

```

library(car)

## Loading required package: carData

library(reshape2)
library(ggplot2)
library(MASS)
library(interactions)

source("clean_data.R")

df <- remove_cols(df, c("Color", "Model"))

# Remove columns with only one observation and affected rows
res <- convert_categorical(df, categorical)
design <- as.data.frame(res$dummy)

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] "LocationPurnea" "LocationRohtak" "LocationRudrapur"
## [10] "LocationSamastipur" "LocationValsad"

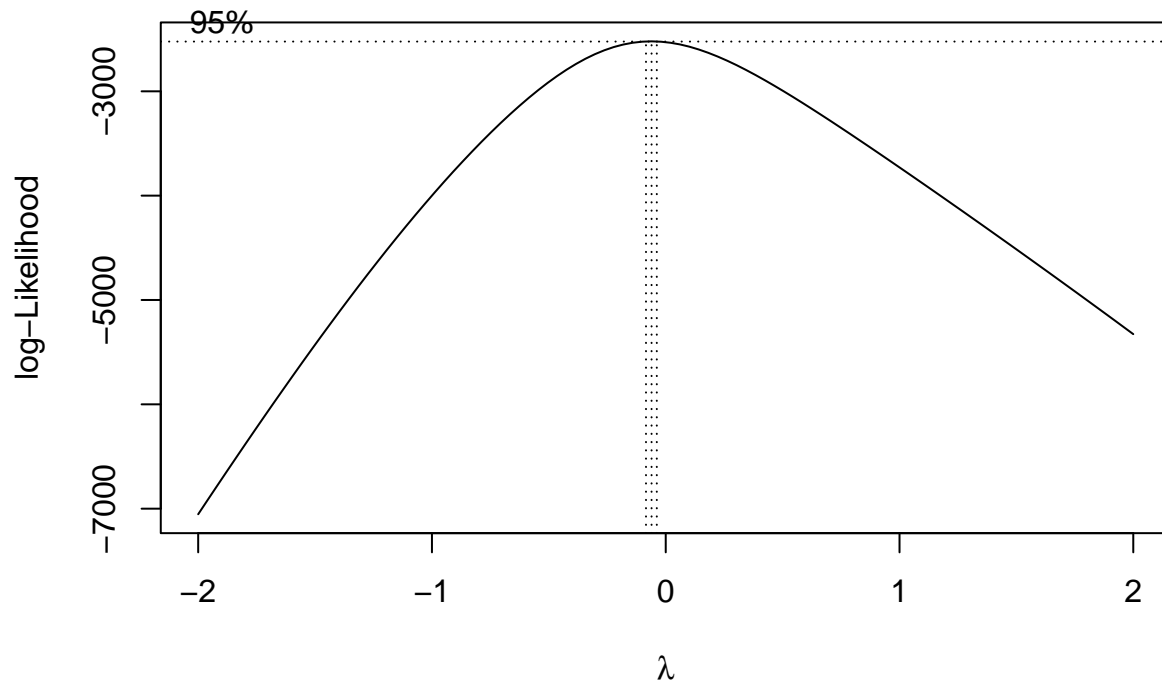
bad_idx

## [1] 662 1009 1169 1077 779 569 728 245 510 264 162
df <- df[-bad_idx, ]

```

Box Cox

```
bc <- boxcox(Price ~., data = df)
```



```
bc$x[which.max(bc$y)]
```

```
## [1] -0.06060606
```

For the sake of interpretability, we will use a log transformation, since $\lambda \approx 0$.

Iteratively Remove Multicollinear Regressors

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

finished <- F
while(!finished) {
  temp <- car::vif(model)[, "GVIF^(1/(2*Df))"]
  worst <- names(which.max(temp))
  if (length(temp) > 0 && temp[worst] > sqrt(5)) {
    x <- remove_cols(x, c(worst))
    model <- lm(log(Price) ~., data = x)

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

removed

## [1] "Max.Torque.Value" "Max.Power.RPM" "Max.Power.Value"
## [4] "Engine" "Fuel.Tank.Capacity" "Height"
which(abs(rstudent(model)) > 4)

## 44 366 492 1536 1681
```

```
## 39 268 345 1071 1171
```

After inspecting these data points, we do not find a good reason to remove them (i.e., they are not clerical errors).

Add Interactions

We consider only numerical interactions due to data sparsity; including categorical values produces NA's for most interactions. We do not expect kilometers driven to interact with torque or width, so those interactions are not considered.

```
names(model$coefficients)[names(model$coefficients) %in% numerical]
```

```
## [1] "Year"          "Kilometer"      "Max.Torque.RPM" "Length"
## [5] "Width"
```

```
model <- lm(
  log(Price) ~ . + Year:Kilometer + Year:Max.Torque.RPM
              + Year:Length + Year:Width
              + Max.Torque.RPM:Length + Max.Torque.RPM:Width
              + Length:Width,
  data = x
)

anova(model)
```

```
## Analysis of Variance Table
```

```
##
```

```
## Response: log(Price)
```

```
##              Df Sum Sq Mean Sq  F value    Pr(>F)
## Make           25 354.44   14.178   482.1594 < 2.2e-16 ***
## Year            1 178.52  178.525  6071.3552 < 2.2e-16 ***
## Kilometer       1   0.98   0.982    33.3800 9.430e-09 ***
## Fuel.Type       2  19.23   9.615    326.9979 < 2.2e-16 ***
## Transmission    1  20.31  20.313    690.7982 < 2.2e-16 ***
## Location        63  16.76   0.266     9.0458 < 2.2e-16 ***
## Owner           3   0.46   0.153     5.2143 0.001397 **
## Seller.Type      2   0.33   0.165     5.6238 0.003697 **
## Max.Torque.RPM   1   1.67   1.669    56.7628 9.050e-14 ***
## Drivetrain       2   5.81   2.905    98.7875 < 2.2e-16 ***
## Length          1  51.72  51.724  1759.0668 < 2.2e-16 ***
## Width           1   5.06   5.063    172.1998 < 2.2e-16 ***
## Seating.Capacity 1   0.02   0.019     0.6507 0.419991
## Year:Kilometer   1   1.10   1.100    37.4187 1.251e-09 ***
## Year:Max.Torque.RPM 1   0.99   0.993    33.7868 7.690e-09 ***
## Year:Length      1   2.04   2.038    69.2926 < 2.2e-16 ***
## Year:Width       1   0.08   0.076     2.5959 0.107377
## Max.Torque.RPM:Length 1   0.28   0.276     9.3872 0.002229 **
## Max.Torque.RPM:Width 1   0.01   0.012     0.3960 0.529287
## Length:Width     1   0.01   0.014     0.4748 0.490893
## Residuals      1329  39.08   0.029
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Interactions with width don't seem to be significant; remove them.

```
model <- lm(
  log(Price) ~ . + Year:Kilometer + Year:Max.Torque.RPM
              + Year:Length + Max.Torque.RPM:Length,
  data = x
)
```

```
anova(model)
```

```
## Analysis of Variance Table
```

```
##
```

```
## Response: log(Price)
```

