

# Fuel Economy Analysis

This demo is an example of performing data mining on historical fuel economy data. We have data from various cars built from year 2000 up to 2012.



## Import Data into Table

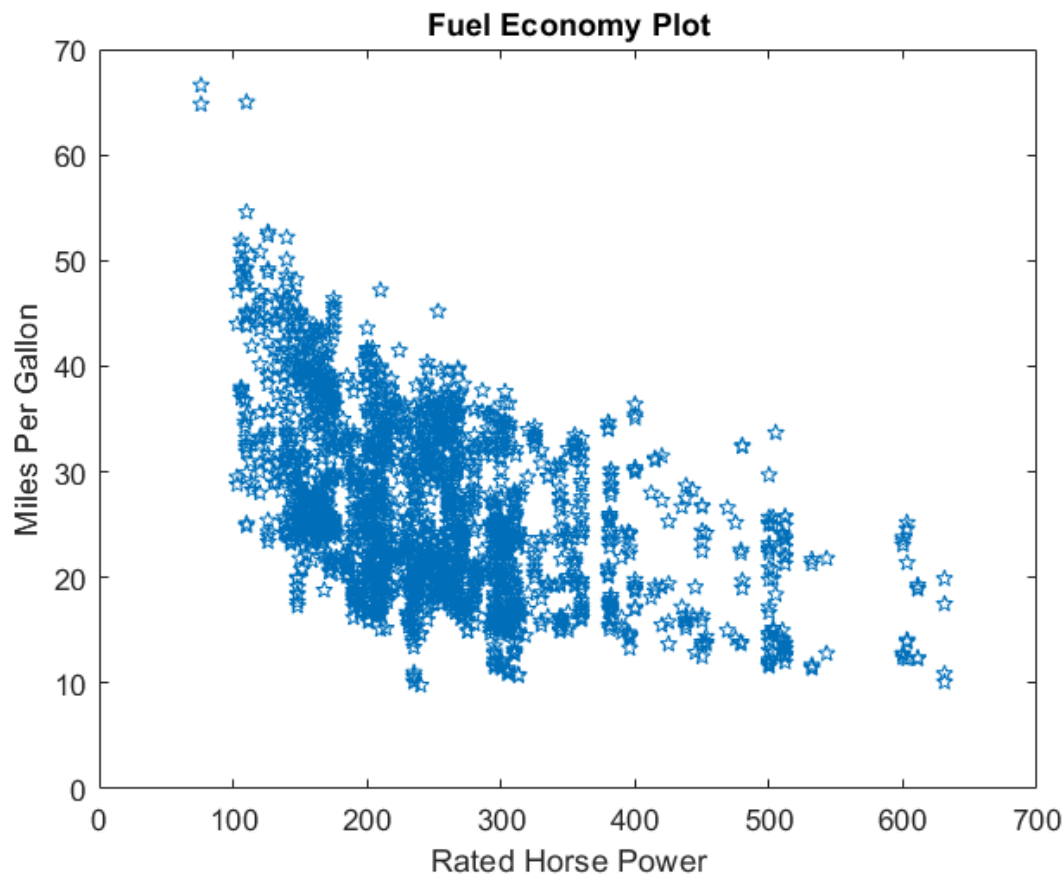
Import from Excel using modified auto-generated function from Import Tool

```
carData = importYearXLS(2007);
```

## Visualize

Plot MPG versus Rated Horsepower

```
createMPGfigure(carData.RatedHP, carData.MPG);
```



## Examine Grouping Effects of Categorical Data

In order to extract all "cars":

```
carIDs = carData.Car_Truck == "car";
```

In order to extract "city" data for "trucks":

```
city_truckIDs = (carData.City_Highway == "city" & carData.Car_Truck == "truck");
```

