

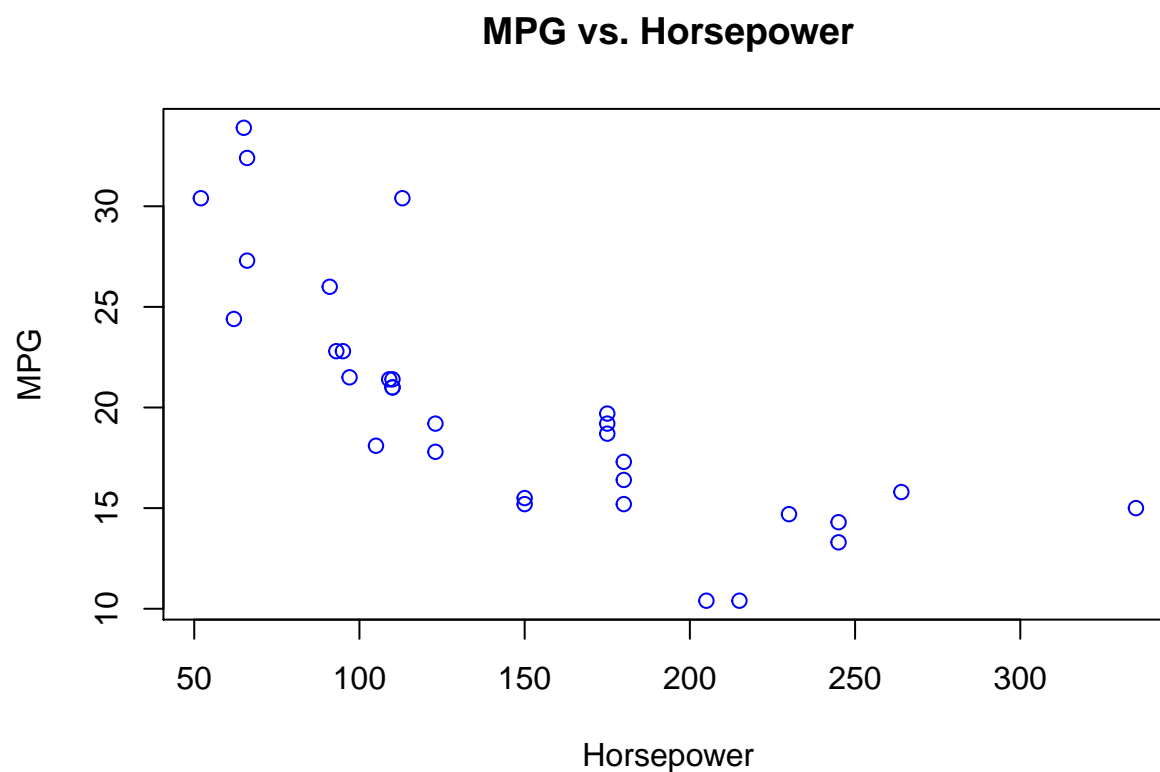
# Homework Sample

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## Problem 1

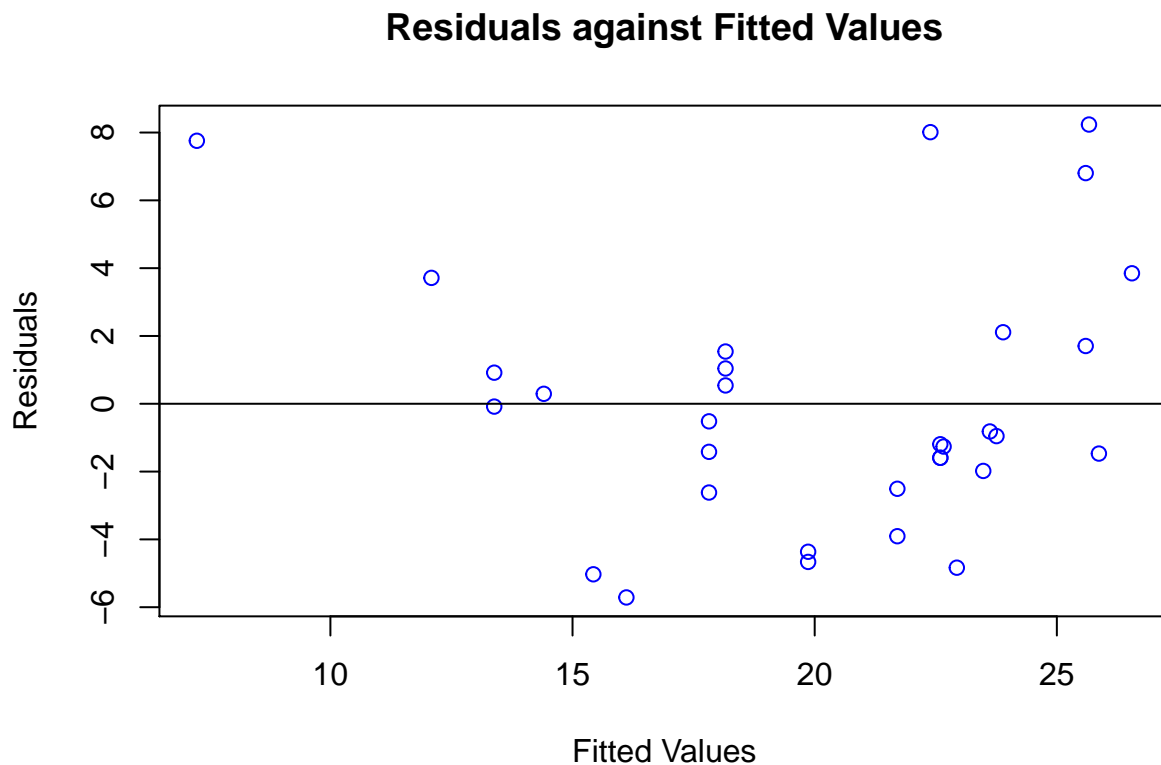
```
CAR <- mtcars
plot(y = CAR$mpg, x = CAR$hp, col = 4,
     main = "MPG vs. Horsepower",
     ylab = "MPG",
     xlab = "Horsepower")
```



There are outliers in the x-axis in *Horsepower* which will likely have large pull on the regression line. It is probably a good idea to remove outliers to increase accuracy in the shortened domain.

## Part (b)

```
model <- lm(mpg ~ hp, data = CAR)
plot(y = model$residuals, x = model$fitted.values, col = 4,
     main = "Residuals against Fitted Values",
     ylab = "Residuals",
     xlab = "Fitted Values")
abline(h = 0, col = 1)
```



Assumptions of mean-zero residuals, constant variance hold well in this plot. There is some shape to the error distribution. It curves up in the right tail end.

## Part (c)

The estimated regression coefficient  $\hat{\beta}_1 = -0.068$ .

## Part (d)

$R^2 = 0.602$ .

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### Part (e)

```
LOWCAR <- CAR[CAR$hp <= 250, ]  
newmodel <- lm(mpg ~ hp , data = LOWCAR)
```

### Part (f)

$\hat{\beta}_1 = -0.09$  for the new model, and  $\hat{\beta}_1 = -0.068$  for the model with outliers. The outliers had a strong upward pull on the slope of the line.

### Part (g)

The  $MSE$  is 10.86 as opposed to 13.99 previously. This is significant given deletion of only 2 of 32 rows.

### Part (h)

Removing the outliers in this case helped the data conform to a linear model. I would remove the outliers to better fit the model within (0, 250) horsepower, and avoid predicting outside of that interval.