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
## Make	25	354.44	14.178	481.9160	< 2.2e-16 ***
## Year	1	178.52	178.525	6068.2914	< 2.2e-16 ***
## Kilometer	1	0.98	0.982	33.3632	9.505e-09 ***
## Fuel.Type	2	19.23	9.615	326.8329	< 2.2e-16 ***
## Transmission	1	20.31	20.313	690.4496	< 2.2e-16 ***
## Location	63	16.76	0.266	9.0412	< 2.2e-16 ***
## Owner	3	0.46	0.153	5.2117	0.001402 **
## Seller.Type	2	0.33	0.165	5.6210	0.003708 **
## Max.Torque.RPM	1	1.67	1.669	56.7341	9.165e-14 ***
## Drivetrain	2	5.81	2.905	98.7377	< 2.2e-16 ***
## Length	1	51.72	51.724	1758.1791	< 2.2e-16 ***
## Width	1	5.06	5.063	172.1129	< 2.2e-16 ***
## Seating.Capacity	1	0.02	0.019	0.6504	0.420108
## Year:Kilometer	1	1.10	1.100	37.3998	1.262e-09 ***
## Year:Max.Torque.RPM	1	0.99	0.993	33.7697	7.752e-09 ***
## Year:Length	1	2.04	2.038	69.2576	< 2.2e-16 ***
## Max.Torque.RPM:Length	1	0.27	0.270	9.1768	0.002498 **
## Residuals	1332	39.19	0.029		

```
## ---
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Seems reasonable now. We will double check our assumptions

```
vif(model, type = "predictor")
```

```
## GVIFs computed for predictors
```

	GVIF	Df	GVIF ^{1/(2*Df)}
## Make	119.109068	25	1.100320
## Year	75.935730	8	1.310774
## Kilometer	330.635985	3	2.629612
## Fuel.Type	5.334325	2	1.519742
## Transmission	1.851632	1	1.360747
## Location	16.448508	63	1.022473
## Owner	1.673586	3	1.089619
## Seller.Type	1.341372	2	1.076186
## Max.Torque.RPM	207.518246	6	1.559869
## Drivetrain	4.003689	2	1.414539
## Length	207.518246	6	1.559869
## Width	4.624641	1	2.150498
## Seating.Capacity	2.882982	1	1.697934
##			Interacts With
## Make			--

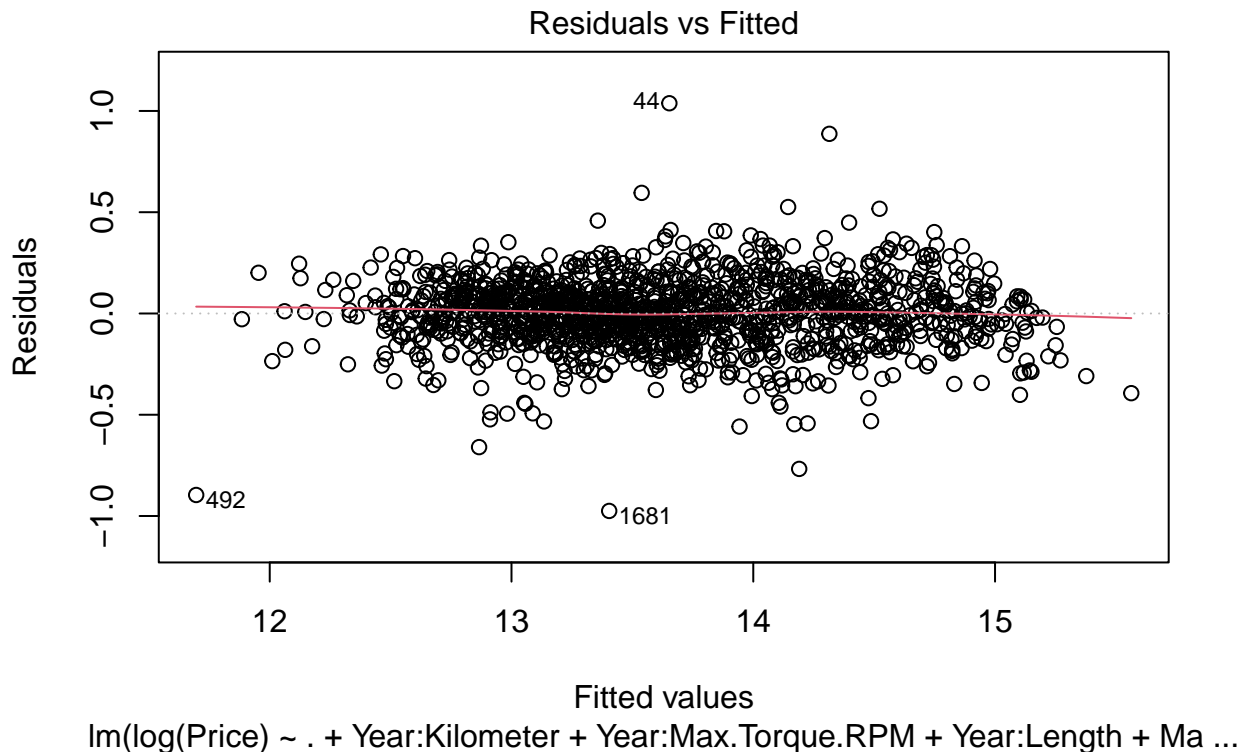
```

