

Imports

```
source("preprocess.R")
```

Load Data and Format

```
# Load data
data <- read.csv("car details v4.csv")
nrow(data)

## [1] 2059

df <- data
df <- remove_null(df)
df <- df[!duplicated(df), ]
nrow(df)

## [1] 1874

# Drop unwanted columns
df <- remove_cols(df, c("Model"))

# Remove the " cc" suffix from Engine
df$Engine <- as.numeric(gsub("[^0-9]", "", df$Engine)) # Convert to numbers, ex. "123 cc" -> 123

# Split max torque and the RPM at which max torque occurs
df <- tidyr::separate(df, Max.Torque, into = c("Max.Torque.Value", "Max.Torque.RPM"), sep = "@")
df$Max.Torque.Value <- as.numeric(gsub("[^0-9]", "", df$Max.Torque.Value))
df$Max.Torque.RPM <- as.numeric(gsub("[^0-9]", "", df$Max.Torque.RPM))

# Include RPM and torque interaction
df$Torque.RPM <- df$Max.Torque.Value * df$Max.Torque.RPM

# Split the max horsepower and the RPM at which max horsepower occurs
df <- tidyr::separate(df, Max.Power, into = c("Max.Power.Value", "Max.Power.RPM"), sep = "@")
df$Max.Power.Value <- as.numeric(gsub("[^0-9]", "", df$Max.Power.Value))
df$Max.Power.RPM <- as.numeric(gsub("[^0-9]", "", df$Max.Power.RPM))

colnames(df)

## [1] "Make" "Price" "Year"
## [4] "Kilometer" "Fuel.Type" "Transmission"
## [7] "Location" "Color" "Owner"
## [10] "Seller.Type" "Engine" "Max.Power.Value"
## [13] "Max.Power.RPM" "Max.Torque.Value" "Max.Torque.RPM"
## [16] "Drivetrain" "Length" "Width"
## [19] "Height" "Seating.Capacity" "Fuel.Tank.Capacity"
## [22] "Torque.RPM"

# remove nulls again in case the power and torque columns are malformed
df <- remove_null(df)

# Include power and RPM interaction
df$Power.RPM <- df$Max.Power.Value * df$Max.Power.RPM

# Numerical and categorical regressors
```

```

# Note that Seating.Capacity was omitted to exempt it from 1.5 IQR
numerical <- c(
  "Price", "Year", "Kilometer", "Engine",
  "Max.Power.Value", "Max.Power.RPM", "Max.Torque.Value", "Max.Torque.RPM",
  "Length", "Width", "Height", "Fuel.Tank.Capacity",
  "Torque.RPM", "Power.RPM"
)
categorical <- c(
  "Make", "Fuel.Type", "Transmission", "Location", "Color",
  "Owner", "Seller.Type", "Drivetrain"
)

# Outliers for numerical regressors
df <- remove_outliers(df, numerical)
nrow(df)

## [1] 1434

# Convert categorical data to factors
res <- convert_categorical(df, categorical)
design <- as.data.frame(res$dummy)
ncol(design)

## [1] 138

# Remove columns with only one observation and affected rows
singles <- c()
bad_idx <- c()
for (col in colnames(design)) {
  if (sum(design[, col] != 0) <= 1) {
    singles <- c(singles, col)
    bad_idx <- c(bad_idx, which(design[, col] != 0))
  }
}
singles

## [1] "MakeFiat" "MakeLexus" "Fuel.TypePetrol + CNG"
## [4] "LocationDak. Kannada" "LocationFaizabad" "LocationGorakhpur"
## [7] "LocationMirzapur" "LocationPurnea" "LocationRohtak"
## [10] "LocationRudrapur" "LocationSamastipur" "LocationValsad"

bad_idx

## [1] 654 996 1153 1063 770 563 765 719 242 505 261 159

design <- remove_cols(design, singles)
design <- design[-bad_idx, ]

nrow(design)

## [1] 1422

ncol(design)

## [1] 126

```

Multicollinearity

```
x <- design
model <- lm(log(Price) ~., data = x)
removed <- c()

finished <- F
while(!finished) {
  temp <- car::vif(model)
  worst <- names(which.max(temp))
  if (temp[worst] > 5) {
    x <- remove_cols(x, c(gsub("^`|`$", "", worst)))
    model <- lm(log(Price) ~., data = x)

    removed <- c(removed, worst)
  } else {
    finished <- T
  }
}

removed
```

```
## [1] "Max.Power.Value"      "ColorWhite"           "Max.Torque.RPM"
## [4] "Max.Power.RPM"        "Fuel.TypeDiesel"      "Max.Torque.Value"
## [7] "LocationMumbai"      "Seller.TypeIndividual" "Engine"
## [10] "`MakeMaruti Suzuki`" "Length"               "Fuel.Tank.Capacity"
```

