

# Chapter 5

## Modeling and inference using GLMMs

### 5.1 Introduction

In this chapter I continue the prescription of Section 4.4 and present a number of examples and consider the inferential goals.

### 5.2 Chestnut blight (gene effects)

Recall the model we developed in the first chapter (1.1) for the chestnut blight example, now modified to include random effects:

$$(5.1) \quad \begin{aligned} Y_i &= 1 && \text{if the virus is transmitted and 0 otherwise,} \\ Y_i | \mathbf{u} &\sim \text{indep. Bernoulli}(p_i), \\ p_i &= \Phi \left( \mu + \sum_s \beta_s \text{MCH}_{is} + \sum_s \gamma_s \text{ASY}_{is} + \mathbf{z}'_{d,i} \mathbf{u}_1 + \mathbf{z}'_{r,i} \mathbf{u}_2 \right), \end{aligned}$$

where  $\mathbf{z}'_{d,i}$  and  $\mathbf{z}'_{r,i}$  are the  $i$ th rows of the model matrices for the donor and recipient random effects, respectively, and we assume

$$(5.2) \quad \begin{aligned} \mathbf{u}_1 &\sim \mathcal{N}(0, \mathbf{I}\sigma_d^2) \text{ independent of} \\ \mathbf{u}_2 &\sim \mathcal{N}(0, \mathbf{I}\sigma_r^2). \end{aligned}$$

One inferential goal might be to test if gene 4 had an effect. To do so we could fit the model described by (5.1) and (5.2) and evaluate the log likelihood. We would next fit the same model, but with  $\beta_4$  and  $\gamma_4$  set equal to zero and compare the value of the log likelihood. A large sample likelihood ratio test could be used to test  $H_0 : \beta_4 = \gamma_4 = 0$ . The inferential goal in this case is to form a hypothesis test of parameters from the linear predictor.