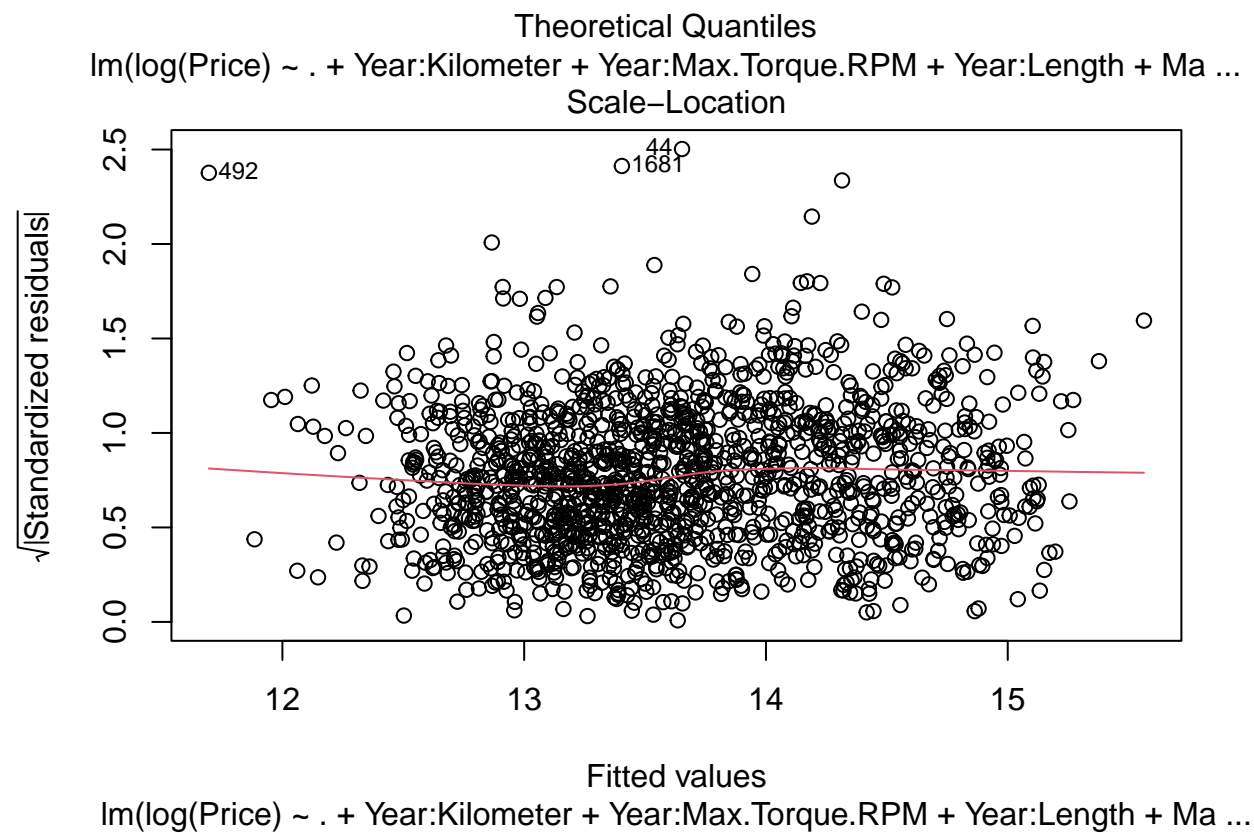
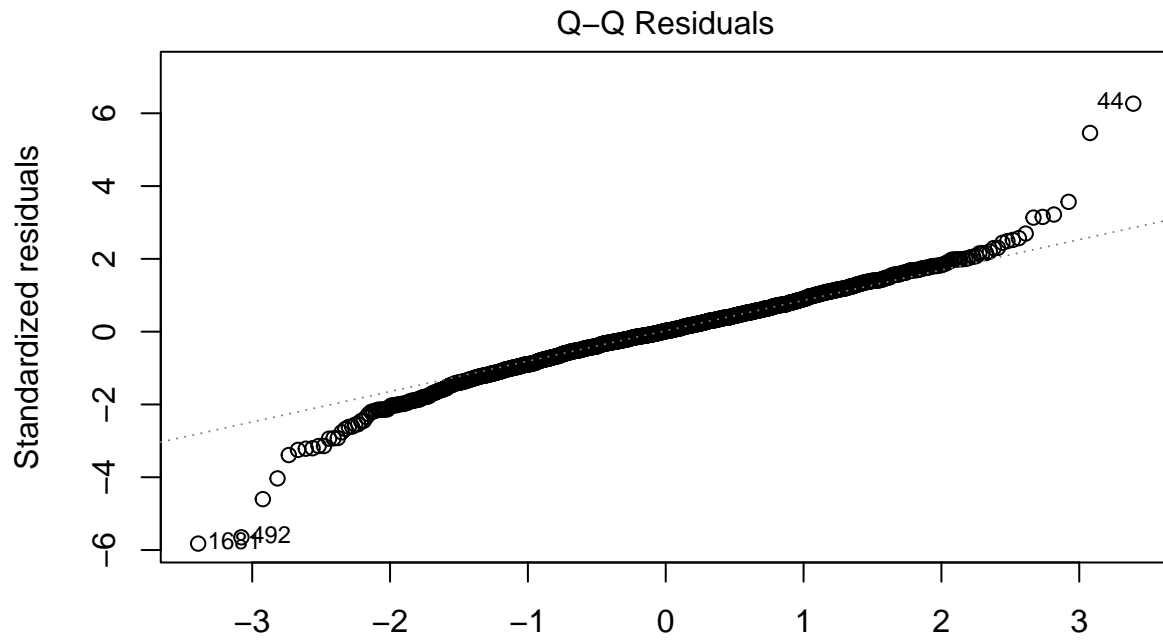
## Year          Kilometer, Max.Torque.RPM, Length
## Kilometer          Year
## Fuel.Type          --
## Transmission          --
## Location          --
## Owner          --
## Seller.Type          --
## Max.Torque.RPM          Year, Length
## Drivetrain          --
## Length          Year, Max.Torque.RPM
## Width          --
## Seating.Capacity          --
##
## Make          Year, Kilometer, Fuel.Type, Transmission, Location, Owner, Seller.Type, Max.Torque.RPM
## Year          Make, Fuel.Type, Transmission, Location, Owner, Seller.Type, Max.Torque.RPM
## Kilometer          Make, Fuel.Type, Transmission, Location, Owner, Seller.Type, Max.Torque.RPM
## Fuel.Type          Make, Year, Kilometer, Transmission, Location, Owner, Seller.Type, Max.Torque.RPM
## Transmission          Make, Year, Kilometer, Fuel.Type, Location, Owner, Seller.Type, Max.Torque.RPM
## Location          Make, Year, Kilometer, Fuel.Type, Transmission, Owner, Seller.Type, Max.Torque.RPM
## Owner          Make, Year, Kilometer, Fuel.Type, Transmission, Location, Seller.Type, Max.Torque.RPM
## Seller.Type          Make, Year, Kilometer, Fuel.Type, Transmission, Location, Owner, Max.Torque.RPM
## Max.Torque.RPM          Make, Kilometer, Fuel.Type, Transmission, Location, Owner, Seller.Type, Max.Torque.RPM
## Drivetrain          Make, Year, Kilometer, Fuel.Type, Transmission, Location, Owner, Seller.Type, Max.Torque.RPM
## Length          Make, Kilometer, Fuel.Type, Transmission, Location, Owner, Seller.Type, Max.Torque.RPM
## Width          Make, Year, Kilometer, Fuel.Type, Transmission, Location, Owner, Seller.Type, Max.Torque.RPM
## Seating.Capacity          Make, Year, Kilometer, Fuel.Type, Transmission, Location, Owner, Seller.Type, Max.Torque.RPM