Leverage Points

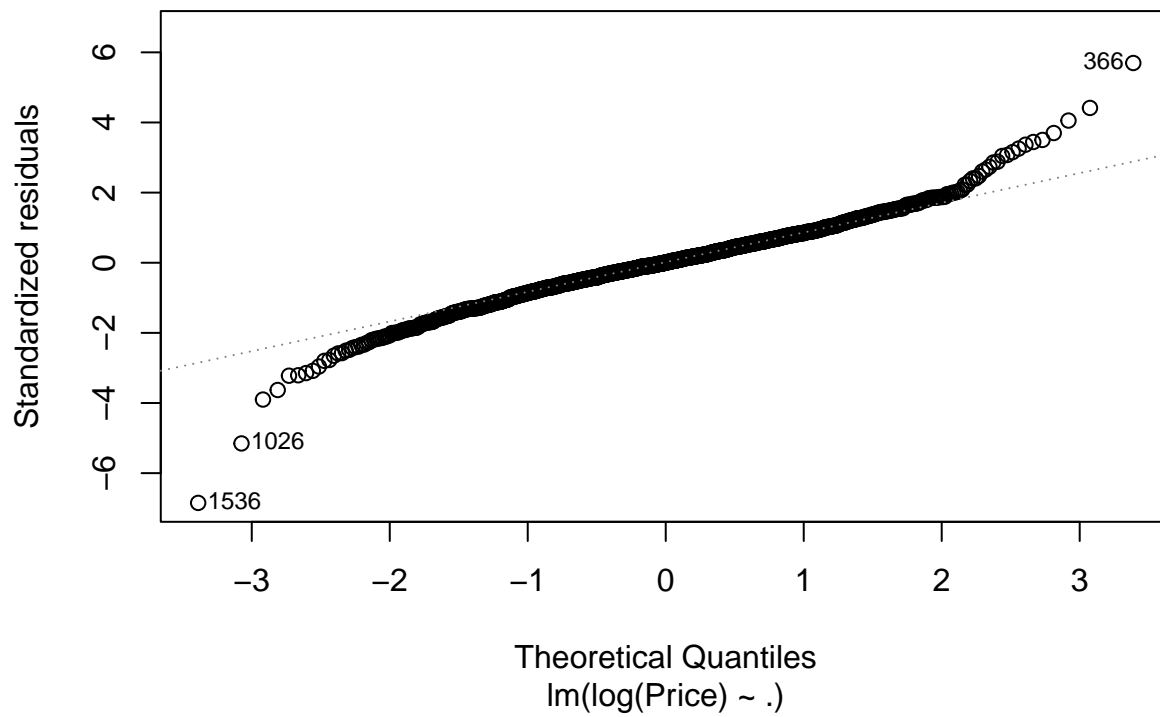
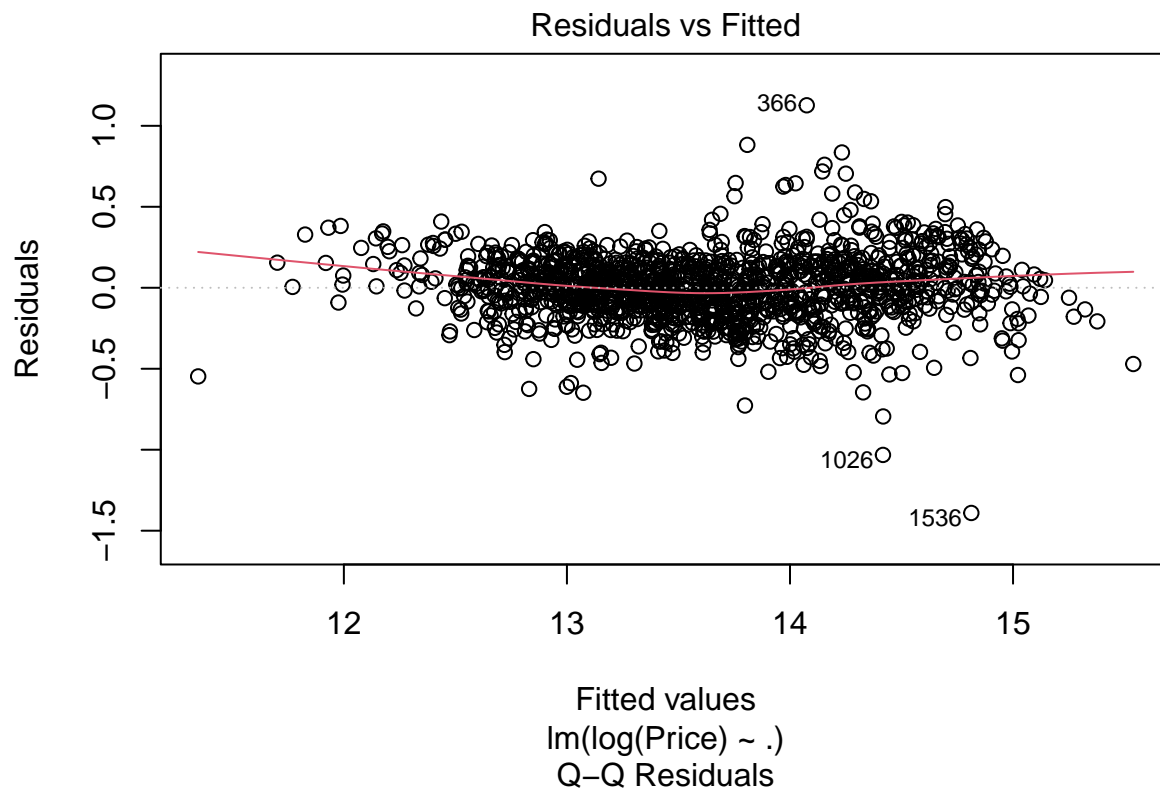
```
which(abs(rstudent(model)) > 4)
```

