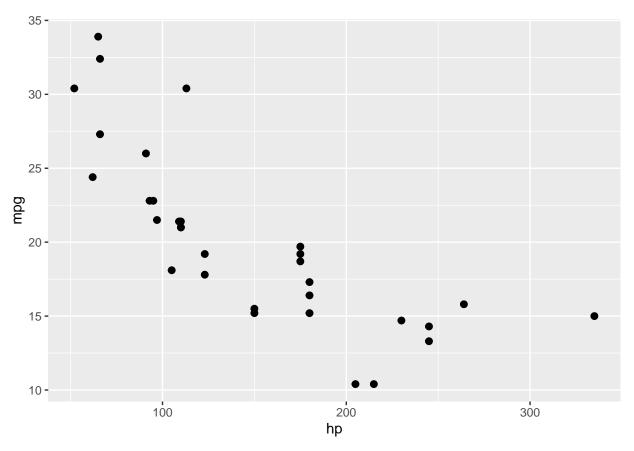
Lab 12

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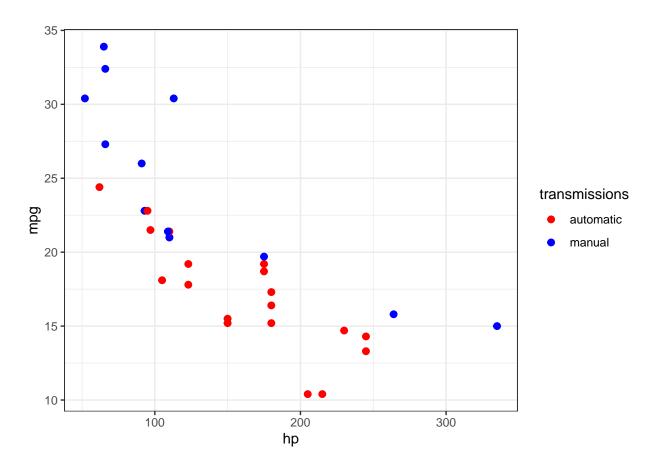
Install the data

```
mtcars %>%
 head() %>%
  select(mpg, hp, am)
                      mpg hp am
## Mazda RX4
                     21.0 110 1
## Mazda RX4 Wag
                     21.0 110 1
## Datsun 710
                     22.8 93 1
## Hornet 4 Drive
                     21.4 110 0
## Hornet Sportabout 18.7 175 0
## Valiant
                     18.1 105 0
We will focus on three variables: mpg, hp, am
mtcars %>%
  ggplot(aes(x = hp, y = mpg)) +
 geom_point(shape = 21, size = .1, stroke = 2)
```



We could also label the points based on the transmission type ${\tt am}$. ${\tt am}$ is a dummy variable:

- ullet 0 for automatic transmissions
- 1 for manual transmissions

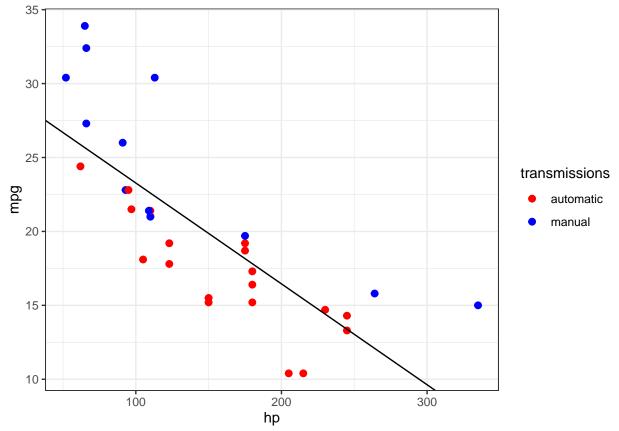


simple linear regression

$$Y = \beta 0 + \beta 1X1 + ei$$

```
mpg_hp_slr = lm(mpg ~ hp, data = mtcars)
summary(mpg_hp_slr)
##
## Call:
## lm(formula = mpg ~ hp, data = mtcars)
##
## Residuals:
##
      Min
               1Q Median
                               ЗQ
                                      Max
## -5.7121 -2.1122 -0.8854 1.5819 8.2360
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
                        1.63392 18.421 < 2e-16 ***
## (Intercept) 30.09886
                          0.01012 -6.742 1.79e-07 ***
## hp
              -0.06823
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
\#\# Residual standard error: 3.863 on 30 degrees of freedom
## Multiple R-squared: 0.6024, Adjusted R-squared: 0.5892
## F-statistic: 45.46 on 1 and 30 DF, p-value: 1.788e-07
```

Add the fitted line to the plot



Obviously, the red points are below the line, meaning that the model overestimates the fuel efficiency of automatic transmission. On the other hand, the blue points are above the line, meaning that the model underestimates the fuel efficiency of manual transmission. Thus, we need to constantly adjust the model.