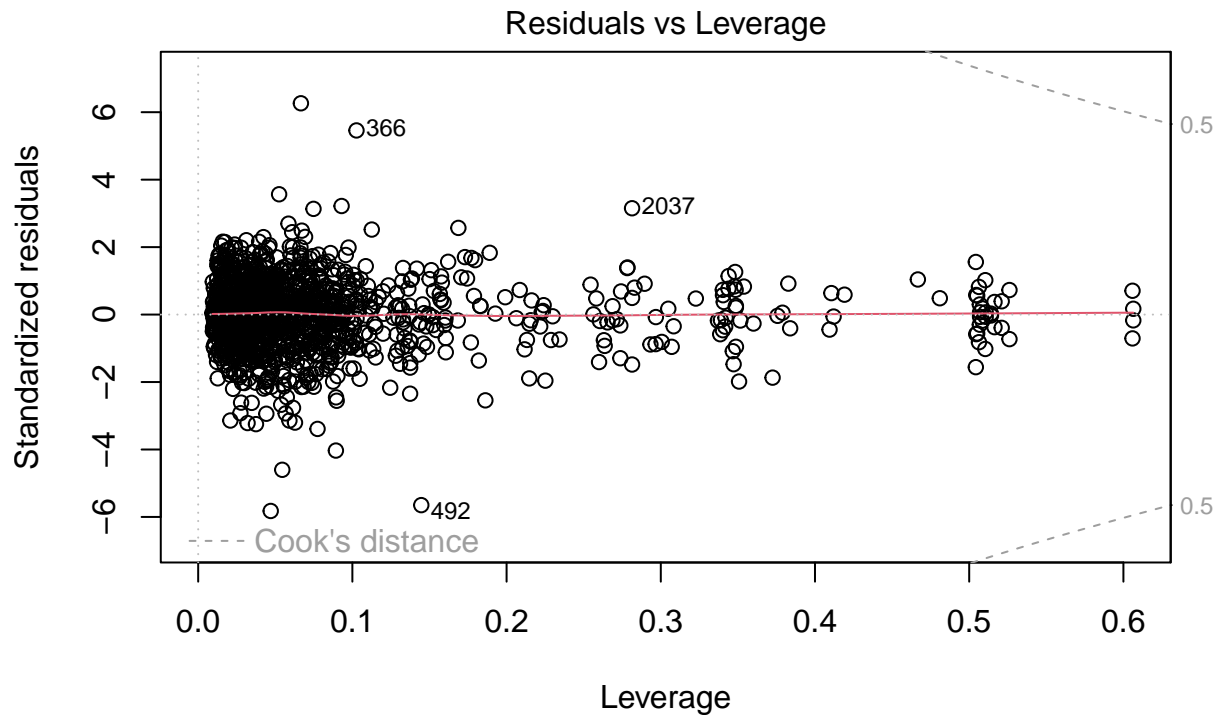
```

We get that $GVIF^{(1/(2 \cdot Df))} > 2.236068 \approx \sqrt{5}$, for “Kilometer,” but this is known to be important is does not exceed the threshold by a lot, so we keep it.

```
plot(model)
```







lm(log(Price) ~ . + Year:Kilometer + Year:Max.Torque.RPM + Year:Length + Ma ...

Backwards Stepwise Search

We do a backwards search since our model already conforms to the linear assumptions and is performing well. We simply wish to reduce the model size now.

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

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

```
## [1] "LocationAhmedabad" "LocationAllahabad" "LocationAmbala Cantt"
## [4] "LocationAmritsar" "LocationAurangabad" "LocationBangalore"
## [7] "LocationBhopal" "LocationBhubaneswar" "LocationBulandshahar"
## [10] "LocationChandigarh" "LocationChennai" "LocationCoimbatore"
## [13] "LocationDehradun" "LocationDelhi" "LocationDharwad"
## [16] "LocationErnakulam" "LocationFaridabad" "LocationGhaziabad"
## [19] "LocationGoa" "LocationGurgaon" "LocationGuwahati"
## [22] "LocationHaldwani" "LocationHyderabad" "LocationIndore"
## [25] "LocationJaipur" "LocationJalandhar" "LocationJamshedpur"
## [28] "LocationKanpur" "LocationKarnal" "LocationKharar"
## [31] "LocationKheda" "LocationKolkata" "LocationKollam"
## [34] "LocationKota" "LocationLucknow" "LocationLudhiana"
## [37] "LocationMangalore" "LocationMeerut" "LocationMirzapur"
## [40] "LocationMohali" "LocationMumbai" "LocationMuzaffarpur"
## [43] "LocationMysore" "LocationNagpur" "LocationNashik"
## [46] "LocationNavi Mumbai" "LocationNoida" "LocationPanchkula"
## [49] "LocationPatna" "LocationPune" "LocationRaipur"
## [52] "LocationRanchi" "LocationRanga Reddy" "LocationRoorkee"
## [55] "LocationSalem" "LocationSurat" "LocationThane"
```