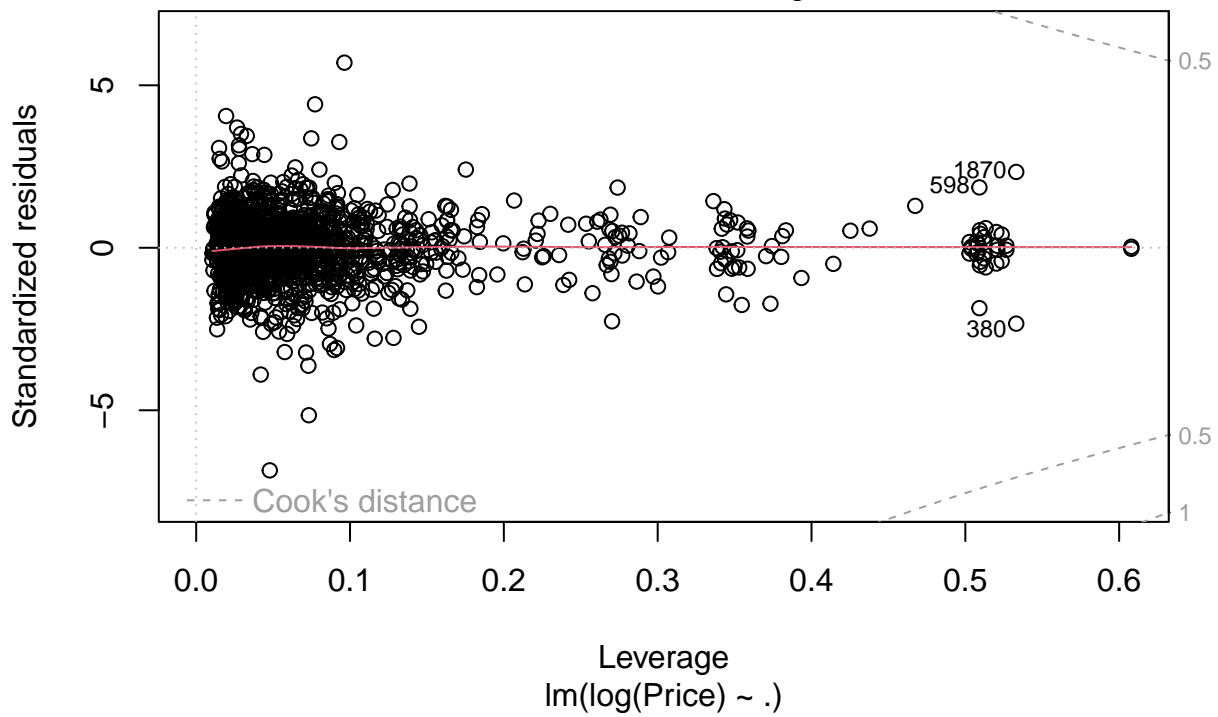
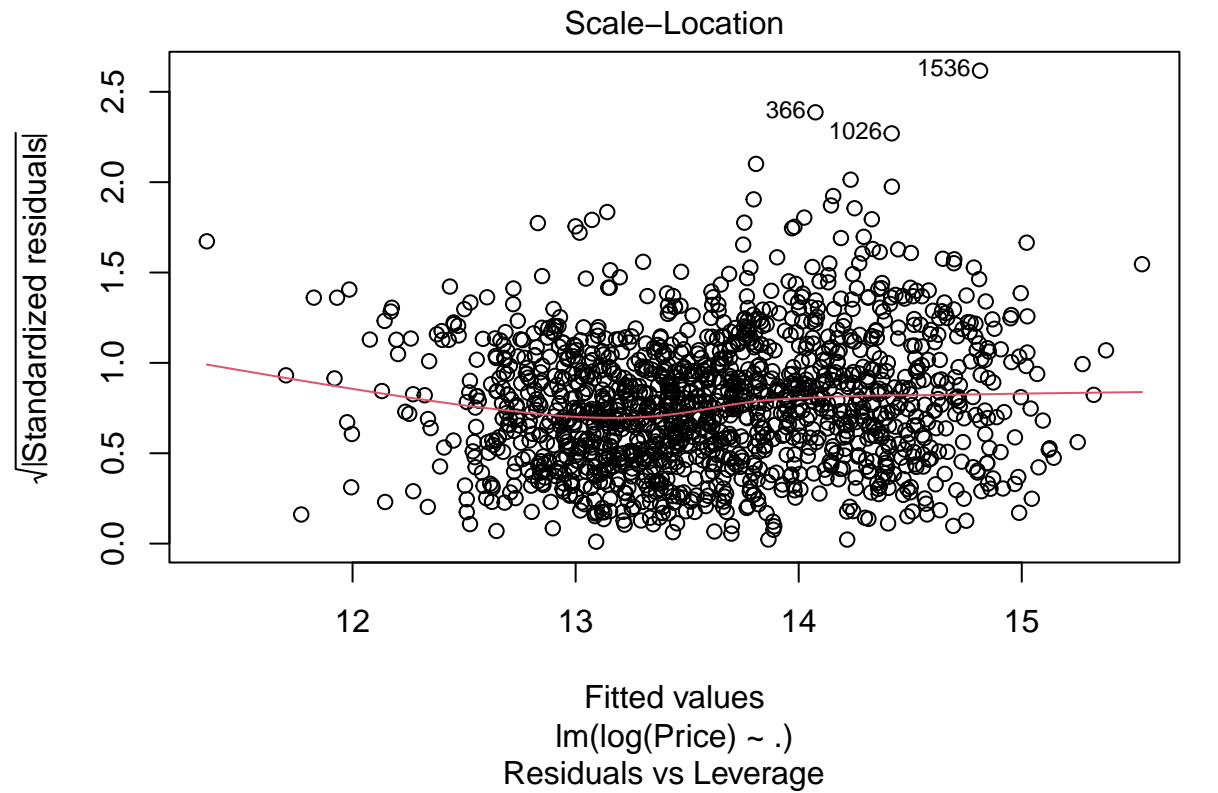
```
## 44 366 610 1026 1536
## 39 265 426 712 1056
```

```
# all of these points were manually inspected, and were not suspected to be clerical errors.
```

