

# 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

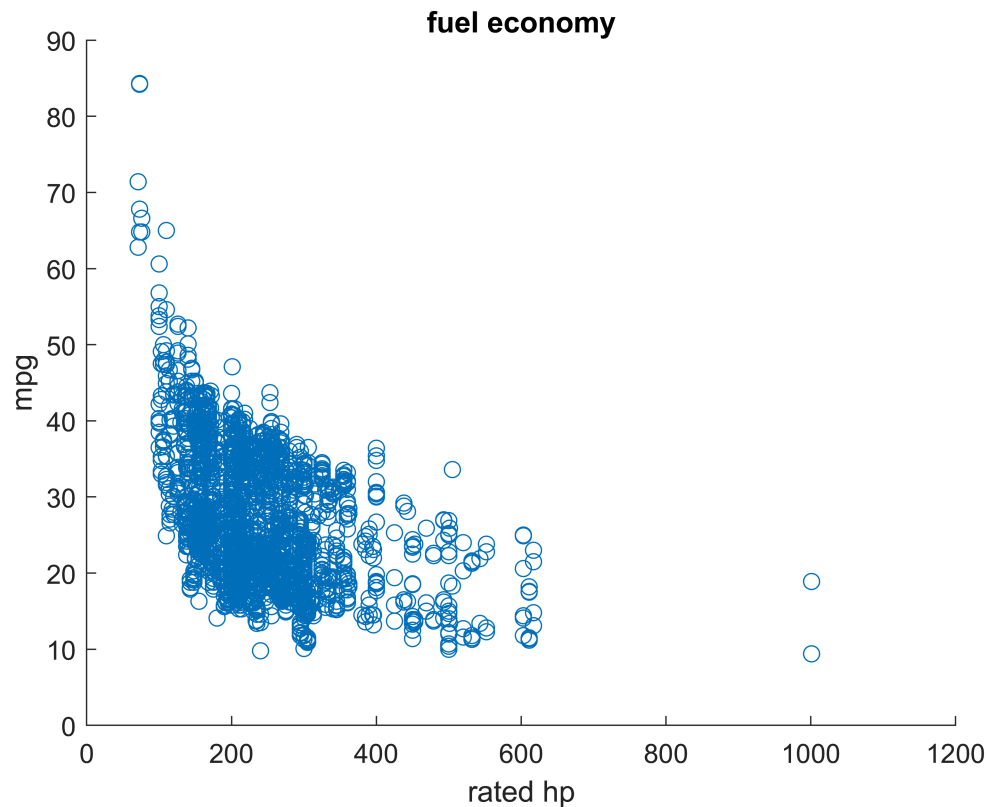
Import from Excel using modified auto-generated function from Import Tool

```
carData = importfile('2006dat.xlsx');
```

## Visualize

Plot MPG versus Rated Horsepower

```
createfigure(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");
```