```
## [58] "LocationUdupi"          "LocationVadodara"      "LocationVaranasi"
## [61] "LocationWarangal"       "LocationYamunanagar"   "LocationZirakpur"
## [64] "OwnerSecond"           "OwnerThird"            "OwnerUnRegistered Car"
## [67] "Seller.TypeCorporate"   "Seller.TypeIndividual" "Seating.Capacity"
```

Inspect Model Coefficients

```
summary(reduced_model)
```

```
##
## Call:
## lm(formula = log(Price) ~ Make + Year + Kilometer + Fuel.Type +
##     Transmission + Max.Torque.RPM + Drivetrain + Length + Width +
##     Year:Kilometer + Year:Max.Torque.RPM + Year:Length + Max.Torque.RPM:Length,
##     data = x)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1.08307 -0.10778  0.00302  0.11095  1.08538
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    9.510e+01  4.616e+01   2.060 0.039566 *
## MakeBMW         1.874e-02  4.321e-02   0.434 0.664637
## MakeChevrolet   -8.118e-01  9.967e-02  -8.144 8.35e-16 ***
## MakeDatsun      -9.312e-01  7.499e-02 -12.418 < 2e-16 ***
## MakeFord        -5.886e-01  4.134e-02 -14.237 < 2e-16 ***
## MakeHonda       -5.243e-01  3.348e-02 -15.659 < 2e-16 ***
## MakeHyundai     -4.913e-01  2.934e-02 -16.744 < 2e-16 ***
## MakeIsuzu       -4.104e-01  1.386e-01  -2.960 0.003124 **
## MakeJaguar       4.304e-02  8.415e-02   0.511 0.609111
## MakeJeep        -3.526e-01  5.495e-02  -6.418 1.89e-10 ***
## MakeKia         -4.353e-01  4.853e-02  -8.970 < 2e-16 ***
## MakeLand Rover   8.907e-02  1.132e-01   0.787 0.431440
## MakeMahindra    -6.243e-01  3.534e-02 -17.663 < 2e-16 ***
## MakeMaruti Suzuki -5.409e-01  3.008e-02 -17.979 < 2e-16 ***
## MakeMercedes-Benz 1.231e-01  3.561e-02   3.457 0.000563 ***
## MakeMG          -6.054e-01  5.996e-02 -10.095 < 2e-16 ***
## MakeMINI        7.174e-01  8.298e-02   8.645 < 2e-16 ***
## MakeMitsubishi  -2.063e-01  1.137e-01  -1.814 0.069817 .
## MakeNissan      -7.579e-01  5.559e-02 -13.634 < 2e-16 ***
## MakeRenault     -7.602e-01  4.110e-02 -18.499 < 2e-16 ***
## MakeSkoda       -4.933e-01  4.013e-02 -12.292 < 2e-16 ***
## MakeSsangyong   -8.659e-01  1.144e-01  -7.570 6.74e-14 ***
## MakeTata        -7.449e-01  3.963e-02 -18.799 < 2e-16 ***
## MakeToyota     -2.811e-01  3.210e-02  -8.757 < 2e-16 ***
## MakeVolkswagen  -5.047e-01  4.029e-02 -12.525 < 2e-16 ***
## MakeVolvo       4.204e-02  6.144e-02   0.684 0.493988
## Year           -4.268e-02  2.289e-02  -1.865 0.062445 .
## Kilometer      -3.578e-04  1.296e-04  -2.761 0.005844 **
## Fuel.TypeDiesel -1.317e-01  4.020e-02  -3.277 0.001075 **
## Fuel.TypePetrol -1.608e-01  3.648e-02  -4.409 1.12e-05 ***
## TransmissionManual -1.581e-01  1.340e-02 -11.799 < 2e-16 ***
## Max.Torque.RPM  1.250e-02  3.153e-03   3.965 7.72e-05 ***
```



```
## DrivetrainFWD      -2.699e-01  2.396e-02 -11.268 < 2e-16 ***
## DrivetrainRWD      -2.116e-01  2.905e-02  -7.283 5.44e-13 ***
## Length             -7.724e-02  1.032e-02  -7.484 1.26e-13 ***
## Width              1.540e-03  1.163e-04  13.239 < 2e-16 ***
## Year:Kilometer      1.768e-07  6.427e-08   2.751 0.006026 **
## Year:Max.Torque.RPM -6.132e-06  1.559e-06  -3.933 8.82e-05 ***
## Year:Length         3.866e-05  5.116e-06   7.555 7.51e-14 ***
## Max.Torque.RPM:Length -3.970e-08  1.460e-08  -2.719 0.006627 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1905 on 1401 degrees of freedom
## Multiple R-squared:  0.9273, Adjusted R-squared:  0.9253
## F-statistic: 458.2 on 39 and 1401 DF,  p-value: < 2.2e-16
```

```
anova(reduced_model)
```

```
## Analysis of Variance Table
```

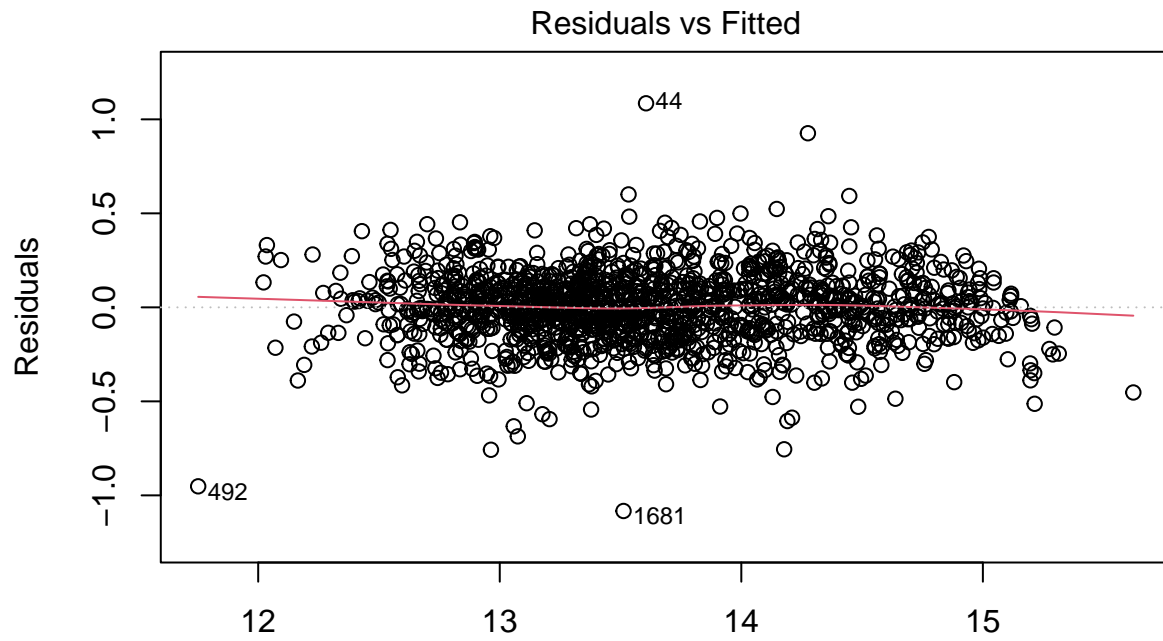
```
##
```