City versus Highway

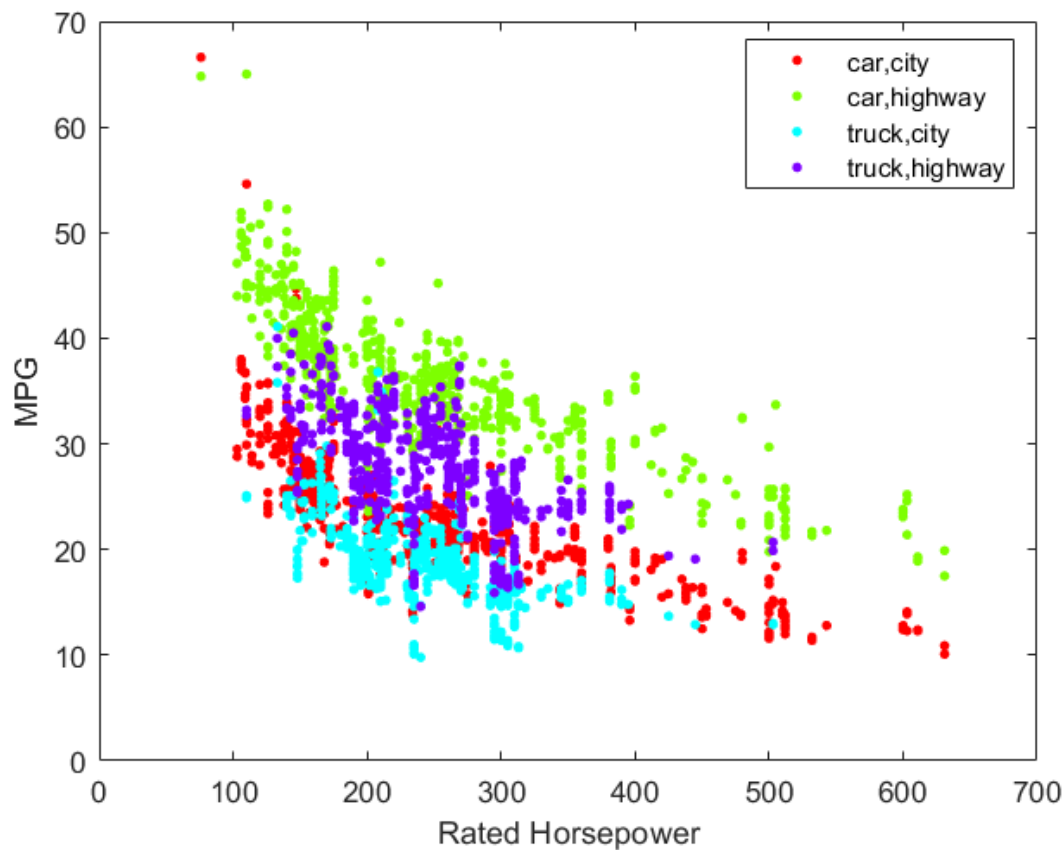
```
cityIDs = carData.City_Highway == "city";
highwayIDs = carData.City_Highway == "highway";
```

## Grouped Visualizations

Scatter plot by group.

```
figure
gscatter(carData.RatedHP, carData.MPG, ...
        {carData.Car_Truck, carData.City_Highway}, ...
```

```
','.', 10, 'on', 'Rated Horsepower', 'MPG')
```



## Extract Data for Curve Fitting

Create these variables for Curve Fitting App

```
RatedHPCity = carData.RatedHP(cityIDs);
MPGCity      = carData.MPG(cityIDs);

% Use the App to develop a curve fit.
```

## Curve Fitting

Equation:

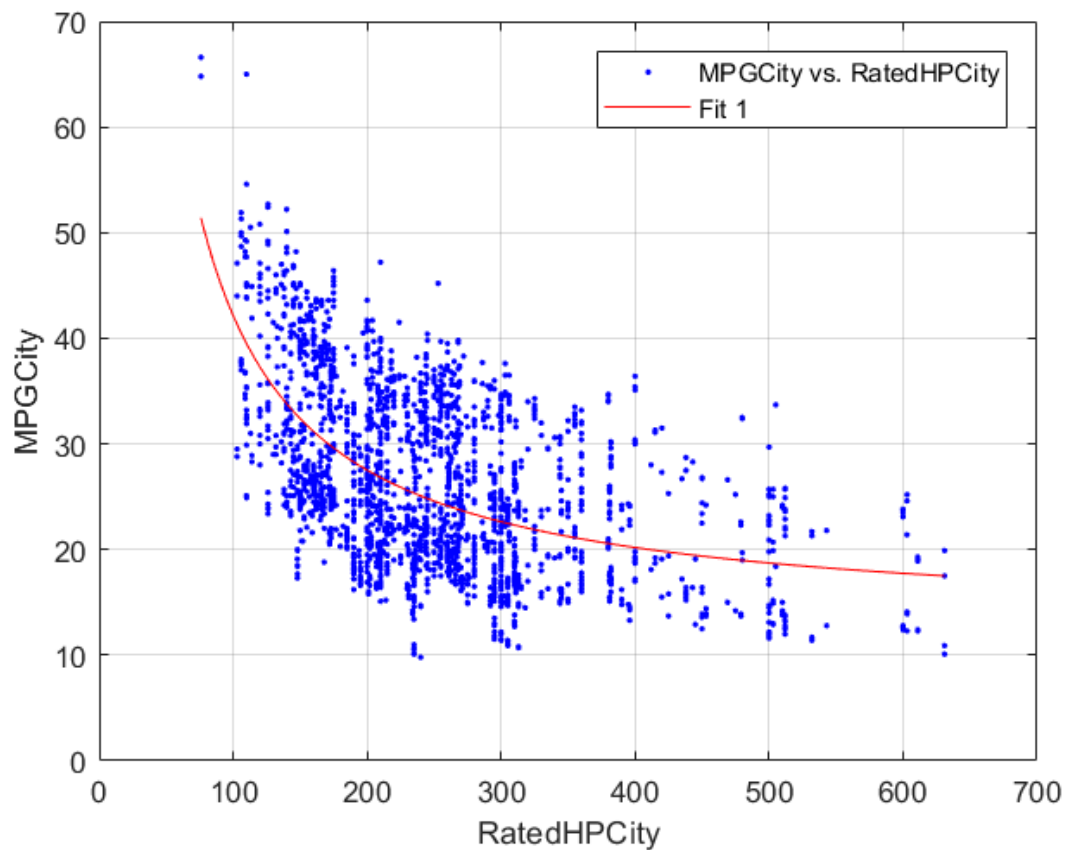
$$\text{MPG} = b_1 + \frac{b_2}{\text{RatedHP}}$$

We can solve this using the Curve Fitting Tool

```
cftool(carData.RatedHP, carData.MPG)
```

The following is a modified version of the auto-generated m-file from cftool.

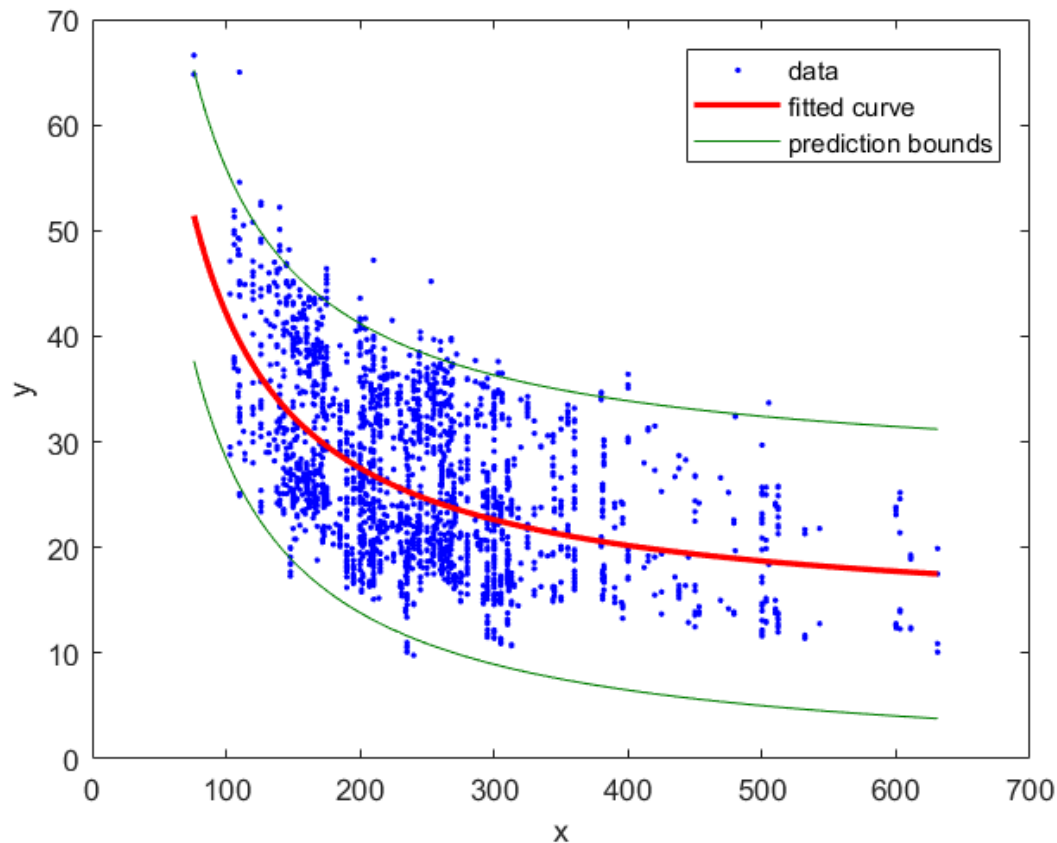
```
cf = createMPGFit(carData.RatedHP, carData.MPG);
```