Diagnostics

```
plot(model)
```





```
summary(model)
```

```
##
## Call:
## lm(formula = log(Price) ~ ., data = x)
##
## Residuals:
```

```

##      Min      1Q   Median      3Q      Max
## -1.39045 -0.10899  0.00104  0.11573  1.12668
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    -1.918e+02  5.543e+00 -34.603 < 2e-16 ***
## MakeBMW         1.250e-01  4.830e-02   2.588 0.009760 **
## MakeChevrolet   -3.156e-01  1.064e-01  -2.966 0.003070 **
## MakeDatsun      -3.711e-01  7.722e-02  -4.806 1.72e-06 ***
## MakeFord        -1.870e-01  3.726e-02  -5.017 5.96e-07 ***
## MakeHonda       -7.359e-02  2.305e-02  -3.193 0.001442 **
## MakeHyundai     -7.965e-02  1.706e-02  -4.670 3.32e-06 ***
## MakeIsuzu       -1.894e-02  1.543e-01  -0.123 0.902303
## MakeJaguar      -1.504e-02  1.294e-01  -0.116 0.907468
## MakeJeep        -5.365e-02  5.786e-02  -0.927 0.353997
## MakeKia         -5.703e-02  4.777e-02  -1.194 0.232764
## `MakeLand Rover` 4.689e-01  1.525e-01   3.076 0.002144 **
## MakeMahindra    -2.820e-01  3.357e-02  -8.401 < 2e-16 ***
## `MakeMercedes-Benz` 4.437e-01  3.884e-02  11.424 < 2e-16 ***
## MakeMG          -4.836e-02  6.330e-02  -0.764 0.445001
## MakeMINI        8.575e-01  9.425e-02   9.099 < 2e-16 ***
## MakeMitsubishi  -3.443e-02  1.358e-01  -0.253 0.799942
## MakeNissan      -1.973e-01  5.793e-02  -3.406 0.000680 ***
## MakeRenault     -2.086e-01  3.627e-02  -5.752 1.09e-08 ***
## MakeSkoda       1.371e-02  3.925e-02   0.349 0.726908
## MakeTata        -3.098e-01  3.599e-02  -8.606 < 2e-16 ***
## MakeToyota      1.807e-01  2.902e-02   6.225 6.46e-10 ***
## MakeVolkswagen  4.132e-02  3.631e-02   1.138 0.255320
## MakeVolvo       2.774e-01  6.614e-02   4.194 2.93e-05 ***
## Year           1.000e-01  2.769e-03  36.117 < 2e-16 ***
## Kilometer      -2.047e-06  2.929e-07  -6.988 4.44e-12 ***
## Fuel.TypePetrol -4.260e-01  1.870e-02 -22.781 < 2e-16 ***
## TransmissionManual -2.075e-01  1.507e-02 -13.770 < 2e-16 ***
## LocationAhmedabad -8.261e-03  3.286e-02  -0.251 0.801561
## LocationAllahabad -2.148e-01  1.217e-01  -1.765 0.077786 .
## `LocationAmbala Cantt` -2.767e-01  6.958e-02  -3.976 7.40e-05 ***
## LocationAmritsar -2.675e-03  1.077e-01  -0.025 0.980188
## LocationAurangabad 2.100e-01  1.221e-01   1.720 0.085686 .
## LocationBangalore 1.480e-01  2.550e-02   5.804 8.13e-09 ***
## LocationBhopal   -1.442e-02  1.225e-01  -0.118 0.906344
## LocationBhubaneswar -3.519e-02  1.275e-01  -0.276 0.782525
## LocationBulandshahar -3.404e-01  1.521e-01  -2.238 0.025399 *
## LocationChandigarh 3.165e-02  5.737e-02   0.552 0.581233
## LocationChennai  1.786e-01  3.788e-02   4.716 2.67e-06 ***
## LocationCoimbatore 2.046e-01  5.222e-02   3.918 9.40e-05 ***
## LocationDehradun  4.099e-02  5.110e-02   0.802 0.422598
## LocationDelhi    -2.179e-02  1.991e-02  -1.094 0.273962
## LocationDharwad   2.544e-02  1.487e-01   0.171 0.864154
## LocationErnakulam 1.505e-01  1.491e-01   1.009 0.313137
## LocationFaridabad -9.745e-02  4.748e-02  -2.052 0.040337 *
## LocationGhaziabad 6.304e-02  1.533e-01   0.411 0.680959
## LocationGoa       5.127e-01  1.072e-01   4.782 1.93e-06 ***
## LocationGurgaon  -1.652e-02  4.702e-02  -0.351 0.725427
## LocationGuwahati  5.416e-02  9.683e-02   0.559 0.576013