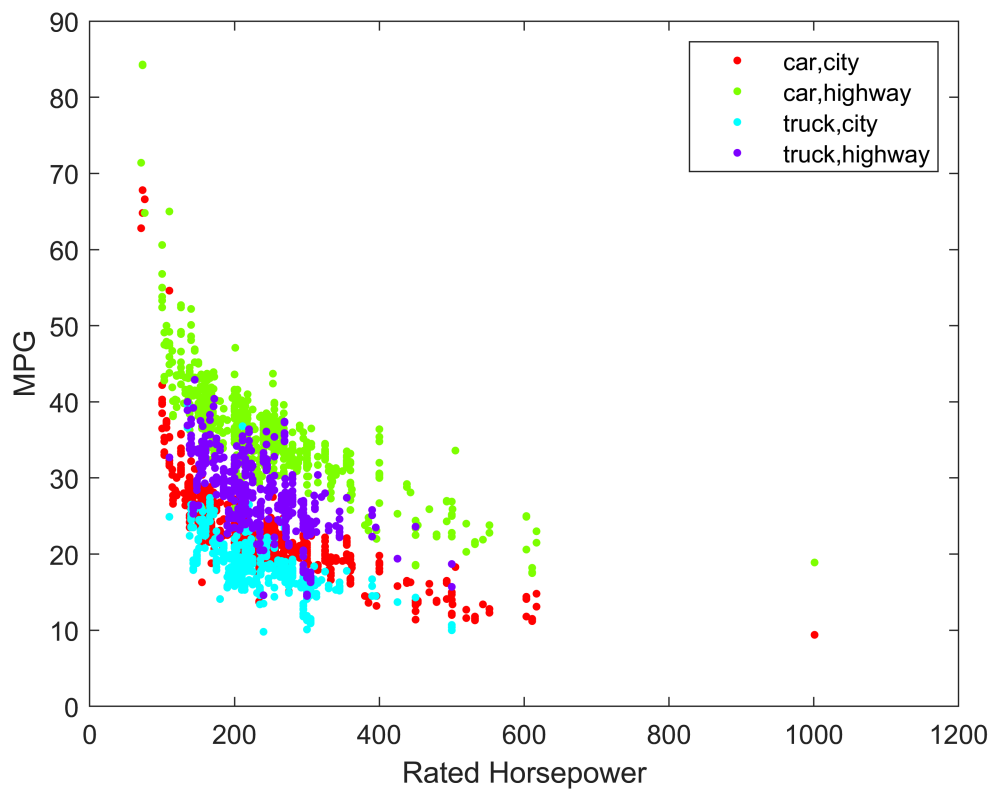
City versus Highway

```
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:

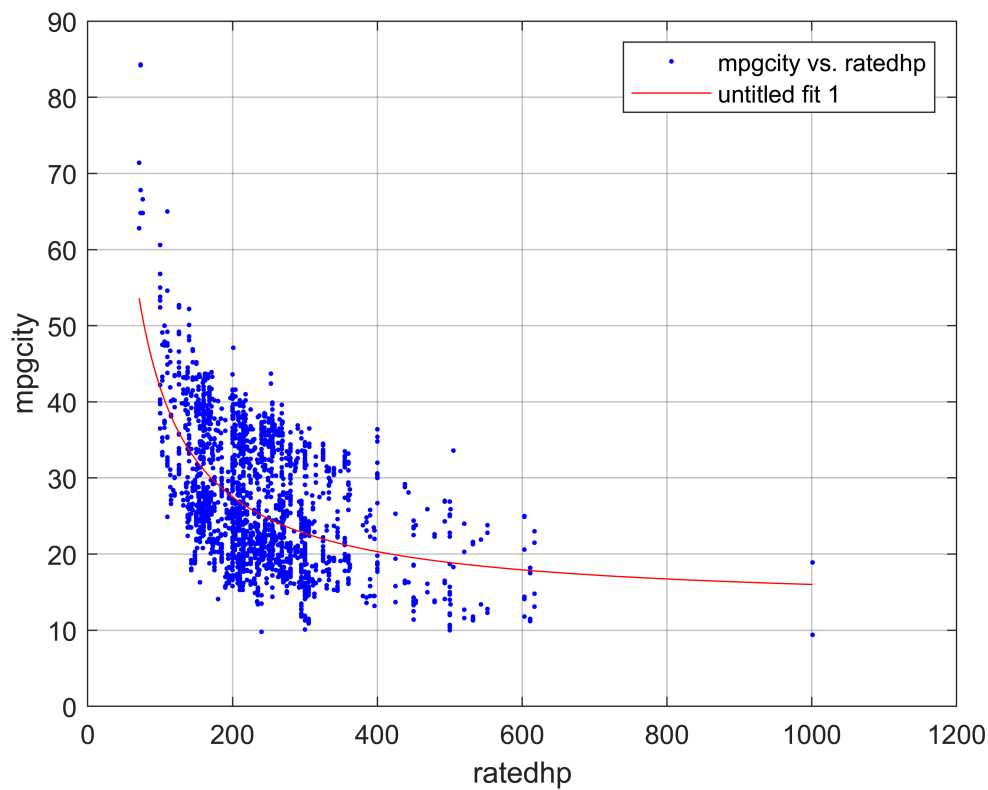
$$\text{MPG} = b_1 + \frac{b_2}{\text{RatedHP}}$$

