

PS2 → grading
Quiz 2 →
PS3 → posted

Lecture 11

Lab 6
Lab 7 and 8

Finish Model Checking

Implementation of WLS and robust variance estimation in R

Key Assumptions by Order of Importance

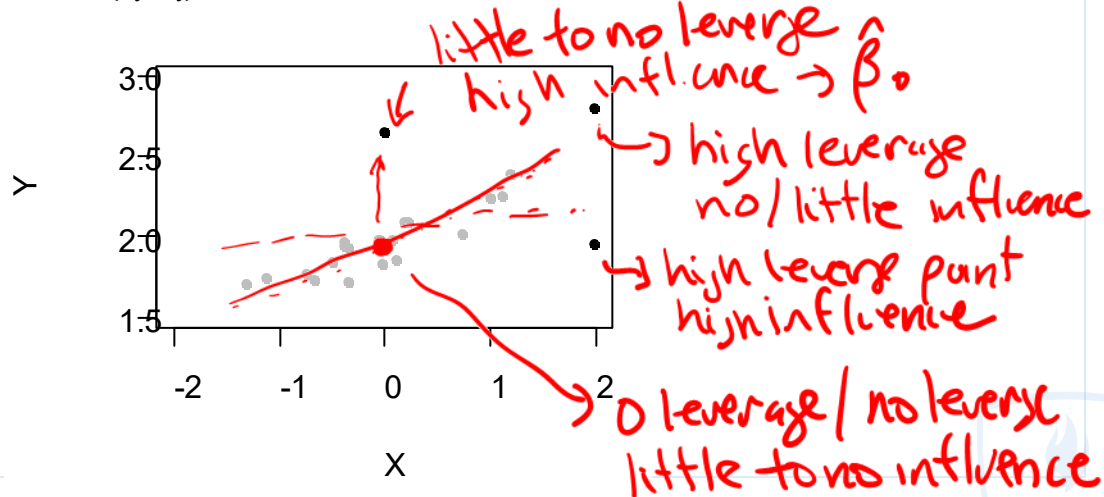
- ▶ $E(Y|X) = X\beta$, i.e. the mean model is “correctly” specified
 - ▶ Misspecification of $X\beta$ can lead to biased β / misinterpretations
 - ▶ Omitted variable Bias
 - ▶ Correct functional form for continuous X
 - ▶ Residuals are independent
 - ▶ This assumption is violated due to the design of the study
 - ▶ Longitudinal study
 - ▶ Clustered design
 - ▶ Show today: ignoring the correlation will impact $Var(\hat{\beta})$ and derive weighted least squares
 - ▶ Variance of residuals is constant
 - ▶ Often the variance is a function of some X
 - ▶ Show today: same impact and solution as violation of independence
 - ▶ Residuals are normally distributed
 - ▶ CLT bootstrap procedure
 - ▶ There are not a small number of highly influential observations
 - ▶ Sensitivity analyses
- WLS
- WLS
- correlated data \Rightarrow resample level 1 units

Leverage and Influence

- ▶ Leverage: A measure of how far an individual's predictors (X_i) are from the mean X

- ▶ Hat matrix:
$$h_{ii} = \frac{(X_i - \bar{X})^2}{\sum (X_i - \bar{X})^2}$$

- ▶ Influence: An observation (Y_i, X_i) such that including this value would greatly change the fitted values: $\hat{\beta}$ and \hat{Y} .



Influence statistics

There are several influence statistics that are used in practice:

$j = 0, 1, \dots, p$

$$\bullet \text{ } DBETA_{ij} = \hat{\beta}_j - \hat{\beta}_{j(-i)}$$

$i = 1, \dots, n$

$$\bullet \text{ } DBETAS_{ij} = \frac{DBETA_{ij}}{\hat{se}(\hat{\beta}_{j(-i)})}$$

$$\bullet \text{ } DFIT_i = \hat{Y}_i - \hat{Y}_{i(-i)}$$

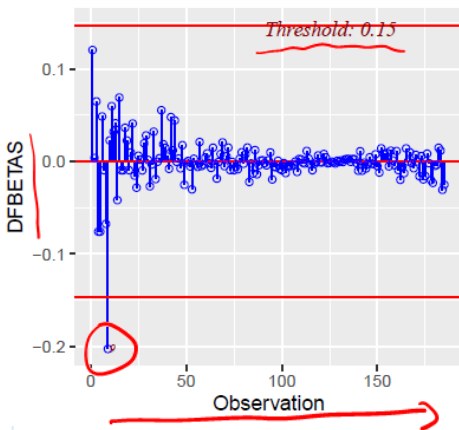
$$\bullet \text{ } DFITS_i = \frac{DFIT_i}{\hat{se}(\hat{Y}_{i(-i)})}$$

\Rightarrow defined for each parameter in the model, SLR
DBeta β_0, β_1
 $j = 0, 1$