```

## LocationHaldwani	-1.613e-01	1.490e-01	-1.083	0.279191	
## LocationHyderabad	1.040e-01	2.783e-02	3.737	0.000195	***
## LocationIndore	1.803e-01	1.063e-01	1.695	0.090289	.
## LocationJaipur	6.025e-02	4.579e-02	1.316	0.188470	
## LocationJalandhar	4.853e-02	5.196e-02	0.934	0.350452	
## LocationJamshedpur	4.408e-02	9.649e-02	0.457	0.647895	
## LocationKanpur	-9.718e-02	4.008e-02	-2.425	0.015458	*
## LocationKarnal	9.823e-02	7.616e-02	1.290	0.197352	
## LocationKharar	-6.913e-02	1.485e-01	-0.465	0.641698	
## LocationKheda	-8.693e-02	1.490e-01	-0.583	0.559765	
## LocationKolkata	-1.576e-01	3.201e-02	-4.925	9.52e-07	***
## LocationKollam	1.506e-01	1.493e-01	1.008	0.313551	
## LocationKota	1.424e-01	1.635e-01	0.871	0.383969	
## LocationLucknow	1.570e-03	3.370e-02	0.047	0.962851	
## LocationLudhiana	-3.600e-02	4.023e-02	-0.895	0.371021	
## LocationMangalore	-5.101e-02	1.062e-01	-0.480	0.631089	
## LocationMeerut	1.963e-02	7.578e-02	0.259	0.795665	
## LocationMohali	5.729e-02	4.724e-02	1.213	0.225407	
## LocationMuzaffarpur	-7.812e-02	1.491e-01	-0.524	0.600437	
## LocationMysore	3.033e-01	8.210e-02	3.694	0.000230	***
## LocationNagpur	1.528e-01	1.222e-01	1.250	0.211614	
## LocationNashik	5.282e-02	8.081e-02	0.654	0.513441	
## `LocationNavi Mumbai`	1.797e-02	6.003e-02	0.299	0.764759	
## LocationNoida	4.097e-02	6.010e-02	0.682	0.495557	
## LocationPanchkula	-7.015e-02	1.214e-01	-0.578	0.563431	
## LocationPatna	1.440e-02	4.005e-02	0.360	0.719213	
## LocationPune	7.107e-02	2.590e-02	2.744	0.006155	**
## LocationRaipur	6.231e-02	5.449e-02	1.143	0.253071	
## LocationRanchi	-8.400e-02	5.459e-02	-1.539	0.124100	
## `LocationRanga Reddy`	-1.628e-01	1.499e-01	-1.086	0.277577	
## LocationRoorkee	1.123e-01	1.237e-01	0.907	0.364479	
## LocationSalem	3.265e-01	1.502e-01	2.174	0.029904	*
## LocationSurat	3.894e-03	1.069e-01	0.036	0.970938	
## LocationThane	1.182e-02	5.749e-02	0.206	0.837177	
## LocationUdupi	1.007e-01	7.543e-02	1.335	0.182226	
## LocationVadodara	-1.895e-01	1.218e-01	-1.556	0.119912	
## LocationVaranasi	-1.597e-01	5.471e-02	-2.918	0.003581	**
## LocationWarangal	-2.123e-01	1.498e-01	-1.417	0.156649	
## LocationYamunanagar	-2.660e-02	1.236e-01	-0.215	0.829635	
## LocationZirakpur	-7.028e-02	5.744e-02	-1.224	0.221357	
## ColorBlack	-1.737e-02	2.454e-02	-0.708	0.479223	
## ColorBlue	-2.190e-02	2.306e-02	-0.950	0.342485	
## ColorBronze	-4.483e-02	4.662e-02	-0.962	0.336465	
## ColorBrown	2.604e-04	2.863e-02	0.009	0.992745	
## ColorGold	-1.939e-02	4.761e-02	-0.407	0.683929	
## ColorGreen	-5.648e-02	6.952e-02	-0.812	0.416687	
## ColorGrey	-8.916e-03	1.920e-02	-0.464	0.642504	
## ColorMaroon	-7.699e-02	4.646e-02	-1.657	0.097728	.
## ColorOrange	2.686e-02	5.866e-02	0.458	0.647088	
## ColorOthers	-3.179e-02	6.603e-02	-0.481	0.630335	
## ColorPurple	-5.198e-02	1.487e-01	-0.349	0.726794	
## ColorRed	-1.298e-02	2.247e-02	-0.578	0.563413	
## ColorSilver	-4.214e-02	1.773e-02	-2.377	0.017613	*
## ColorYellow	-4.104e-02	1.271e-01	-0.323	0.746793	

```
## OwnerSecond          -2.712e-02  1.656e-02  -1.638  0.101685
## OwnerThird           -4.831e-02  5.886e-02  -0.821  0.411950
## `OwnerUnRegistered Car` 7.793e-02  1.117e-01   0.698  0.485480
## Seller.TypeCorporate   -9.918e-03  3.810e-02  -0.260  0.794666
## DrivetrainFWD         -2.307e-01  2.926e-02  -7.885  6.58e-15 ***
## DrivetrainRWD         -1.525e-01  3.416e-02  -4.465  8.69e-06 ***
## Width                 2.148e-03  1.120e-04  19.180  < 2e-16 ***
## Height               -1.709e-04  9.080e-05  -1.882  0.060067 .
## Seating.Capacity       5.839e-02  1.268e-02   4.606  4.50e-06 ***
## Torque.RPM            -8.338e-08  5.825e-08  -1.431  0.152527
## Power.RPM             1.297e-06  6.875e-08  18.860  < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.2081 on 1308 degrees of freedom
## Multiple R-squared:  0.9172, Adjusted R-squared:  0.9101
## F-statistic: 128.3 on 113 and 1308 DF,  p-value: < 2.2e-16
```

Model Reduction

```
reduced_model <- step(model, direction = "backward", data = x, trace = 0)
old <- names(model$coefficients)
new <- names(reduced_model$coefficients)

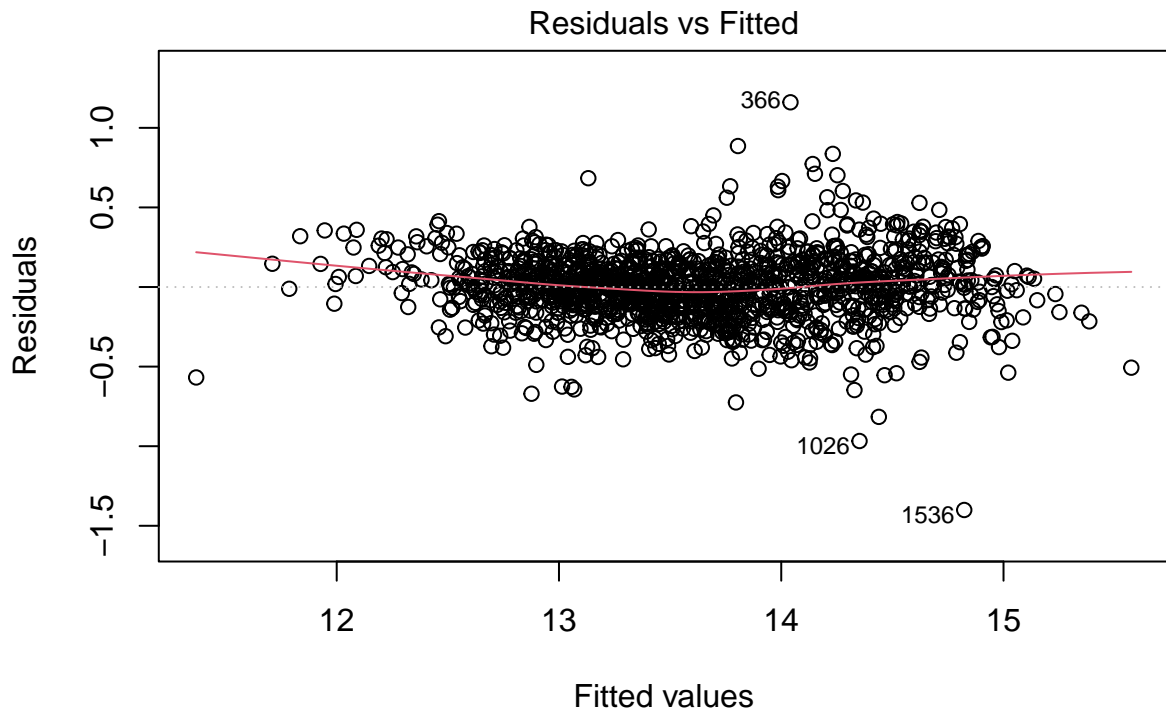
old[!(old %in% new)]
```

```
## [1] "MakeIsuzu"          "MakeJaguar"
## [3] "MakeJeep"           "MakeKia"
## [5] "MakeMG"             "MakeMitsubishi"
## [7] "MakeSkoda"          "MakeVolkswagen"
## [9] "LocationAhmedabad"  "LocationAmritsar"
## [11] "LocationBhopal"     "LocationBhubaneswar"
## [13] "LocationChandigarh" "LocationDehradun"
## [15] "LocationDharwad"    "LocationErnakulam"
## [17] "LocationGhaziabad"  "LocationGurgaon"
## [19] "LocationGuwahati"   "LocationHaldwani"
## [21] "LocationJaipur"     "LocationJalandhar"
## [23] "LocationJamshedpur" "LocationKarnal"
## [25] "LocationKharar"     "LocationKheda"
## [27] "LocationKollam"     "LocationKota"
## [29] "LocationLucknow"    "LocationMangalore"
## [31] "LocationMeerut"     "LocationMohali"
## [33] "LocationMuzaffarpur" "LocationNagpur"
## [35] "LocationNashik"     "`LocationNavi Mumbai`"
## [37] "LocationNoida"      "LocationPanchkula"
## [39] "LocationPatna"      "LocationRaipur"
## [41] "`LocationRanga Reddy`" "LocationRoorkee"
## [43] "LocationSurat"      "LocationThane"
## [45] "LocationUdupi"      "LocationYamunanagar"
## [47] "ColorBlack"         "ColorBlue"
## [49] "ColorBronze"        "ColorBrown"
## [51] "ColorGold"          "ColorGreen"
## [53] "ColorGrey"          "ColorOrange"
## [55] "ColorOthers"        "ColorPurple"
```