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.

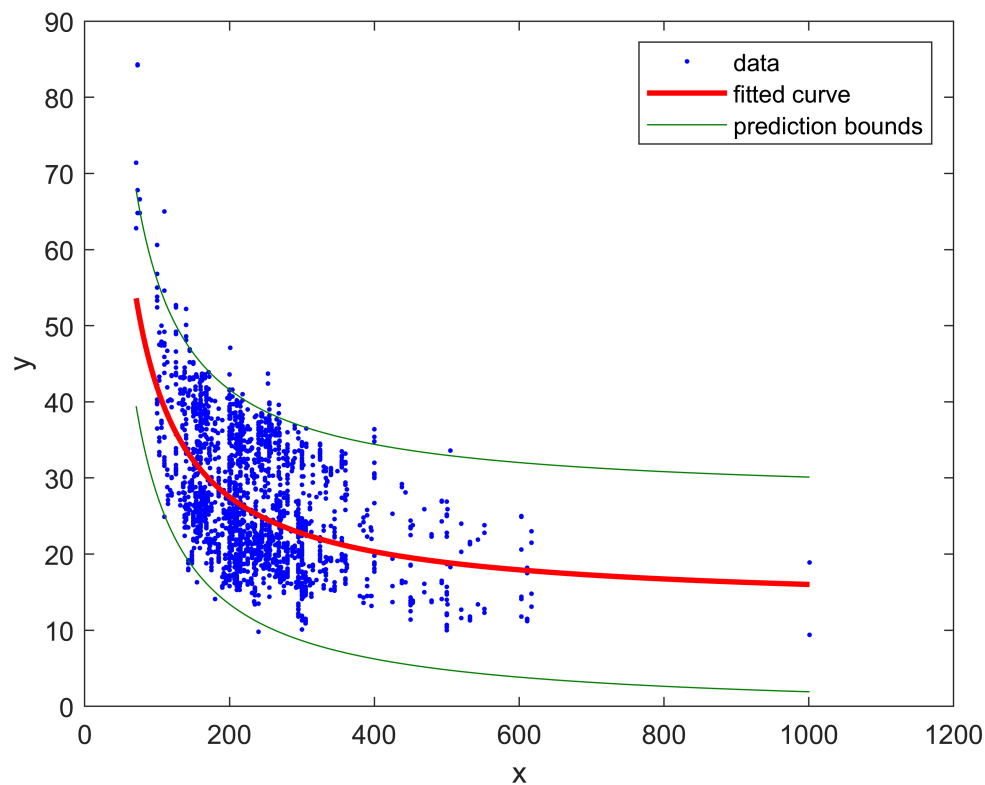
```
cf = createFit(carData.RatedHP, carData.MPG);
```



## 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
```

