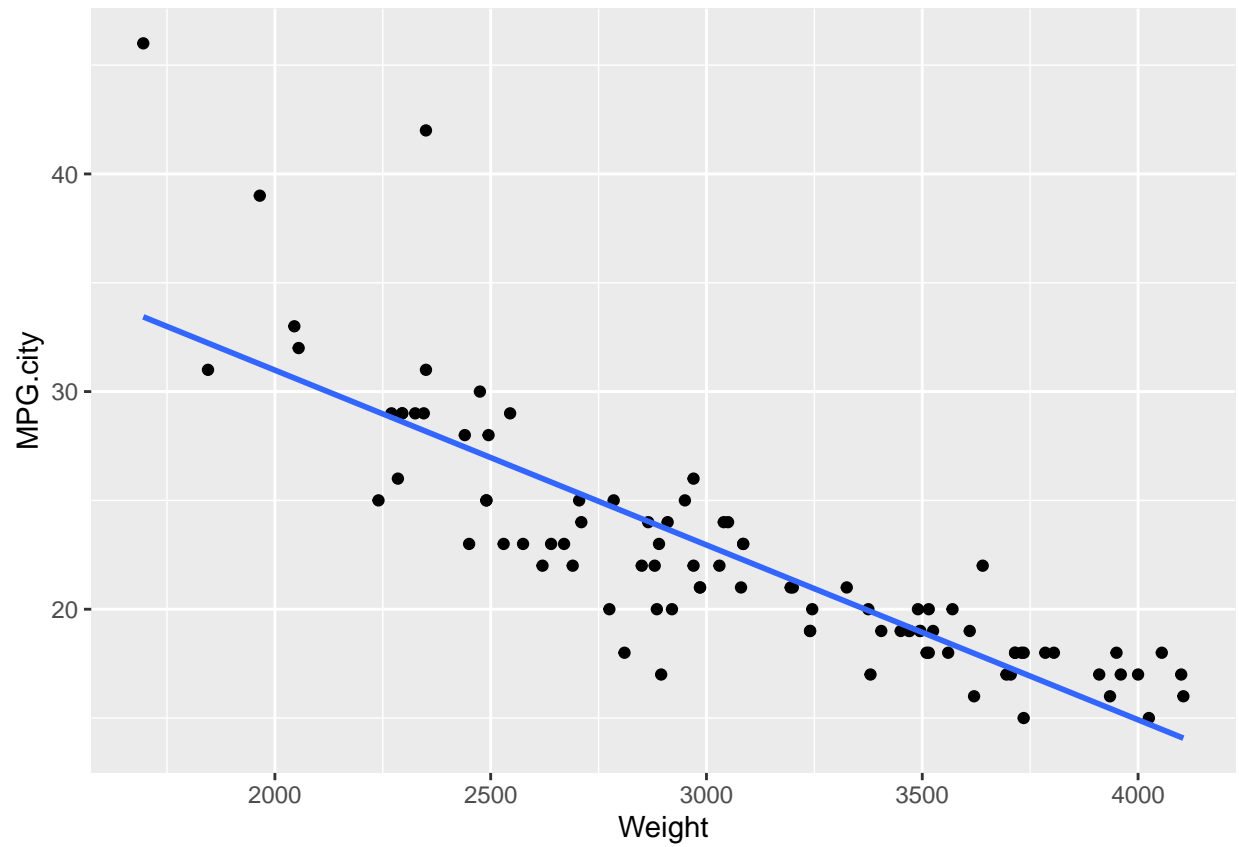
```
## [1] "Correlation between Fuel.tank.capacity and EngineSize : 0.759306224903401"  
## [1] "x = Fuel.tank.capacity , y = Horsepower"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```



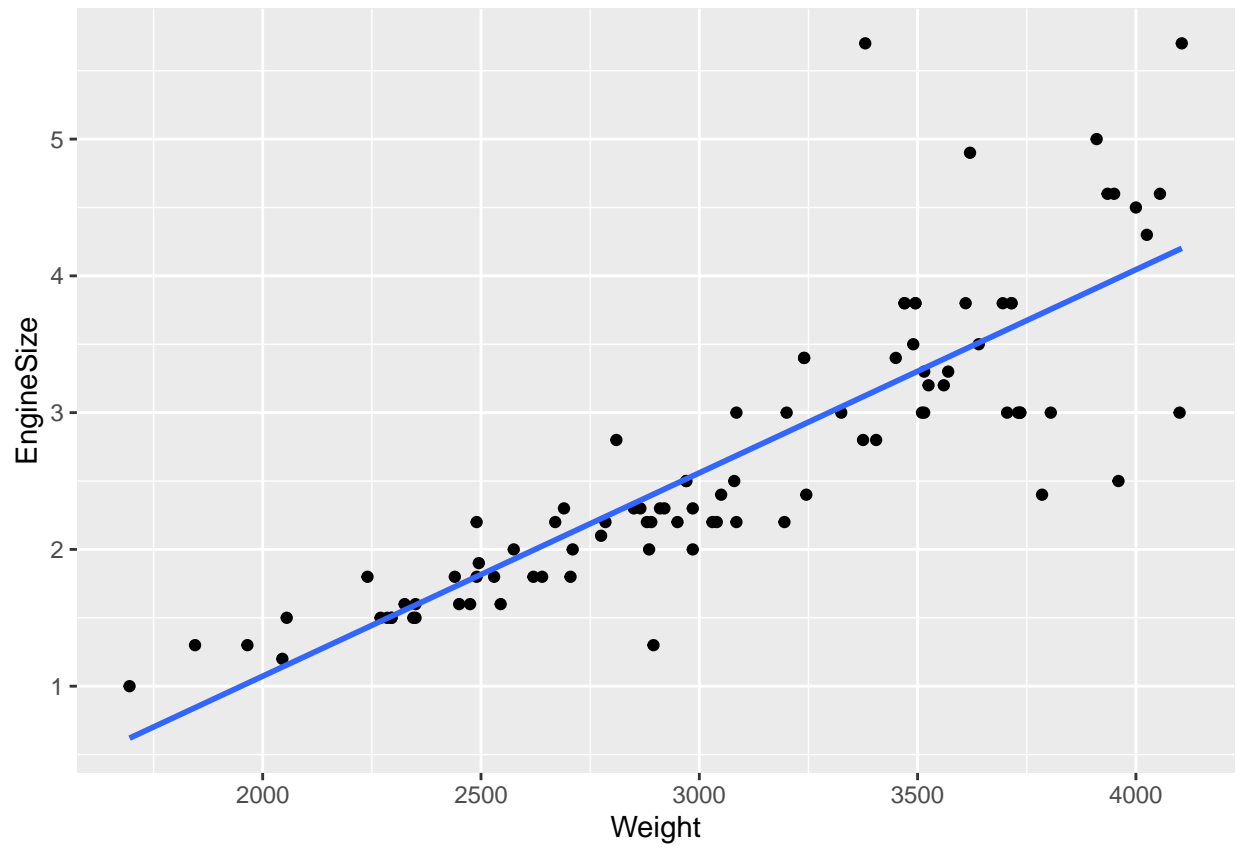
```
## [1] "Correlation between Fuel.tank.capacity and Horsepower : 0.711790317259719"  
## [1] "x = Fuel.tank.capacity , y = Weight"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```