Multiple regression model

$$Y = \beta 0 + \beta 1X1 + \beta 2X2 + ei$$

The model looks like

```
mpg_hp_add = lm(mpg ~ hp + am, data = mtcars)
summary(mpg_hp_add)
```

```
##
## Call:
```

```
## lm(formula = mpg ~ hp + am, data = mtcars)
##
## Residuals:
##
                                3Q
      Min
                1Q
                   Median
                                       Max
##
   -4.3843 -2.2642 0.1366
                           1.6968
                                   5.8657
##
## Coefficients:
##
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) 26.584914
                           1.425094
                                    18.655 < 2e-16 ***
## hp
              -0.058888
                           0.007857
                                    -7.495 2.92e-08 ***
## am
               5.277085
                           1.079541
                                      4.888 3.46e-05 ***
##
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.909 on 29 degrees of freedom
## Multiple R-squared: 0.782, Adjusted R-squared: 0.767
## F-statistic: 52.02 on 2 and 29 DF, p-value: 2.55e-10
```

As we mentioned, X2(am) is a dummy variable, it's only takes the values 0 and 1. We can write two separate versions, one for manual transmissions and the other for automatic transmissions.

For automatic transmissions - X2 = 0

$$Y = \beta 0 + \beta 1X1 + ei$$

For manual transmissions - X2 = 1

$$Y = (\beta 0 + \beta 2) + \beta 1X1 + ei$$

These models have the same slope $\beta 1$, but different intercepts, which differ by $\beta 2$. So the change in mpg is the same for both models, but the average mpg difference between the two transmission types is $\beta 2$.

```
summary(mpg_hp_add)$coefficients[1] # b0
## [1] 26.58491
summary(mpg_hp_add)$coefficients[2] # b1
## [1] -0.0588878
summary(mpg_hp_add)$coefficients[3] # b2
## [1] 5.277085
The estimated slope and intercepts can then be calculated by combining these.
intercept_auto = coef(mpg_hp_add)[1]
intercept_auto
## (Intercept)
##
      26.58491
intercept_manu = coef(mpg_hp_add)[1] + coef(mpg_hp_add)[3]
intercept_manu
## (Intercept)
##
        31.862
```

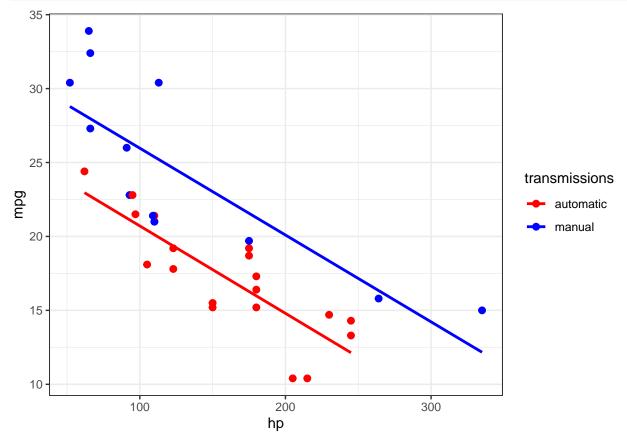
```
slope_auto = coef(mpg_hp_add)[2]
slope_auto

##     hp
## -0.0588878

slope_manu = coef(mpg_hp_add)[2]
slope_manu

##     hp
## -0.0588878
```

Re-Plot



The above plot makes it abundantly clear that $\beta 2$ is significant, but let us test it mathematically.

• Hypothesis Test: $\beta 2 = 0$ vs $\beta 2 \neq 0$

t-test

```
summary(mpg_hp_add)$coefficients["am", ]

## Estimate Std. Error t value Pr(>|t|)
## 5.277085e+00 1.079541e+00 4.888270e+00 3.460318e-05
```

F test

```
anova(mpg_hp_slr, mpg_hp_add)

## Analysis of Variance Table

##

## Model 1: mpg ~ hp

## Model 2: mpg ~ hp + am

## Res.Df RSS Df Sum of Sq F Pr(>F)

## 1 30 447.67

## 2 29 245.44 1 202.24 23.895 3.46e-05 ***

## ---

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

According to the t-test and f-test, the p-values are the same, but the F test statistic is the t test statistic squared.

Interpretations

- bo = 26.584914 is the estimated average mpg for an automatic transmission car with 0 hp.
- b0+b2 = 26.584914 + 5.277085 = 31.862 is the estimated average mpg for a manual transmission car with 0 hp.
- b2 = 5.277085 is the estimated difference in average mpg for cars with manual transmissions as compared to those with automatic transmission, for any hp.
- b1 = -0.058888 is the estimated change in average mpg corresponds to increase in one unit of hp, no matter manual or automatic transmissions.