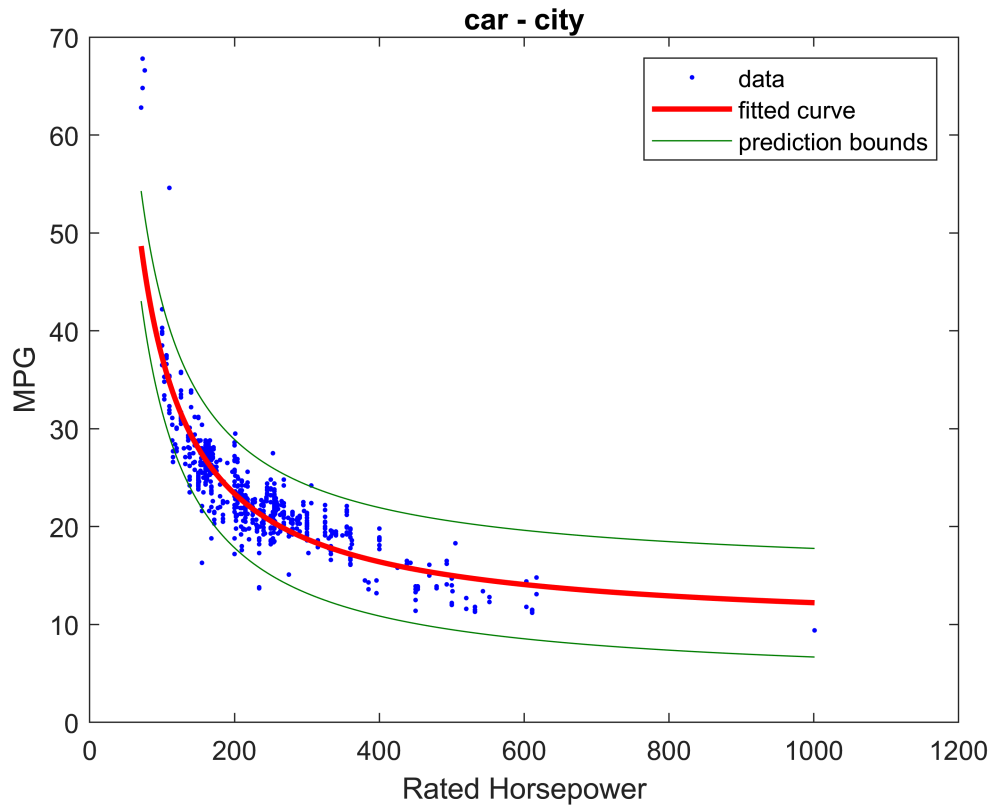


## Plot of Data and Model (for different groups)

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

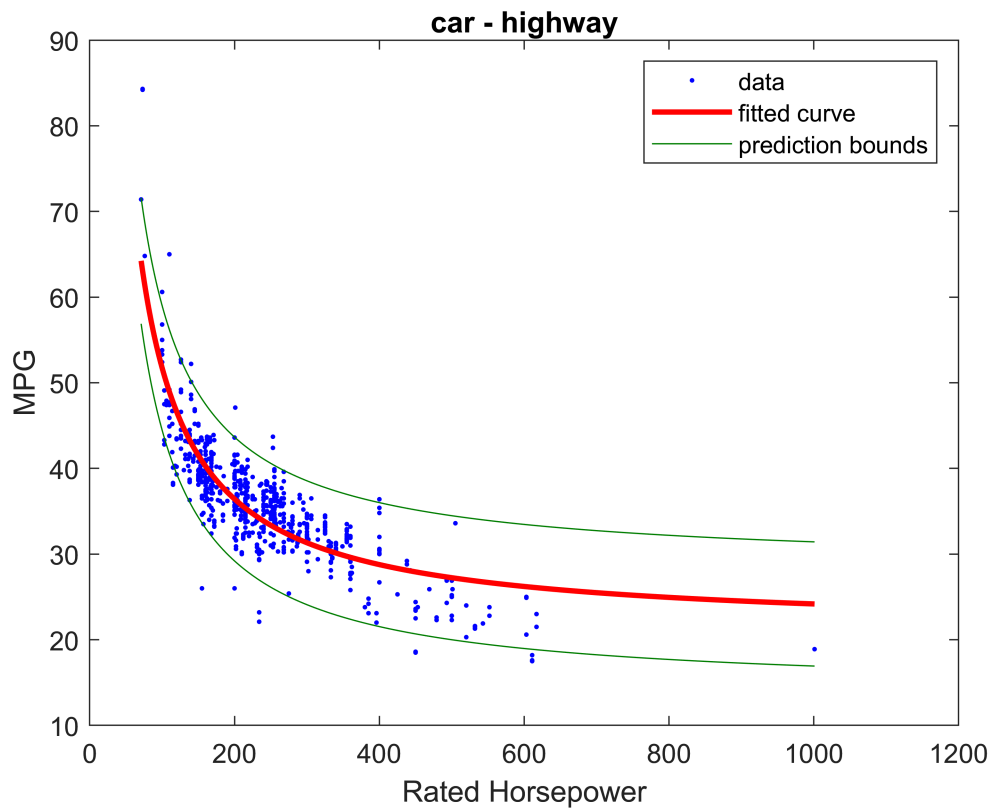
Model different combinations:

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



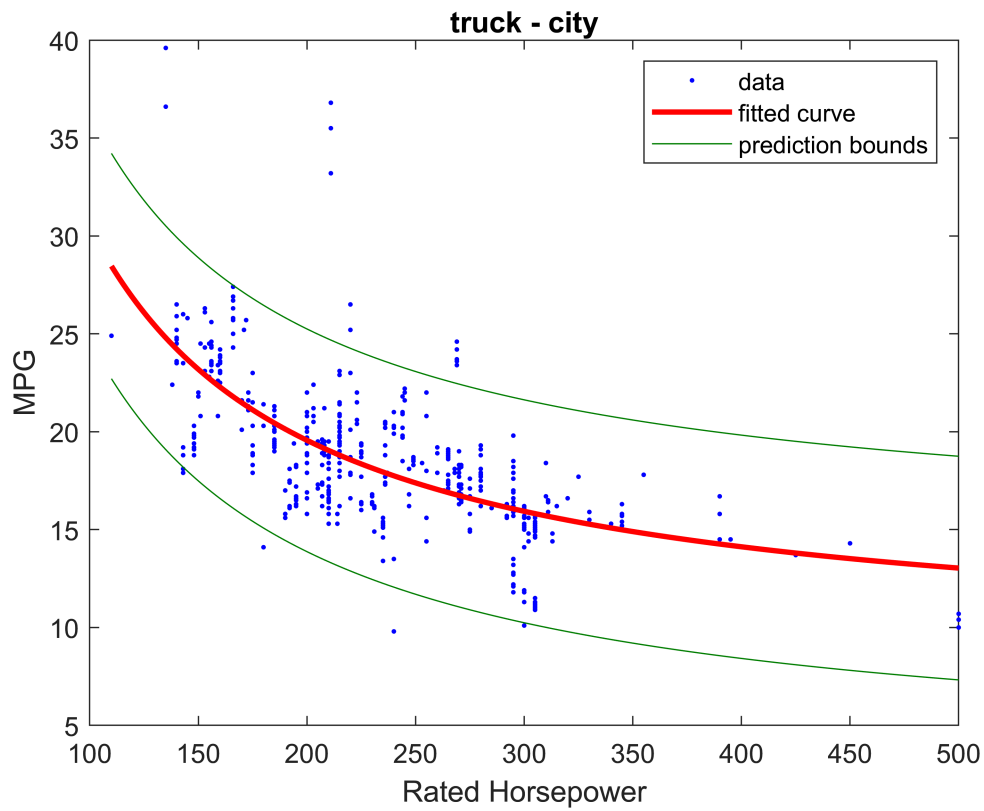
```
mdl =
  Linear model:
  mdl(x) = a + b*1/x
  Coefficients (with 95% confidence bounds):
    a =      9.443  (8.874, 10.01)
    b =     2784  (2675, 2894)
gof = struct with fields:
    sse: 5.7126e+03
    rsquare: 0.7763
    dfe: 721
    adjrsquare: 0.7760
    rmse: 2.8148
```