## Plot Data and Model

The result from the Curve Fitting Toolbox has a `plot` method for displaying the result graphically. We can choose to display the prediction bounds for the fit.

```
figure
hh = plot(cf, 'r', carData.RatedHP, carData.MPG, 'predobs', 0.95);
hh(2).LineWidth = 2;
for ii = [3 4]
    hh(ii).LineStyle = '-';
    hh(ii).Color = [0 0.5 0];
end
```



## Plot of Data and Model (for different groups)

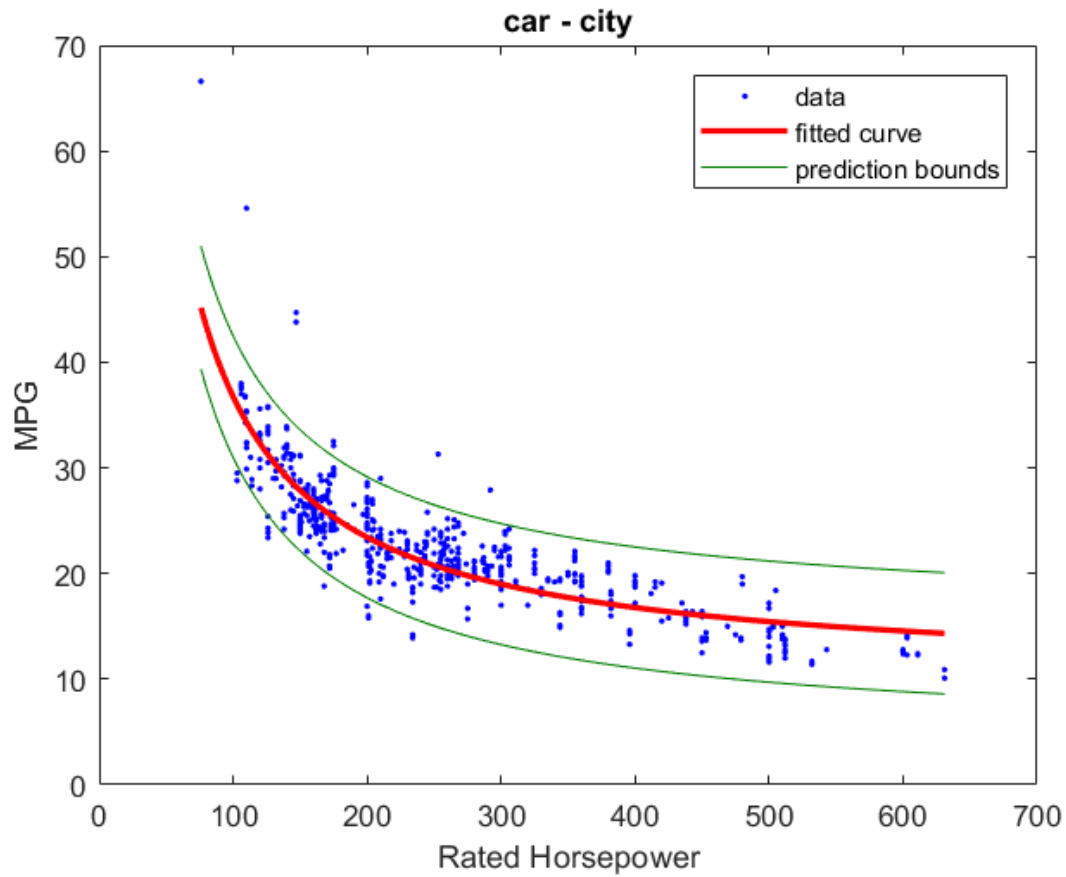
We will apply the similar modeling technique to the data for different combinations of groups (Car-Truck and City-Highway)