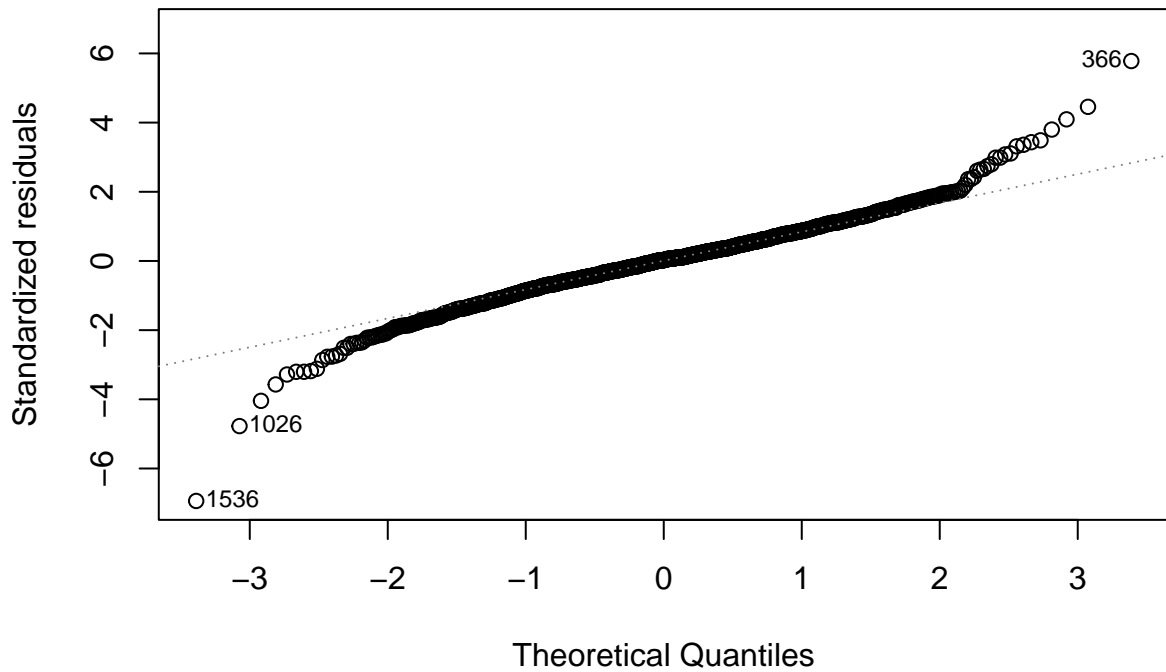


```
## [57] "ColorRed"          "ColorYellow"
## [59] "OwnerSecond"       "OwnerThird"
## [61] "`OwnerUnRegistered Car`" "Seller.TypeCorporate"

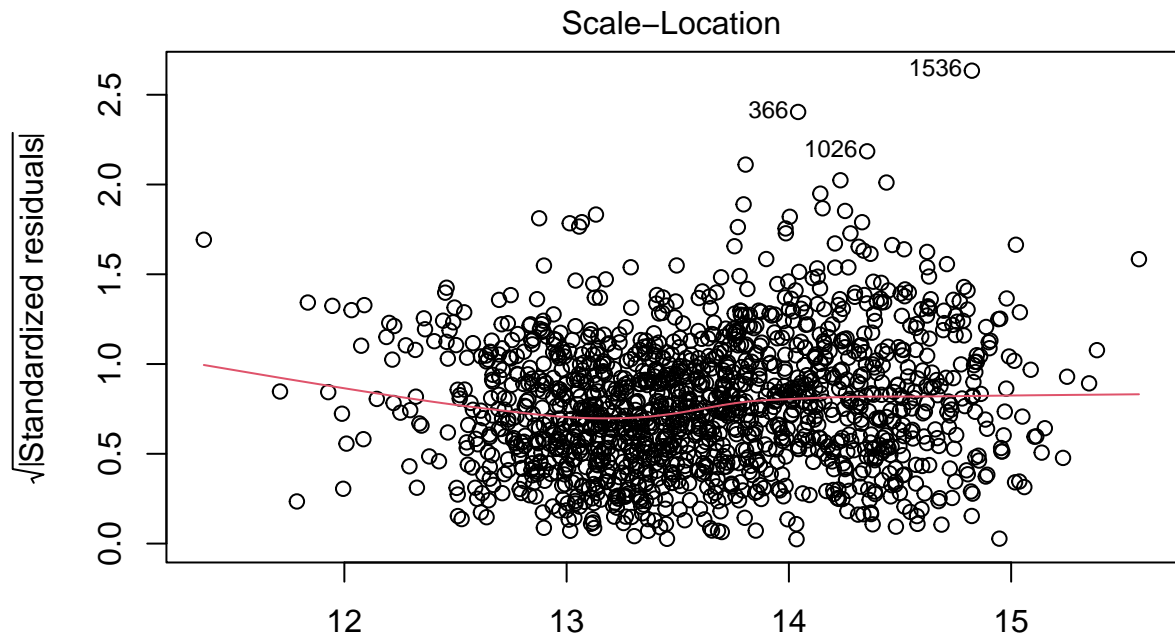
plot(reduced_model)
```



