

36-463/663: Multilevel & Hierarchical Models

Spring 2022

HW03 – Due Tue Feb 15, 11:59pm

- Please turn the homework in, as a single pdf, online in GradeScope using the link provided on the Assignment page on canvas.cmu.edu. Upload *one* file per person.
- Reading:
 - See the subfolder “reading” in the week05 folder on Canvas.
- There are two main exercises, each with “parts”...

Exercises

1. Return to the CD4 data from HW02, and consider the model

```
lmer.1 <- lmer(sqrt.CD4PCT ~ 1 + VISIT * treatmnt + (1 + VISIT | newpid),  
              data=cd4)
```

Bring together all the residuals (except random effects residuals) that we have talked about, using library(HLMdiag):

```
> r.1 <- hlm_resid(lmer.1, level=1, include.ls=F)  
> r.1s <- hlm_resid(lmer.1, level=1, include.ls=F, standardize=T)  
> r.2 <- hlm_resid(lmer.1, level="newpid", include.ls=F)  
> r.2s <- hlm_resid(lmer.1, level="newpid", include.ls=F, standardize=T)  
> names(r.1); names(r.1s); names(r.2); names(r.2s)
```

Read the help file for `hlm_resid` if necessary to understand what all the components are. (You can create all these residuals “by hand” as I have shown in lecture slides and in the accompanying R files, but using HLMdiag is simpler and more reliable.)

- (a) Make a facets plot of the marginal residuals, as a function of the marginal fitted values (use `scales="free_x"` if needed to make the plot legible). Explain in a sentence or two why a facets plot is not very useful for assessing model fit for this problem, whether we look at the first 12 children, or all 251 children).
- (b) Make an ungrouped (that is, no facets) scatter plot of marginal residuals as a function of marginal fitted values, using the full data set (not just the first 12 children). Color the points for `treatmnt=1` kids and `treatmnt=2` kids with different colors. Overlay a smooth fit (`geom_smooth` is the easiest to use here). Explain, in a couple of sentences (optionally with some math):
 - What is causing the dominant structure in this plot, and why that dominant structure is essentially irrelevant for checking the relationship between `sqrt.CD4PCT` and `VISIT`; and
 - What in this plot makes you happy or unhappy about having a linear relationship between `sqrt.CD4PCT` and `VISIT` in the model.
- (c) Make an ungrouped (that is, no facets) scatter plot of conditional residuals as a function of conditional fitted values, using the full data set (not just the first 12 children). Color the points for `treatmnt=1` kids and `treatmnt=2` kids with different colors. Overlay a smooth fit (`geom_smooth` is the easiest to use here). Explain, in a couple of sentences (optionally with some math):

- Why the dominant structure in the marginal residuals is not also present in this plot of conditional residuals
 - What might be causing the trend you see in this plot to be different from the trend in the plot of the marginal residuals.
- (d) Use standardized residuals and standardized random effects estimates to assess the normality of ϵ_i , η_{0j} and η_{1j} in the fitted model, and to check for any outliers. Include qq plots for each, and accompany each plot with a sentence or two describing what is good or bad in that plot.

2. Continuing with the CD4 data...

- (a) Make a table giving values of AIC, BIC, DIC, and cAIC (you compared two of these on the last assignment, using just AIC, BIC and DIC):

```
sqrt.CD4PCT ~ 1 + visage + (1+visage|newpid)
sqrt.CD4PCT ~ 1 + visage + treatmnt + (1+visage|newpid)
sqrt.CD4PCT ~ 1 + visage * treatmnt + (1+visage|newpid)
sqrt.CD4PCT ~ 1 + VISIT + (1+VISIT|newpid)
sqrt.CD4PCT ~ 1 + VISIT + treatmnt + (1+VISIT|newpid)
sqrt.CD4PCT ~ 1 + VISIT * treatmnt + (1+VISIT|newpid)
```

Comment briefly on any similarities or differences in how the different criteria choose fixed effects.

- (b) Now let's look at the random effects in the model

```
sqrt.CD4PCT ~ 1 + VISIT + treatmnt + (1+VISIT|newpid)
```

Again, make a table giving values of AIC, BIC, DIC, and cAIC for the following models:

```
sqrt.CD4PCT ~ 1 + VISIT + treatmnt
sqrt.CD4PCT ~ 1 + VISIT + treatmnt + (1|newpid)
sqrt.CD4PCT ~ 1 + VISIT + treatmnt + (0+VISIT|newpid)
sqrt.CD4PCT ~ 1 + VISIT + treatmnt + (1+VISIT|newpid)
```

You'll fit the first model with `lm()`, and the others with `lmer()`. Comment briefly on any similarities or differences in how the different criteria choose random effects. (Note that only BIC and cAIC have a strong theoretical justification here).

- (c) Repeat part (b) but with the interaction `VISIT * treatmnt` in each model instead of just the main effects `VISIT + treatmnt`.