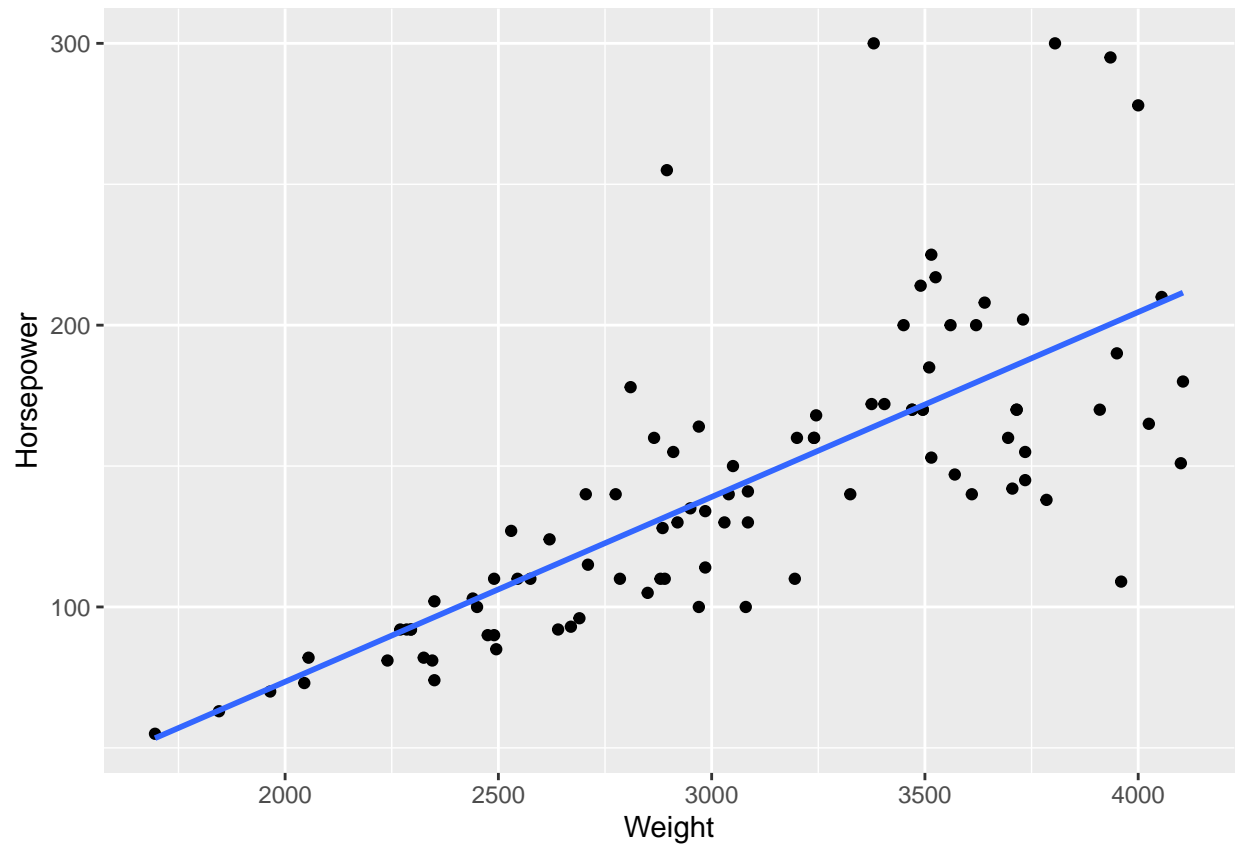
```
## [1] "Correlation between Fuel.tank.capacity and Weight : 0.894018054491314"  
## [1] "x = Weight , y = MPG.city"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```



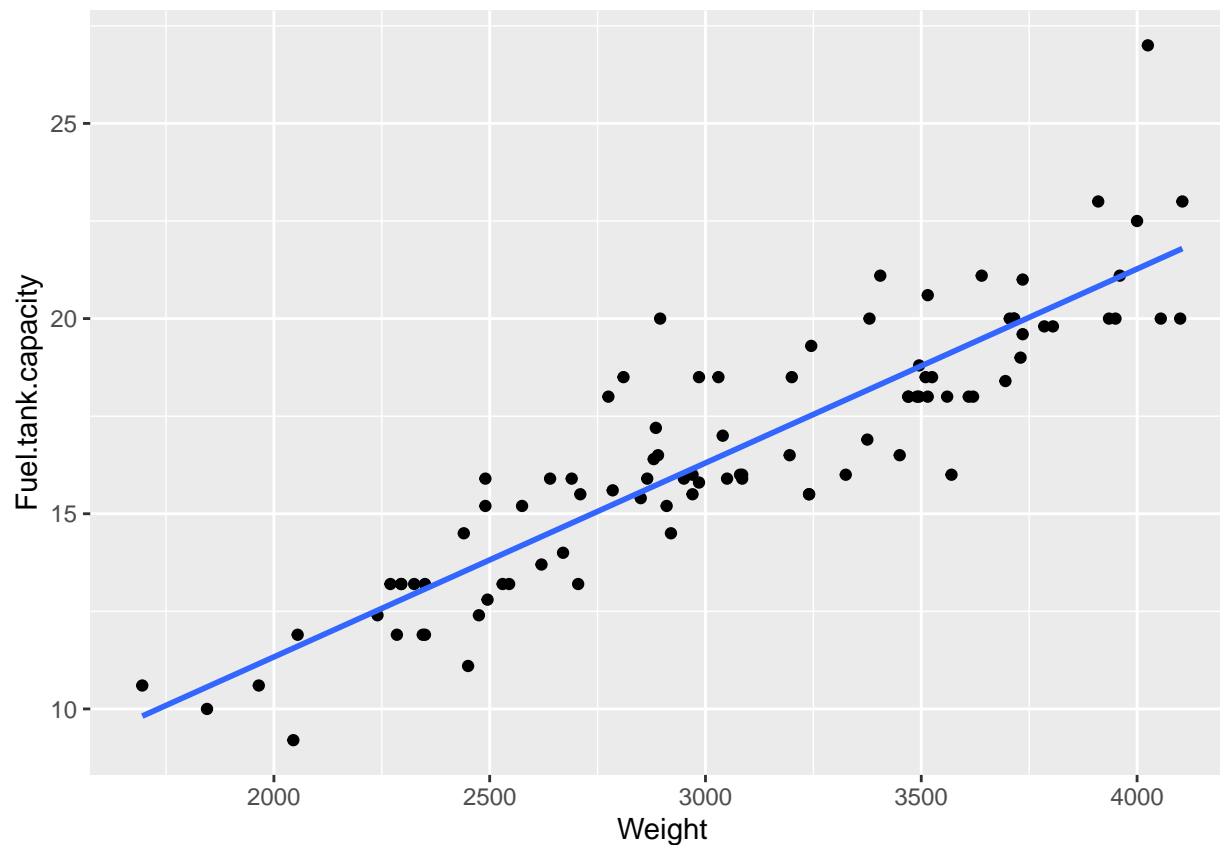
```
## [1] "Correlation between Weight and MPG.city : -0.84313854799682"  
## [1] "x = Weight , y = EngineSize"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```



```
## [1] "Correlation between Weight and EngineSize : 0.845075335037262"  
## [1] "x = Weight , y = Horsepower"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```