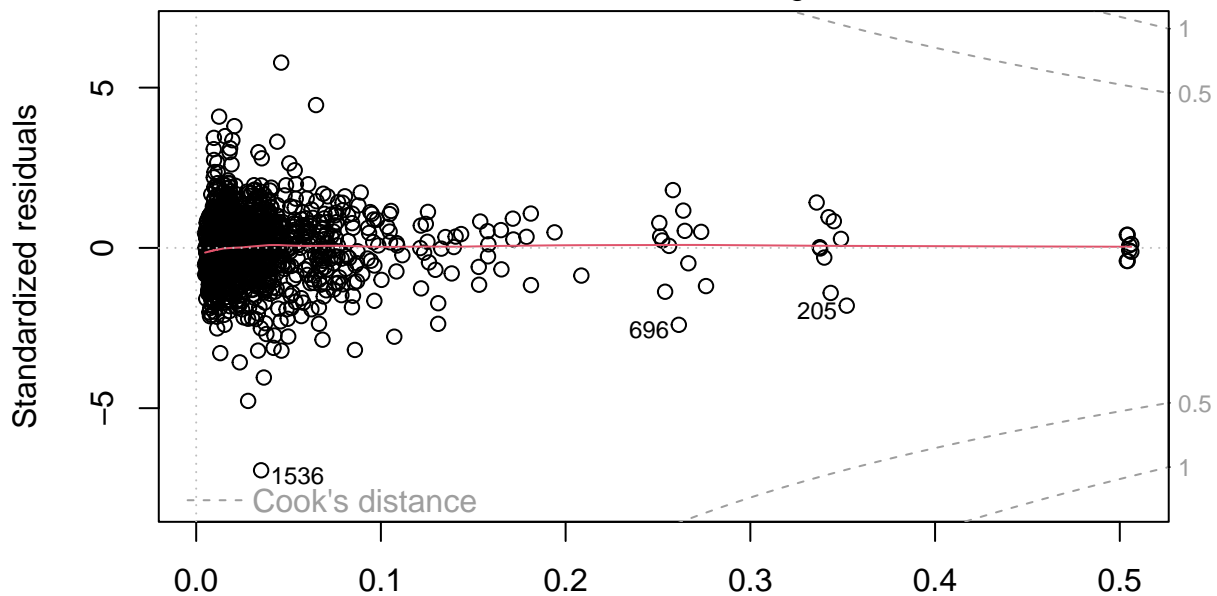
$\text{lm}(\log(\text{Price}) \sim \text{MakeBMW} + \text{MakeChevrolet} + \text{MakeDatsun} + \text{MakeFord} + \text{MakeHonda})$
Q-Q Residuals



$\text{lm}(\log(\text{Price}) \sim \text{MakeBMW} + \text{MakeChevrolet} + \text{MakeDatsun} + \text{MakeFord} + \text{MakeHonda})$



Fitted values
 $\text{lm}(\log(\text{Price}) \sim \text{MakeBMW} + \text{MakeChevrolet} + \text{MakeDatsun} + \text{MakeFord} + \text{MakeHonda} + \text{MakeHyundai} + \text{MakeLand Rover} + \text{MakeMahindra} + \text{MakeMercedes-Benz} + \text{MakeMINI} + \text{MakeNissan} + \text{MakeRenault} + \text{MakeSubaru} + \text{MakeSuzuki} + \text{MakeToyota} + \text{MakeVauxhall})$
 Residuals vs Leverage



