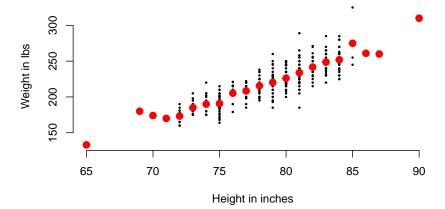
Business Statistics 41000 NBA height and weigth

Mladen Kolar

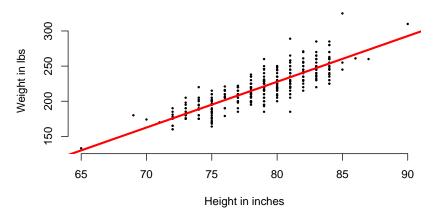
NBA height and weight: $E(Y \mid X = x)$



A few heights have only one observation. Is that problematic?

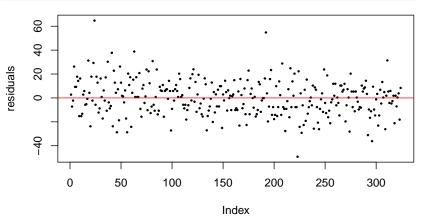
The least square fit line

$$\widetilde{\text{weight}} = -293.33 + 6.513 \cdot \text{height}$$



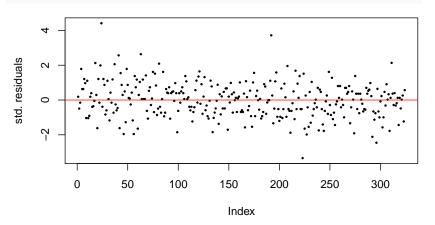
We can plot residuals

```
plot(fit$residuals, pch=20, cex=0.5, ylab="residuals")
abline(h=0, col="red")
```



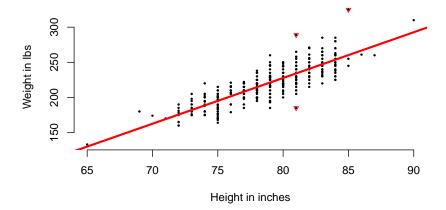
... or standardized residuals

plot(rstandard(fit), pch=20, cex=0.5, ylab="std. residuals")
abline(h=0, col="red")



```
outlier = abs(rstandard(fit)) > 3
nba[outlier, ]
```

```
##
                   name weight height team
## 24
       oneal, shaquille
                            325
                                    85
                                         pho
## 192
            davis, glen
                            289
                                     81
                                         bos
## 223
          brewer, corey
                            185
                                     81
                                         min
```



summary(fit)

```
##
## Call:
## lm(formula = weight ~ height, data = nba)
##
## Residuals:
##
     Min 1Q Median 3Q Max
## -49.21 -9.70 0.33 9.26 64.74
##
## Coefficients:
##
             Estimate Std. Error t value Pr(>|t|)
## (Intercept) -293.329 17.825 -16.5 <2e-16 ***
## height
          6.513 0.226 28.9 <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 14.8 on 322 degrees of freedom
## Multiple R-squared: 0.721, Adjusted R-squared: 0.72
## F-statistic: 833 on 1 and 322 DF, p-value: <2e-16
```

The anova() command, which stands for "analysis of variance".

```
anova(fit)
```

Let's check that the sum-of-squares regression divided by the sum-of-squares total equals \mathbb{R}^2 from the previous slide.

Forecasting interval

```
new_height = data.frame(height=c(75, 81,85))
predict.lm(fit, newdata=new_height, interval="pred", level=0.95)
##
        fit
               lwr
                      upr
## 1 195.13 166.00 224.27
## 2 234.21 205.11 263.31
## 3 260.26 231.05 289.47
predict.lm(fit, newdata=new_height, interval="pred", level=0.90)
##
        fit
               lwr
                      upr
## 1 195.13 170.70 219.56
## 2 234.21 209.81 258.61
## 3 260.26 235.77 284.76
```