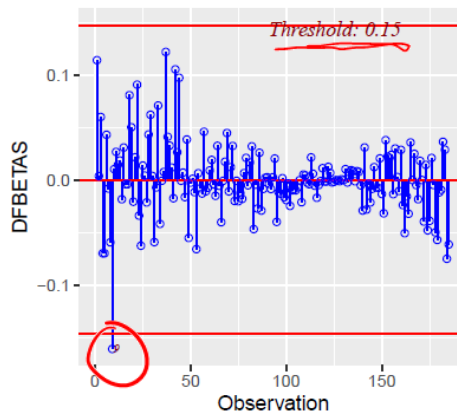


Example: Nepali Anthropometry Data

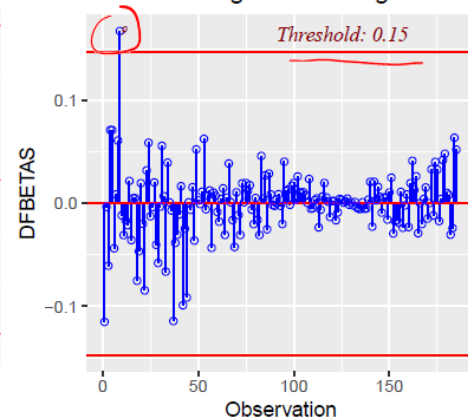
Influence Diagnostics for (Intercept)



Influence Diagnostics for agesp6



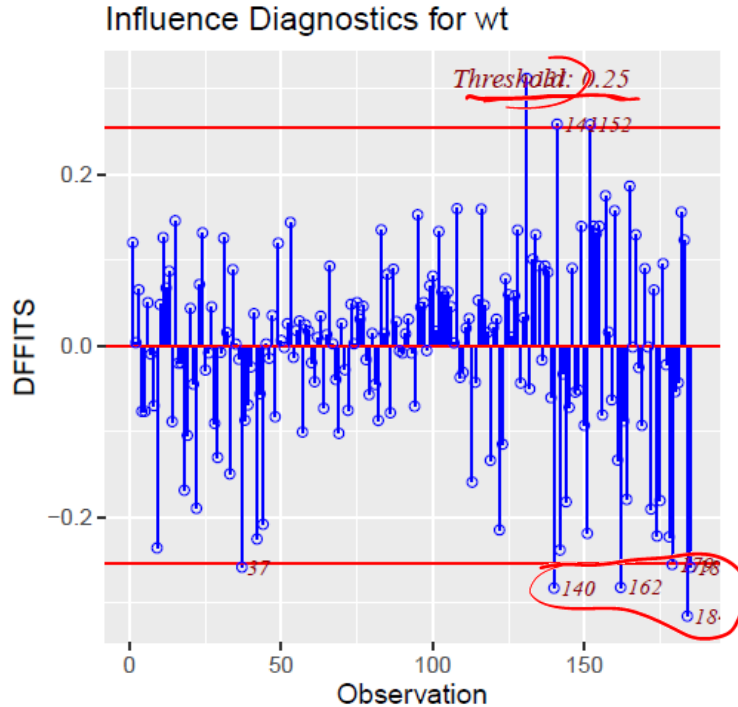
Influence Diagnostics for age



$$WT = \beta_0 + \beta_1 \text{age} + \beta_2 (\text{age} - 6)^+ + \epsilon$$

Obs 9

Example: Nepali Anthropometry Data



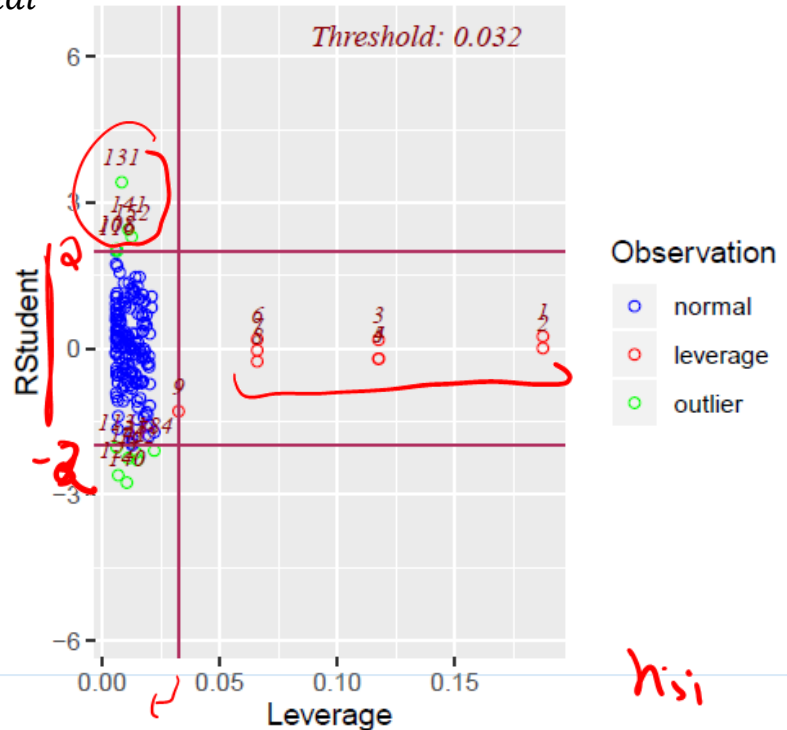
Example: Nepali Anthropometry Data

► *RStudent*: studentized residual

$$\frac{y_i - \hat{y}_{(i)}}{se(y_i - \hat{y}_{(i)})} = \frac{y_i - \hat{y}_i}{\sqrt{MSE_{(i)}(1 - h_{ii})}}$$

mean 0
sd 1

Outlier and Leverage Diagnostics for wt



Implementation of WLS in R

- ▶ For the remainder of the lecture, we will work through some analyses to demonstrate how to fit WLS in R
- ▶ In addition, I will show one approach for obtaining robust variance estimates for different working correlation models. Here we will use the *gee* package in R.
- ▶ See Handout 11.