Model different combinations:

```
[mdl,gof] = modelMPG(carData, 'car', 'city')
```

```
mdl =
  Linear model:
  mdl(x) = a + b*1/x
  Coefficients (with 95% confidence bounds):
    a =      10.12  (9.528, 10.72)
    b =     2663  (2546, 2779)

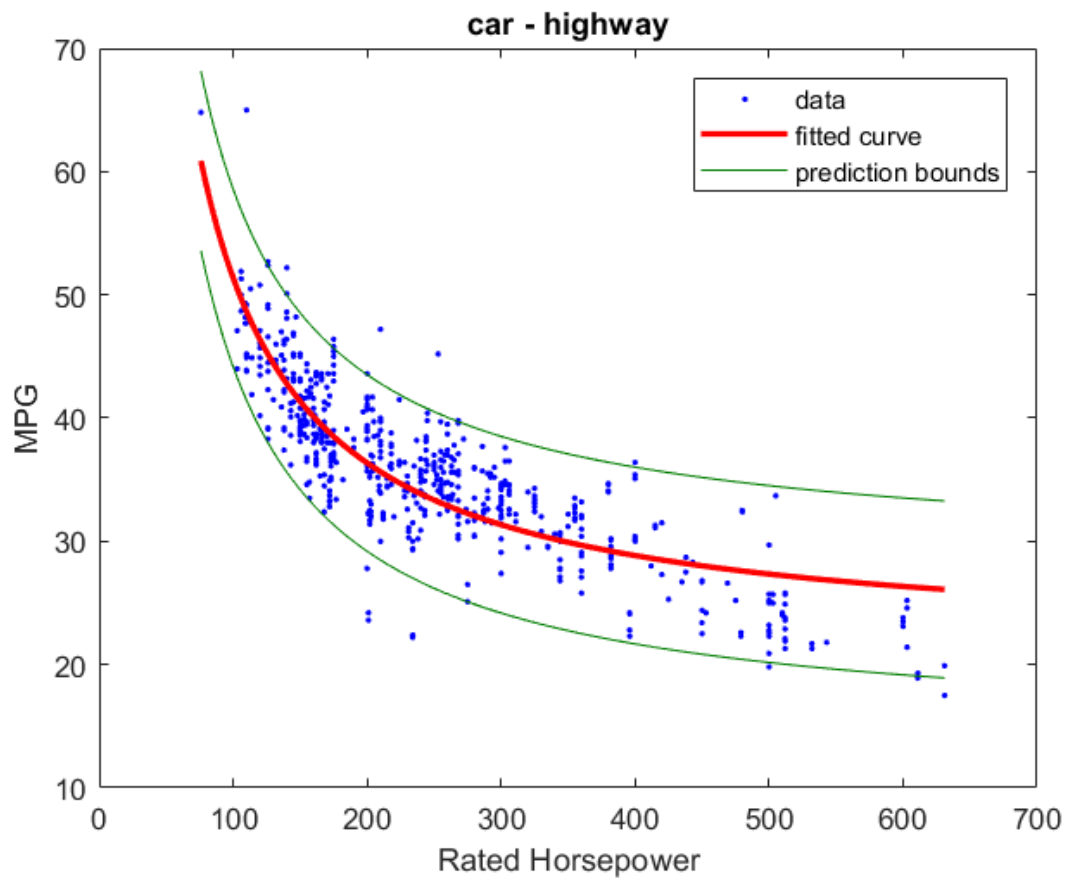
gof = struct with fields:
    sse: 5.6906e+03
   rsquare: 0.7496
    dfe: 670
  adjrsquare: 0.7493
    rmse: 2.9143
```



```
[mdl,gof] = modelMPG(carData, 'car', 'highway')
```

```
mdl =
  Linear model:
  mdl(x) = a + b*1/x
  Coefficients (with 95% confidence bounds):