Leverage
 $\text{lm}(\log(\text{Price}) \sim \text{MakeBMW} + \text{MakeChevrolet} + \text{MakeDatsun} + \text{MakeFord} + \text{MakeHonda} + \text{MakeHyundai} + \text{MakeLand Rover} + \text{MakeMahindra} + \text{MakeMercedes-Benz} + \text{MakeMINI} + \text{MakeNissan} + \text{MakeRenault} + \text{MakeSubaru} + \text{MakeSuzuki} + \text{MakeToyota} + \text{MakeVauxhall})$

```
summary(reduced_model)
```

```
##
## Call:
## lm(formula = log(Price) ~ MakeBMW + MakeChevrolet + MakeDatsun +
##     MakeFord + MakeHonda + MakeHyundai + `MakeLand Rover` + MakeMahindra +
##     `MakeMercedes-Benz` + MakeMINI + MakeNissan + MakeRenault +
```

```

##      MakeTata + MakeToyota + MakeVolvo + Year + Kilometer + Fuel.TypePetrol +
##      TransmissionManual + LocationAllahabad + `LocationAmbala Cantt` +
##      LocationAurangabad + LocationBangalore + LocationBulandshahar +
##      LocationChennai + LocationCoimbatore + LocationDelhi + LocationFaridabad +
##      LocationGoa + LocationHyderabad + LocationIndore + LocationKanpur +
##      LocationKolkata + LocationLudhiana + LocationMysore + LocationPune +
##      LocationRanchi + LocationSalem + LocationVadodara + LocationVaranasi +
##      LocationWarangal + LocationZirakpur + ColorMaroon + ColorSilver +
##      DrivetrainFWD + DrivetrainRWD + Width + Height + Seating.Capacity +
##      Torque.RPM + Power.RPM, data = x)
##
## Residuals:
##      Min        1Q    Median        3Q        Max
## -1.40073 -0.11213  0.00538  0.11549  1.16036
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    -1.933e+02  5.133e+00 -37.656 < 2e-16 ***
## MakeBMW         1.132e-01  4.562e-02   2.482 0.013187 *
## MakeChevrolet   -3.115e-01  1.041e-01  -2.992 0.002819 **
## MakeDatsun      -3.838e-01  7.499e-02  -5.118 3.52e-07 ***
## MakeFord        -1.771e-01  3.556e-02  -4.979 7.21e-07 ***
## MakeHonda       -7.556e-02  2.126e-02  -3.554 0.000392 ***
## MakeHyundai     -7.756e-02  1.537e-02  -5.046 5.13e-07 ***
## `MakeLand Rover` 4.537e-01  1.487e-01   3.051 0.002324 **
## MakeMahindra    -2.798e-01  3.095e-02  -9.040 < 2e-16 ***
## `MakeMercedes-Benz` 4.282e-01  3.620e-02  11.829 < 2e-16 ***
## MakeMINI        8.278e-01  8.661e-02   9.557 < 2e-16 ***
## MakeNissan      -2.085e-01  5.592e-02  -3.728 0.000200 ***
## MakeRenault     -2.088e-01  3.420e-02  -6.106 1.32e-09 ***
## MakeTata        -3.115e-01  3.385e-02  -9.200 < 2e-16 ***
## MakeToyota      1.782e-01  2.669e-02   6.674 3.60e-11 ***
## MakeVolvo       2.771e-01  6.414e-02   4.320 1.67e-05 ***
## Year            1.008e-01  2.566e-03  39.268 < 2e-16 ***
## Kilometer       -1.954e-06  2.761e-07  -7.078 2.33e-12 ***
## Fuel.TypePetrol -4.239e-01  1.781e-02 -23.805 < 2e-16 ***
## TransmissionManual -2.025e-01  1.436e-02 -14.103 < 2e-16 ***
## LocationAllahabad -2.166e-01  1.193e-01  -1.816 0.069649 .
## `LocationAmbala Cantt` -3.052e-01  6.672e-02  -4.574 5.22e-06 ***
## LocationAurangabad 1.905e-01  1.198e-01   1.591 0.111891
## LocationBangalore 1.350e-01  2.258e-02   5.980 2.85e-09 ***
## LocationBulandshahar -3.981e-01  1.467e-01  -2.714 0.006736 **
## LocationChennai  1.670e-01  3.574e-02   4.671 3.29e-06 ***
## LocationCoimbatore 1.868e-01  4.948e-02   3.775 0.000167 ***
## LocationDelhi    -3.239e-02  1.689e-02  -1.918 0.055365 .
## LocationFaridabad -1.039e-01  4.518e-02  -2.299 0.021643 *
## LocationGoa      5.029e-01  1.045e-01   4.811 1.67e-06 ***
## LocationHyderabad 9.109e-02  2.473e-02   3.684 0.000239 ***
## LocationIndore   1.532e-01  1.041e-01   1.472 0.141173
## LocationKanpur   -1.146e-01  3.724e-02  -3.078 0.002124 **
## LocationKolkata  -1.773e-01  2.962e-02  -5.987 2.73e-09 ***
## LocationLudhiana -5.805e-02  3.735e-02  -1.555 0.120295
## LocationMysore   2.719e-01  7.908e-02   3.438 0.000603 ***
## LocationPune     5.717e-02  2.346e-02   2.437 0.014940 *

```

```

## LocationRanchi      -9.385e-02  5.259e-02  -1.785  0.074541 .
## LocationSalem       2.944e-01  1.460e-01   2.016  0.043986 *
## LocationVadodara    -2.021e-01  1.197e-01  -1.689  0.091495 .
## LocationVaranasi    -1.820e-01  5.252e-02  -3.466  0.000544 ***
## LocationWarangal    -2.346e-01  1.465e-01  -1.601  0.109503
## LocationZirakpur    -8.295e-02  5.469e-02  -1.517  0.129561
## ColorMaroon         -7.123e-02  4.492e-02  -1.586  0.113003
## ColorSilver         -3.811e-02  1.567e-02  -2.433  0.015121 *
## DrivetrainFWD       -2.362e-01  2.750e-02  -8.589  < 2e-16 ***
## DrivetrainRWD       -1.486e-01  3.229e-02  -4.601  4.59e-06 ***
## Width               2.154e-03  1.073e-04  20.067  < 2e-16 ***
## Height              -2.229e-04  8.132e-05  -2.741  0.006209 **
## Seating.Capacity     6.032e-02  1.212e-02   4.976  7.30e-07 ***
## Torque.RPM          -7.670e-08  5.512e-08  -1.392  0.164255
## Power.RPM           1.275e-06  6.428e-08  19.841  < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.2055 on 1370 degrees of freedom
## Multiple R-squared:  0.9154, Adjusted R-squared:  0.9123
## F-statistic: 290.7 on 51 and 1370 DF,  p-value: < 2.2e-16

npar <- length(model$coefficients) + 1; n <- length(model$residuals)
extractAIC(model)[2] + 2*npar*(npar + 1)/(n - npar - 1)

## [1] -4335.158

npar <- length(reduced_model$coefficients) + 1; n <- length(reduced_model$residuals)
extractAIC(reduced_model)[2] + 2*npar*(npar + 1)/(n - npar - 1)

## [1] -4444.41

```

Interactions