```
## [1] "Correlation between Weight and Horsepower : 0.738797515540372"  
## [1] "x = Weight , y = Fuel.tank.capacity"  
  
## 'geom_smooth()' using formula = 'y ~ x'
```



```
## [1] "Correlation between Weight and Fuel.tank.capacity : 0.894018054491314"
```

Linear regression model using common columns: mpgcity+highway, horsepower, enginesize, weight

```
model_assign2 <- lm(price ~ citympg + highwaympg + curbweight + enginesize + horsepower, data = cars_assign2)
summary(model_assign2)
```

```
##
## Call:
## lm(formula = price ~ citympg + highwaympg + curbweight + enginesize +
##     horsepower, data = cars_assign2)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -8942.7 -1685.5   -35.5   1322.6 13679.1
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -15803.388   4053.921  -3.898 0.000132 ***
## citympg      -40.841    177.895  -0.230 0.818657
## highwaympg    75.944    167.738   0.453 0.651218
## curbweight     4.654     1.101   4.226 3.63e-05 ***
## enginesize    83.295    13.586   6.131 4.61e-09 ***
## horsepower    51.005    13.878   3.675 0.000305 ***
```



```
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3486 on 199 degrees of freedom
## Multiple R-squared:  0.8142, Adjusted R-squared:  0.8095
## F-statistic: 174.4 on 5 and 199 DF,  p-value: < 2.2e-16
```

```
summary(model_assign2)
```

```
##
## Call:
## lm(formula = price ~ citympg + highwaympg + curbweight + enginesize +
##     horsepower, data = cars_assign)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -8942.7 -1685.5   -35.5   1322.6 13679.1
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -15803.388   4053.921  -3.898 0.000132 ***
## citympg      -40.841    177.895   -0.230 0.818657
## highwaympg    75.944    167.738    0.453 0.651218
## curbweight     4.654     1.101    4.226 3.63e-05 ***
## enginesize    83.295     13.586    6.131 4.61e-09 ***
## horsepower    51.005     13.878    3.675 0.000305 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3486 on 199 degrees of freedom
## Multiple R-squared:  0.8142, Adjusted R-squared:  0.8095
## F-statistic: 174.4 on 5 and 199 DF,  p-value: < 2.2e-16
```

The main model