    a =      21.33  (20.58, 22.09)
    b =     3005   (2857, 3153)

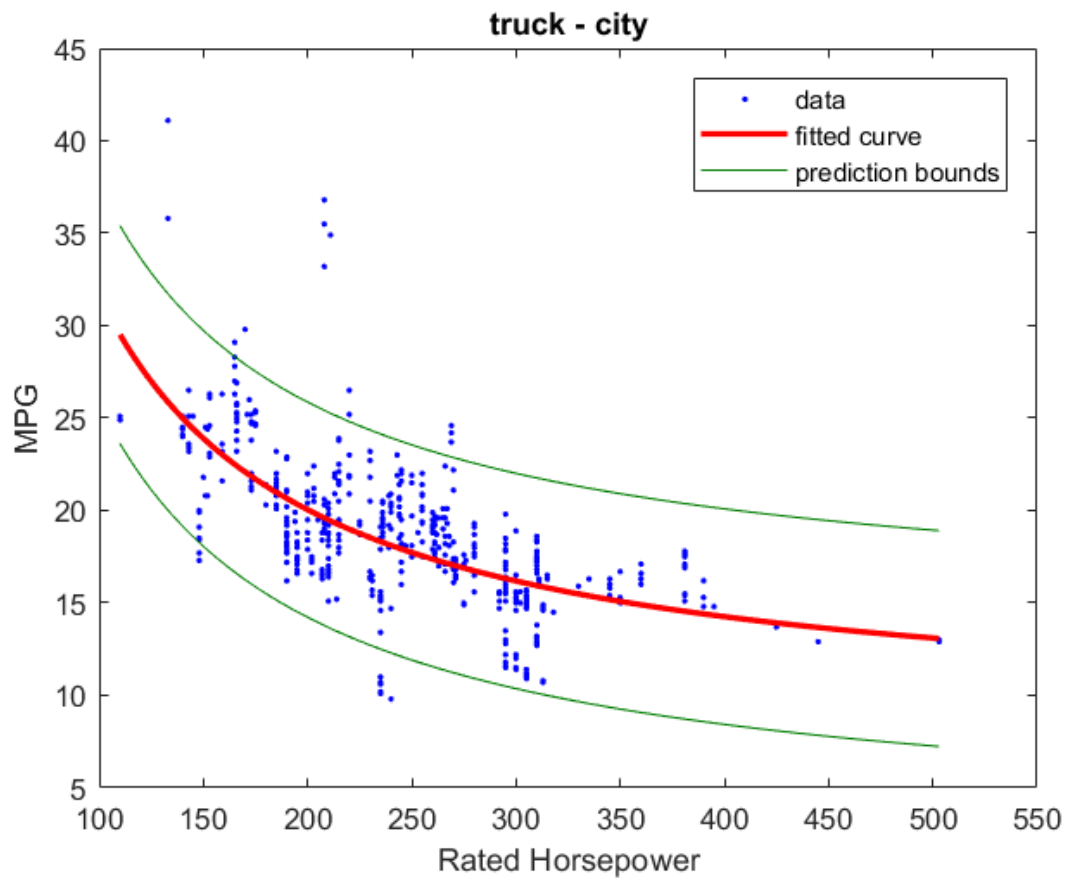
gof = struct with fields:
    sse: 8.8948e+03
   rsquare: 0.7033
    dfe: 669
 adjrsquare: 0.7028
    rmse: 3.6463
```



```
[mdl,gof] = modelMPG(carData, 'truck', 'city')
```

```
mdl =
  Linear model:
  mdl(x) = a + b*1/x
  Coefficients (with 95% confidence bounds):
    a =      8.473  (7.579, 9.368)
    b =     2314   (2115, 2514)

gof = struct with fields:
    sse: 5.4782e+03
   rsquare: 0.4540
    dfe: 625
  adjrsquare: 0.4531
    rmse: 2.9606
```

