## Use of Generalized Inverse Matrices for One-Way ANOVA

For most of the course we will assume that the  $\mathbf{X}'\mathbf{X}$  matrix is nonsingular, so the least-squares solution is unique. Here we examine an approach that can be used when the  $\mathbf{X}'\mathbf{X}$  matrix is singular. Suppose that we have data from two groups, with sample sizes  $n_1$  and  $n_2$ , respectively, with total sample size  $n = n_1 + n_2$ . We use the overparametrized model,

$$Y_{ij} = \mu + \alpha_j + \epsilon_{ij},$$

so that the model matrix  $\mathbf{X}$  has the following form:

$$\mathbf{X} = \begin{bmatrix} 1 & 1 & 0 \\ 1 & 1 & 0 \\ 1 & 1 & 0 \\ \vdots & \vdots & \vdots \\ 1 & 1 & 0 \\ 1 & 0 & 1 \\ 1 & 0 & 1 \\ \vdots & \vdots & \vdots \\ 1 & 0 & 1 \end{bmatrix},$$

and the  $\mathbf{X}'\mathbf{X}$  matrix is:

$$\mathbf{X'X} = \begin{bmatrix} n & n_1 & n_2 \\ n_1 & n_1 & 0 \\ n_2 & 0 & n_2 \end{bmatrix},$$

which only has rank 2.

A generalized inverse of an m x n matrix **B** is defined to be any n x m matrix **B**<sup>-</sup> that satisfies

## $BB^{-}B=B.$

If the model

$$\mathbf{y} = \mathbf{X}\beta + \epsilon$$

has an  $\mathbf{X}'\mathbf{X}$  matrix that is singular then a generalized inverse can be used to obtain a least squares solution:

$$\mathbf{b} = (\mathbf{X}'\mathbf{X})^{-}\mathbf{X}'\mathbf{y}.$$

This solution, however, is not unique. We will discuss one way to obtain a generalized inverse and use it for the problem above.