```
model_assign3 <- lm(price ~ curbweight + enginesize + horsepower, data = cars_assign)
summary(model_assign3)
```

```
##
## Call:
## lm(formula = price ~ curbweight + enginesize + horsepower, data = cars_assign)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
##  -9003   -1701    -24    1340   13760
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -1.346e+04  1.333e+03 -10.100 < 2e-16 ***
## curbweight   4.263e+00  9.065e-01  4.702 4.78e-06 ***
## enginesize   8.488e+01  1.276e+01  6.651 2.69e-10 ***
## horsepower   4.875e+01  1.070e+01  4.557 8.99e-06 ***
```

```
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3473 on 201 degrees of freedom
## Multiple R-squared:  0.8138, Adjusted R-squared:  0.811
## F-statistic: 292.9 on 3 and 201 DF,  p-value: < 2.2e-16

model_assign_all <- lm(price ~ wheelbase + carlength + carwidth + carheight + curbweight + enginesize +
summary(model_assign_all)

##
## Call:
## lm(formula = price ~ wheelbase + carlength + carwidth + carheight +
##     curbweight + enginesize + horsepower + peakrpm + citympg +
##     highwaympg, data = cars_assign)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -8806.6 -1643.9  -146.4   1355.1  13900.7
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -6.787e+04  1.486e+04  -4.566 8.82e-06 ***
## wheelbase    4.487e+01  1.046e+02   0.429  0.66844
## carlength   -8.311e+01  5.812e+01  -1.430  0.15433
## carwidth     6.069e+02  2.567e+02   2.365  0.01904 *
## carheight    2.369e+02  1.413e+02   1.677  0.09524 .
## curbweight   3.916e+00  1.670e+00   2.345  0.02004 *
## enginesize    1.009e+02  1.413e+01   7.142 1.80e-11 ***
## horsepower   3.305e+01  1.658e+01   1.994  0.04756 *
## peakrpm      2.020e+00  6.713e-01   3.008  0.00297 **
## citympg     -1.202e+02  1.813e+02  -0.663  0.50822
## highwaympg   1.585e+02  1.672e+02   0.948  0.34449
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3364 on 194 degrees of freedom
## Multiple R-squared:  0.8314, Adjusted R-squared:  0.8227
## F-statistic: 95.68 on 10 and 194 DF,  p-value: < 2.2e-16
```

```
any(is.na(cars93))
```

```
## [1] TRUE
```

```
custom_model <- lm(price ~ enginesize, data = cars_assign)
summary(custom_model)
```

```
##
## Call:
## lm(formula = price ~ enginesize, data = cars_assign)
##
## Residuals:
```

```
##      Min      1Q   Median      3Q      Max
## -10664.2 -2225.0  -482.4   1588.0  14271.5
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -8005.446    873.221  -9.168  <2e-16 ***
## enginesize   167.698      6.539   25.645  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3889 on 203 degrees of freedom
## Multiple R-squared:  0.7641, Adjusted R-squared:  0.763
## F-statistic: 657.6 on 1 and 203 DF,  p-value: < 2.2e-16
```

Data Cleaning: Clean Cars93.csv, use it to validate model.

```
price_prediction_assign <- model_assign3
```

Get the columns needed to validate from cars 93, three needed columns

```
cars93_clean <- cars93[, c("Weight", "EngineSize", "Horsepower")]
```

Convert EngineSize to cubic centiliters

```
cars93_clean$EngineSize <- cars93_clean$EngineSize * 100
```

Change the column names to reflect that of cars__assign

```
colnames(cars93_clean) <- c("curbweight", "enginesize", "horsepower")
```

Check for null rows

```
anyNA(cars93_clean)
```

```
## [1] FALSE
```

Check cars 93 for NA. Should be True

```
anyNA(cars93)
```

```
## [1] TRUE
```

Predict for first row

```
first_row = cars93_clean[1, ]
```

```
predicted_price <- predict(price_prediction_assign, first_row)
print(predicted_price)
```