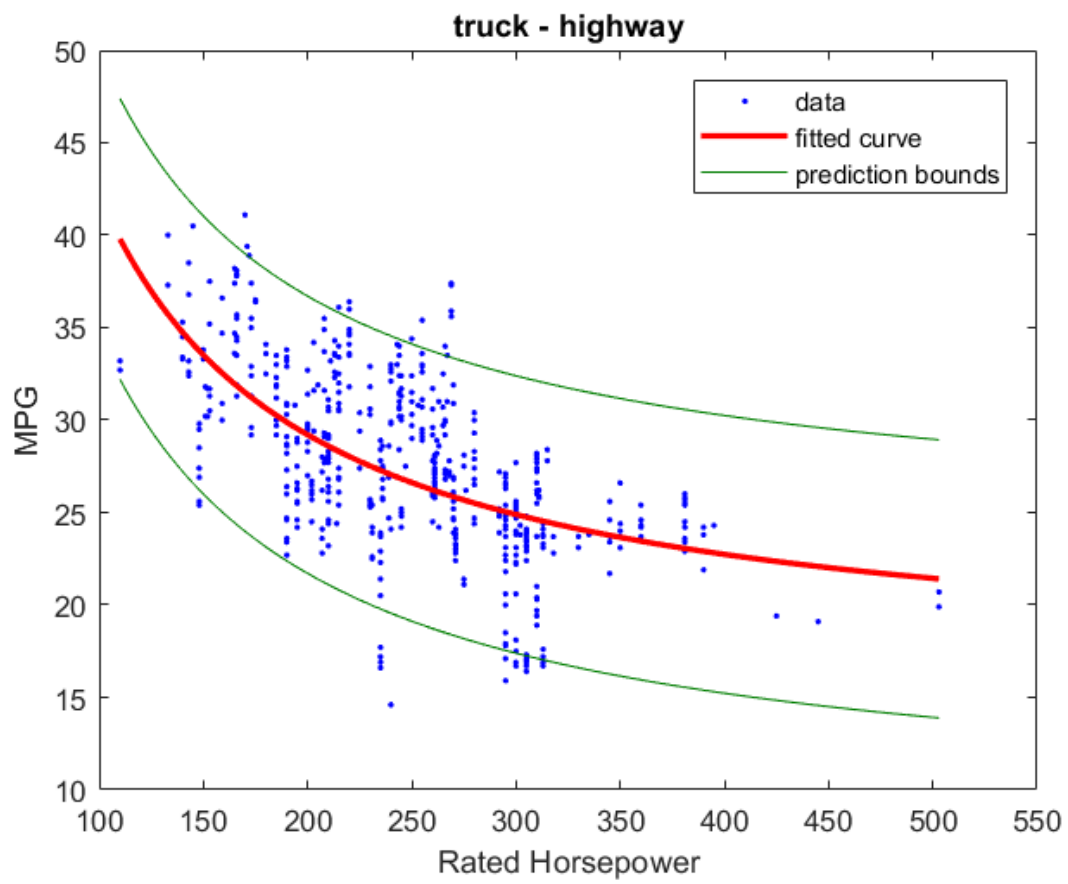


```
[mdl,gof] = modelMPG(carData, 'truck', 'highway')
```

```
mdl =
  Linear model:
  mdl(x) = a + b*1/x
  Coefficients (with 95% confidence bounds):
    a =      16.26  (15.11, 17.42)
    b =     2589   (2332, 2846)

gof = struct with fields:
    sse: 9.0679e+03
   rsquare: 0.3857
    dfe: 623
  adjrsquare: 0.3847
    rmse: 3.8151
```





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