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



```
mdl =
  Linear model:
  mdl(x) = a + b*1/x
  Coefficients (with 95% confidence bounds):
    a =      21.12  (20.37, 21.86)
    b =     3061   (2918, 3205)
gof = struct with fields:
    sse: 9.6775e+03
    rsquare: 0.7108
    dfe: 715
    adjrsquare: 0.7104
    rmse: 3.6790
```

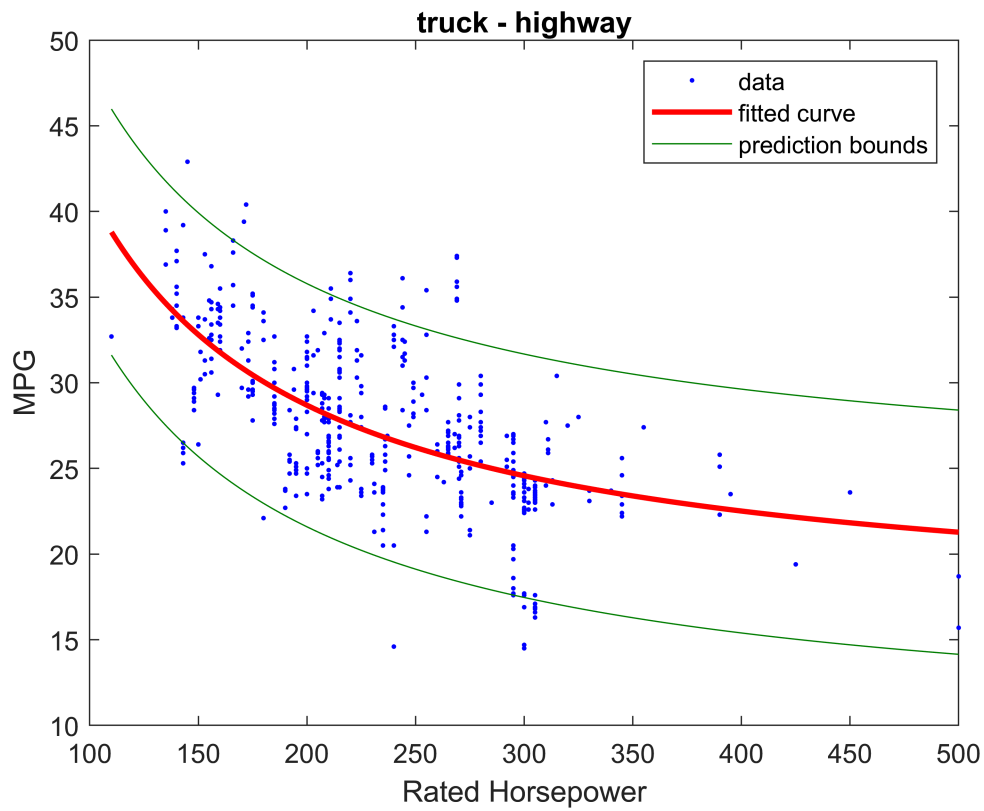
```
[mdl,gof] = modelMPG(carData, 'truck', 'city')
```



```
mdl =
  Linear model:
  mdl(x) = a + b*1/x
  Coefficients (with 95% confidence bounds):
    a =      8.685  (7.729, 9.641)
    b =     2175  (1977, 2373)
gof = struct with fields:
    sse: 4.4074e+03
    rsquare: 0.4689
    dfe: 527
    adjrsquare: 0.4679
    rmse: 2.8919
```

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





```
mdl =
  Linear model:
  mdl(x) = a + b*1/x
  Coefficients (with 95% confidence bounds):
    a =      16.34  (15.13, 17.54)
    b =     2471  (2221, 2721)
gof = struct with fields:
    sse: 6.8494e+03
    rsquare: 0.4176
    dfe: 526
    adjrsquare: 0.4165
    rmse: 3.6086
```

Copyright 2017 The MathWorks, Inc.