

Homework1

BIOS6643 Fall 2021

8/20/2021

Question 1 PCA

Consider the eNO data, and how we applied PCA to the data for graphical purposes (see Graphs slides). Determine the slope of the regression of Post (Y_2) on Pre (Y_1) values (i.e., a standard ‘baseline as covariate’ model), and compare this to the ‘slope’ of the $PC1$ axis. Compare the slopes numerically and superimpose the lines on a scatterplot of Post versus Pre values.

In order to do this, recall $PC1 = aY_1 + bY_2$, where a and b are chosen to maximize the variance of $PC1$ (recall $a = 0.51$, $b = 0.86$ for the data; see the slides).

Note: in terms of Y_2 versus Y_1 , the ‘slope’ of the $PC1$ axis is simply b/a ; to create a line to graph for $PC1$, you can have it go through the joint sample mean of Y_1 and Y_2 . This exercise helps demonstrate the ‘regression’ principle in a regression line.

Question 2 GLM, GzLM, and LMM

In a paragraph, explain the difference between a general linear model (GLM; not a generalized linear model, which I denote with GzLM and which will be discussed more later) and a linear mixed model (LMM).

Question 3 Profiled likelihood, restricted likelihood, and Likelihood functions

In a short paragraph, explain the difference between a profiled likelihood and a restricted likelihood for a linear mixed model, and how and why they are used. Which one is a re-expression of the standard likelihood?

Question 4 Variance in LMM

Derive $Var[\hat{\beta}]$ in a full-rank linear mixed model, given the algebraic form of $\hat{\beta}$ that is obtained via ML estimation.

NOTE: there are two types of variance, model-based and empirical (or sandwich estimator). The difference is whether the middle \mathbf{V} is determined via the model or using squared residual quantities; derive *the model-based form*. To answer this question, work with the ‘complete data’ form of $\hat{\beta}$.