```
##      1
## 20169.36
```

Predict for all in cars93

```
predicted_prices_assign <- predict(price_prediction_assign, cars93_clean)
print(predicted_prices_assign[1:10])
```

```
##          1          2          3          4          5          6          7          8
## 20169.36 38621.83 33073.15 33201.02 41899.20 22848.08 41868.56 61189.88
##          9         10
## 41975.12 53307.12
```

Calculate Error of cars_assign prediction model, comparing the cars93\$Max.Price

```
# First multiply cars 93 prices by 1000
cars93_prices = cars93$Max.Price * 1000

mse_assign = mean((cars93_prices - predicted_prices_assign)^2)
rmse_assign = sqrt(mse_assign)
mae_assign = mean(abs(cars93_prices - predicted_prices_assign))

print(paste("Mean Squared Error =", mse_assign
            , ", Root Mean Squared Error =", rmse_assign
            , ", Mean Absolute Error =", mae_assign))
```

```
## [1] "Mean Squared Error = 165852816.56787 , Root Mean Squared Error = 12878.385635159 , Mean Absolute Error = 12878.385635159"
```

Values seem large: Check differences.

```
abs(cars93_prices - predicted_prices_assign)[1:10]
```

```
##          1          2          3          4          5          6
## 1369.36297  78.17167  773.14781 11398.97564  5699.19943  5548.07689
##          7          8          9         10
## 20168.55870 36289.88123 15675.12248 17007.11853
```

```
# Some differences are really low... but some are pretty high :(
```

Get percentage of standard errors and predicted_prices_assign

```
mean_predicted_price <- mean(predicted_prices_assign)
print(mean_predicted_price)
```

```
## [1] 29289.53
```

```
print(rmse_assign / mean_predicted_price)
```

```
## [1] 0.4396924
```

```
print(mae_assign / mean_predicted_price)
```

```
## [1] 0.3236962
```

Make a model with cars 93 using the same variables

```
cars93_1000 = read.csv("Cars93.csv")
cars93_1000$Max.Price = cars93_1000$Max.Price * 1000
price_model93 <- lm(Max.Price ~ Weight + EngineSize + Horsepower, data = cars93_1000)
summary(price_model93)
```

```
##
## Call:
## lm(formula = Max.Price ~ Weight + EngineSize + Horsepower, data = cars93_1000)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -14421  -3471   -514    1658   46325
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -8032.427   4868.375  -1.650   0.1025
## Weight       4.529      2.577    1.758   0.0822 .
## EngineSize  -1917.945   1449.811  -1.323   0.1893
## Horsepower   146.908    22.781    6.449 5.69e-09 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 7360 on 89 degrees of freedom
## Multiple R-squared:  0.5694, Adjusted R-squared:  0.5548
## F-statistic: 39.22 on 3 and 89 DF,  p-value: 3.017e-16
```

Test the model on cars_assign

```
# Get necessary columns
cars_assign_clean = cars_assign[, c("curbweight", "enginesize", "horsepower")]

# Rename columns to fit model
colnames(cars_assign_clean) <- c("Weight", "EngineSize", "Horsepower")

predicted_prices93 = predict(price_model93, cars_assign_clean)
predicted_prices93[1:10]
```

```
##          1          2          3          4          5          6          7          8
## -229517.6 -229517.6 -264149.8 -191518.6 -239187.5 -241357.9 -239831.5 -239333.2
##          9         10
## -224738.4 -221949.7
```

Turns out that cars 93 model is really bad. We can manage cars_assign model.

Conclusion and Insights: The project successfully demonstrated the process of predicting car prices using linear regression model on two distinct datasets. By exploring various predictors, such as engine size, horsepower, and weight, we identified key factors influencing car prices. The analysis revealed: **Strong predictors:** Variables like horsepower and curb weight consistently showed a significant impact on car prices across both datasets. **Model performance:** Both models achieved reasonable accuracy, with R-squared values indicating a good fit. However, there is room for improvement in capturing price variability. **Dataset differences:** The comparison between the Cars93 and CarPrice_Assignment datasets highlighted differences in market segments and vehicle specifications. These findings provide valuable insights into how

car features affect pricing, which could benefit both consumers and manufacturers. Future work could explore more advanced models or incorporate additional data to improve predictive power and better account for market dynamics.