```
## Response: log(Price)
```

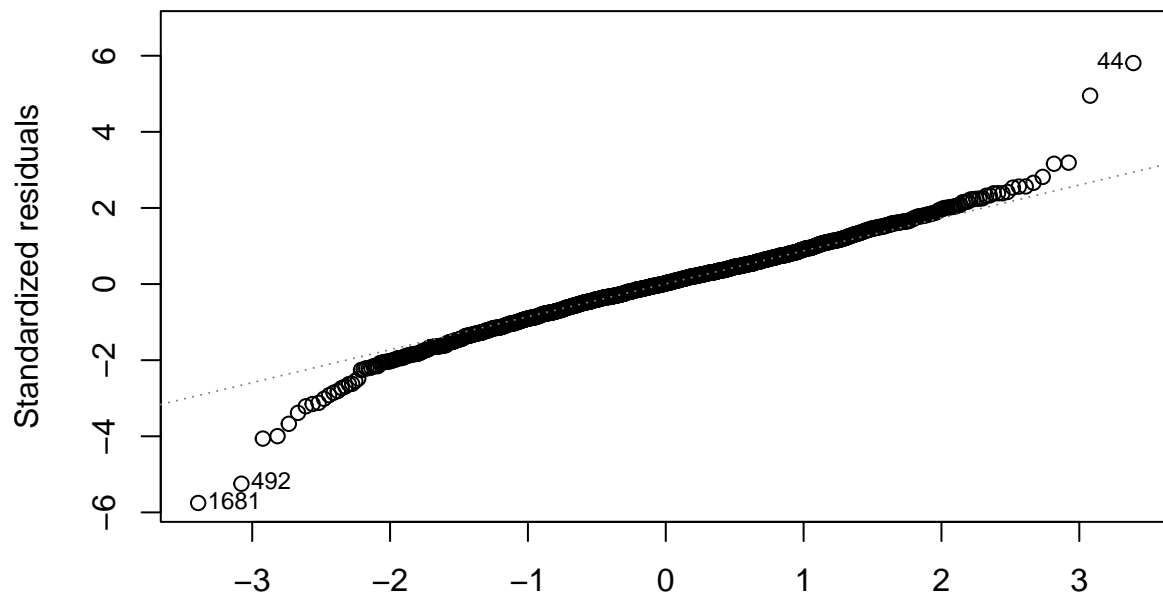
```
##              Df Sum Sq Mean Sq  F value    Pr(>F)
## Make          25 354.44   14.178   390.8754 < 2.2e-16 ***
## Year           1 178.52  178.525 4921.9071 < 2.2e-16 ***
## Kilometer      1   0.98   0.982   27.0604 2.265e-07 ***
## Fuel.Type      2  19.23   9.615  265.0896 < 2.2e-16 ***
## Transmission   1  20.31  20.313  560.0141 < 2.2e-16 ***
## Max.Torque.RPM 1   1.14   1.139   31.4057 2.518e-08 ***
## Drivetrain     2   6.53   3.267   90.0706 < 2.2e-16 ***
## Length         1  56.67  56.670 1562.3991 < 2.2e-16 ***
## Width          1   5.64   5.636  155.3813 < 2.2e-16 ***
## Year:Kilometer  1   0.92   0.922   25.4311 5.184e-07 ***
## Year:Max.Torque.RPM 1   1.25   1.247   34.3901 5.618e-09 ***
## Year:Length     1   2.19   2.187   60.3045 1.559e-14 ***
## Max.Torque.RPM:Length 1   0.27   0.268    7.3935 0.006627 **
## Residuals     1401  50.82   0.036
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Final Model Verification

