Linear Models

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STATS 780/CSE 780

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Introduction

- Before going any further, please note that there are entire courses devoted to linear models.
- Today I want to give you a flavour there is a limit to what I can cover in one "lecture".
- \bullet I recommend Faraway (2014) $^{\rm a}$ for further reading.

^aFaraway, J.J. (2014). Linear Models with R. 2nd edn. Boca Raton: Chapman & Hall/CRC Press.

Basic Model Description

- Consider $(\mathbf{X}_1, Y_1), (\mathbf{X}_2, Y_2), \dots, (\mathbf{X}_n, Y_n).$
- ullet In simple terms, we can think of a linear model with J predictors as

$$Y_i = \beta_0 + \beta_1 X_{i1} + \beta_2 X_{i2} + \dots + \beta_J X_{iJ} + \varepsilon_i = \beta_0 + \sum_{j=1}^J \beta_j X_{ij} + \varepsilon_i,$$

for $i=1,\ldots,n$, where the $\beta_j\in\mathbb{R}$ are (unknown) parameters and $\varepsilon_i\stackrel{\text{i.i.d.}}{\sim} \mathsf{N}(0,\sigma^2).$

• Note that the predictors can be transformed, giving tremendous flexibility.

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Recall: Fathers & Sons

- Height for 1,078 fathers and sons (in inches).
- Very famous example used by Pearson.
- We looked at some pictures, and the first one is just a (simple) linear model:

$$Y_i = \beta_0 + \beta_1 X_i + \varepsilon_i$$

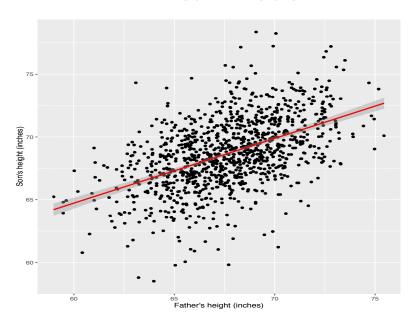
where Y_i is the height of the *i*th son and X_i is the height of the *i*th father.

The plotted (red) line is defined by

$$\hat{Y} = \hat{\beta_0} + \hat{\beta_1} X.$$

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A Linear Model



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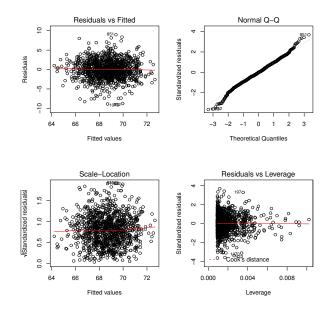
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Output from R

```
>data(father.son, package="UsingR")
>model<-lm(sheight~fheight, father.son)</pre>
>summary(model)
lm(formula = sheight ~ fheight, data = father.son)
Residuals:
            1Q Median
-8.8772 -1.5144 -0.0079 1.6285 8.9685
Coefficients:
           Estimate Std. Error t value Pr(>|t|)
                       1.83235 18.49
(Intercept) 33.88660
                                         <2e-16 ***
fheight
            0.51409
                       0.02705 19.01
                                         <2e-16 ***
Signif. codes: 0 *** 0.001 ** 0.01 * 0.05 . 0.1 1
Residual standard error: 2.437 on 1076 degrees of freedom
Multiple R-squared: 0.2513, Adjusted R-squared: 0.2506
F-statistic: 361.2 on 1 and 1076 DF, \, p-value: < 2.2e-16
```

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Residual Plots



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Comments

- What can we actually learn from this model?
- Now, we will turn to R and look at some data sets.
- Learning in this way is, I think, effective in this environment.
- However, I strongly recommend that you do some background reading.
- I suggest using Faraway (2014) or the first edition thereof.

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