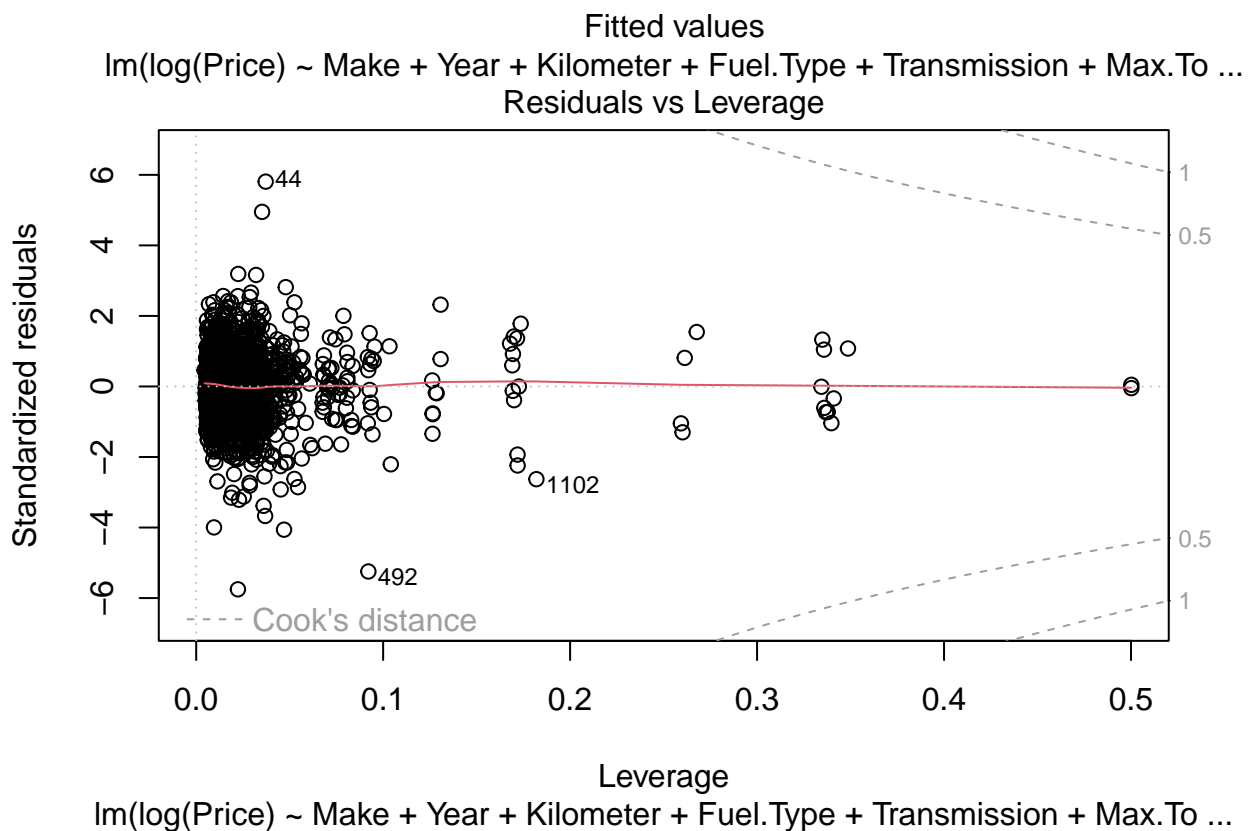
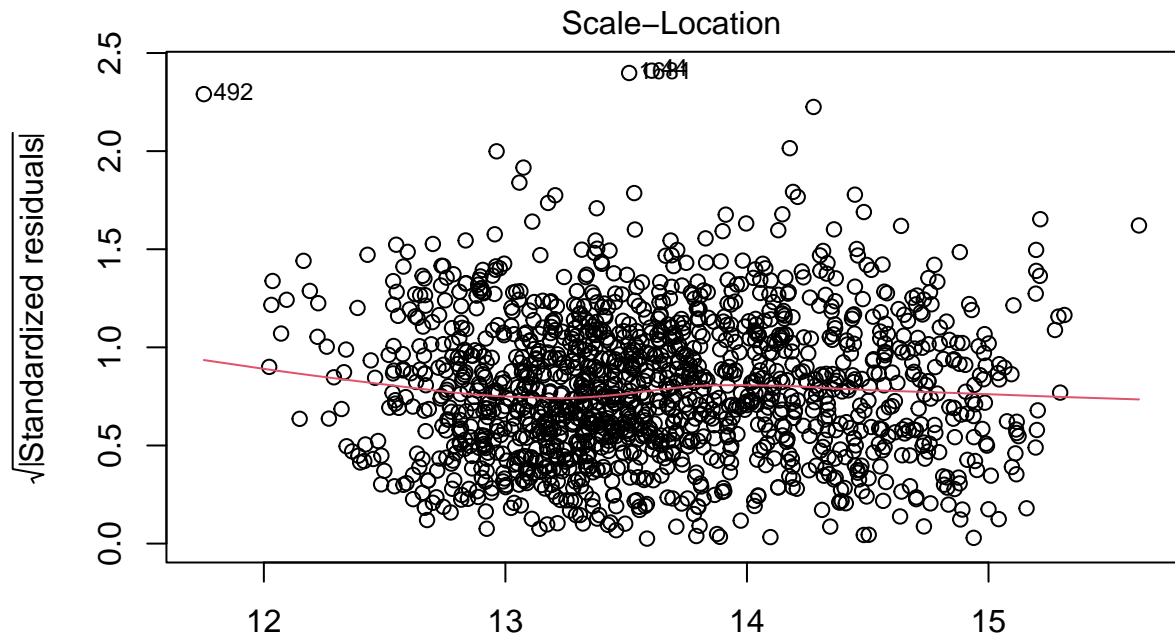
```
plot(reduced_model)
```



Fitted values
 $\text{lm}(\log(\text{Price}) \sim \text{Make} + \text{Year} + \text{Kilometer} + \text{Fuel.Type} + \text{Transmission} + \text{Max.To ...)$
 Q-Q Residuals



Theoretical Quantiles
 $\text{lm}(\log(\text{Price}) \sim \text{Make} + \text{Year} + \text{Kilometer} + \text{Fuel.Type} + \text{Transmission} + \text{Max.To ...)$



Normalized Coefficients

```
normalized <- remove_cols(x, old[!(old %in% new)]) # removed from multicollinearity test
normalized <- remove_cols(x, c("Seating.Capacity", "Seller.Type"))
```

```

for (col in colnames(normalized)) {
  if (col != "Price" && !(col %in% categorical)) {
    normalized[, col] <- (normalized[, col] - mean(normalized[, col])) / sd(normalized[, col])
  }
}

normalized_model <- lm(log(Price) ~ . + Year:Kilometer + Year:Max.Torque.RPM
  + Year:Length + Max.Torque.RPM:Length, data = normalized)

sort(abs(normalized_model$coefficients), decreasing = T)

```

##	(Intercept)	MakeSsangyong	MakeDatsun
##	14.479239336	0.872512749	0.857115032
##	MakeChevrolet	MakeRenault	MakeNissan
##	0.798238688	0.749256440	0.733447399
##	MakeTata	MakeMINI	MakeMG
##	0.718998608	0.704054157	0.596345412
##	MakeMahindra	MakeFord	MakeMaruti Suzuki
##	0.596297888	0.572746578	0.523222676
##	MakeVolkswagen	MakeSkoda	MakeHonda
##	0.518324656	0.509078901	0.485869081
##	MakeHyundai	MakeIsuzu	LocationSalem
##	0.465042221	0.437700433	0.420870279
##	MakeKia	MakeJeep	LocationMysore
##	0.415501427	0.333574675	0.320886296
##	Year	LocationBulandshahar	LocationAurangabad
##	0.313783661	0.313704055	0.293113281
##	LocationCoimbatore	DrivetrainFWD	MakeMitsubishi
##	0.269393557	0.263692508	0.249563035
##	MakeToyota	LocationGoa	LocationBangalore
##	0.248826742	0.248093841	0.246496783
##	LocationKollam	Length	DrivetrainRWD
##	0.226788237	0.226439416	0.221018253
##	LocationChennai	LocationIndore	LocationErnakulam
##	0.216077914	0.209724881	0.208988890
##	LocationDharwad	Fuel.TypePetrol	LocationHyderabad
##	0.207341744	0.189472148	0.185525949
##	LocationJalandhar	MakeLand Rover	LocationNagpur
##	0.181224043	0.180846132	0.167588274
##	LocationGuwahati	Fuel.TypeDiesel	LocationUdupi
##	0.167182283	0.164673775	0.148042941
##	MakeMercedes-Benz	TransmissionManual	Width
##	0.145432782	0.143902253	0.135371333
##	LocationPune	LocationJaipur	LocationAmritsar
##	0.131030713	0.122967074	0.121805119
##	OwnerThird	LocationKota	LocationMohali
##	0.121210169	0.117146416	0.115934495
##	LocationThane	LocationMumbai	LocationRaipur
##	0.102055354	0.100701797	0.099272043
##	LocationMirzapur	LocationKarnal	LocationNoida
##	0.098291704	0.096670247	0.095972455
##	LocationAmbala Cantt	LocationNavi Mumbai	LocationAllahabad
##	0.092130013	0.092020794	0.090599537

##	LocationKheda	LocationKanpur	OwnerUnRegistered Car
##	0.088221751	0.086058622	0.082215522
##	LocationBhubaneswar	LocationKolkata	LocationDelhi
##	0.080119830	0.076002757	0.071607021
##	LocationFaridabad	LocationMuzaffarpur	LocationRanga Reddy
##	0.069764717	0.068179314	0.066742019
##	MakeJaguar	LocationWarangal	LocationAhmedabad
##	0.065901543	0.063180846	0.062246218
##	LocationPatna	LocationDehradun	LocationNashik
##	0.062160128	0.059305473	0.058196769
##	Max.Torque.RPM	LocationMangalore	LocationKharar
##	0.056697498	0.056620215	0.056298195
##	LocationVaranasi	LocationRoorkee	LocationRanchi
##	0.048878372	0.047629067	0.046096495
##	LocationLudhiana	Year:Length	LocationSurat
##	0.043273053	0.041510876	0.041112278
##	LocationMeerut	Kilometer	LocationChandigarh
##	0.038541407	0.034232361	0.031825300
##	MakeVolvo	LocationBhopal	LocationPanchkula
##	0.030938615	0.029053023	0.027400801
##	LocationZirakpur	LocationGhaziabad	LocationYamunanagar
##	0.026027821	0.023076512	0.023015393
##	LocationLucknow	OwnerSecond	Year:Max.Torque.RPM
##	0.021634758	0.020113875	0.018733584
##	Max.Torque.RPM:Length	Year:Kilometer	LocationJamshedpur
##	0.017715033	0.017494262	0.017254036
##	MakeBMW	LocationHaldwani	LocationGurgaon
##	0.012673995	0.010949026	0.010276737
##	LocationVadodara		
##	0.009552204		

Confidence Intervals

```
confint(reduced_model, level = 0.95)
```

##		2.5 %	97.5 %
##	(Intercept)	4.547396e+00	1.856477e+02
##	MakeBMW	-6.602817e-02	1.035016e-01
##	MakeChevrolet	-1.007281e+00	-6.162404e-01
##	MakeDatsun	-1.078298e+00	-7.841001e-01
##	MakeFord	-6.697386e-01	-5.075326e-01
##	MakeHonda	-5.900076e-01	-4.586369e-01
##	MakeHyundai	-5.488761e-01	-4.337524e-01
##	MakeIsuzu	-6.823497e-01	-1.384600e-01
##	MakeJaguar	-1.220379e-01	2.081188e-01
##	MakeJeep	-4.604327e-01	-2.448569e-01
##	MakeKia	-5.305453e-01	-3.401349e-01
##	MakeLand Rover	-1.329586e-01	3.111020e-01
##	MakeMahindra	-6.936231e-01	-5.549561e-01
##	MakeMaruti Suzuki	-5.998648e-01	-4.818426e-01
##	MakeMercedes-Benz	5.324385e-02	1.929706e-01
##	MakeMG	-7.229815e-01	-4.877246e-01
##	MakeMINI	5.546073e-01	8.801579e-01
##	MakeMitsubishi	-4.293712e-01	1.673447e-02

## MakeNissan	-8.669420e-01	-6.488544e-01
## MakeRenault	-8.408366e-01	-6.796076e-01
## MakeSkoda	-5.720421e-01	-4.145890e-01
## MakeSsangyong	-1.090227e+00	-6.414793e-01
## MakeTata	-8.226646e-01	-6.671974e-01
## MakeToyota	-3.440356e-01	-2.181091e-01
## MakeVolkswagen	-5.837269e-01	-4.256451e-01
## MakeVolvo	-7.849495e-02	1.625703e-01
## Year	-8.757891e-02	2.221399e-03
## Kilometer	-6.120361e-04	-1.035533e-04
## Fuel.TypeDiesel	-2.106099e-01	-5.288107e-02
## Fuel.TypePetrol	-2.323631e-01	-8.925600e-02
## TransmissionManual	-1.843935e-01	-1.318227e-01
## Max.Torque.RPM	6.315101e-03	1.868420e-02
## DrivetrainFWD	-3.169270e-01	-2.229383e-01
## DrivetrainRWD	-2.685680e-01	-1.545884e-01
## Length	-9.748973e-02	-5.699824e-02
## Width	1.311510e-03	1.767784e-03
## Year:Kilometer	5.069930e-08	3.028336e-07
## Year:Max.Torque.RPM	-9.190592e-06	-3.073084e-06
## Year:Length	2.861917e-05	4.869220e-05
## Max.Torque.RPM:Length	-6.834755e-08	-1.105992e-08