## Preliminary Examination: LINEAR MODELS

Answer all questions and show all work. Q 1 is 30 points; Q 2 is 35 points, and Q3 is 35 points.

1. Define

$$
\mathbf{Y}=\left(\begin{array}{c}
Y_{1} \\
Y_{2} \\
Y_{3} \\
Y_{4} \\
Y_{5} \\
Y_{6}
\end{array}\right), \mathbf{X}=\left(\begin{array}{rrrrr}
1 & -1 & 1 & 0 & 0 \\
1 & -1 & 0 & 1 & 0 \\
1 & 0 & 1 & 0 & -1 \\
1 & 1 & 0 & -1 & 0 \\
1 & 1 & 0 & 0 & -1 \\
1 & 1 & -1 & 0 & 0
\end{array}\right), \text { and } \boldsymbol{\beta}=\left(\begin{array}{c}
\mu \\
\theta_{1} \\
\theta_{2} \\
\theta_{3} \\
\theta_{4}
\end{array}\right)
$$

Assume the linear model $\mathbf{Y}=\mathbf{X} \boldsymbol{\beta}+\boldsymbol{\varepsilon}$, where $E(\varepsilon)=\mathbf{0}$, and $\operatorname{var}(\varepsilon)=\sigma^{2} \mathbf{I}$. Let $\mathbf{x}_{0}, \mathbf{x}_{1}$, $\mathbf{x}_{2}, \mathbf{x}_{3}$, and $\mathbf{x}_{4}$ denote the columns of $\mathbf{X}$. Note that $\mathbf{x}_{1}+\mathbf{x}_{2}+\mathbf{x}_{3}+\mathbf{x}_{4}=\mathbf{0}$.
a. Find $E(\mathbf{Y})$ in terms of $\mu, \theta_{1}, \theta_{2}, \theta_{3}$, and $\theta_{4}$.
b. Let $\boldsymbol{\lambda}=\left(\lambda_{0}, \lambda_{1}, \lambda_{2}, \lambda_{3}, \lambda_{4}\right)^{\prime}$. Give the necessary and sufficient condition(s) for $\boldsymbol{\lambda}^{\prime} \boldsymbol{\beta}$ to be estimable in the form of $\boldsymbol{\lambda}^{\prime} \mathbf{c}_{i}=0, i=1, \ldots, s$. What is the value of $s$ ?
c. Is $\mu-\theta_{1}+\theta_{2}+\theta_{3}-\theta_{4}$ estimable? Justify your answer.
2. Suppose that data $\left\{\left(x_{i j}, y_{i j}\right): i=1, \ldots, k, j=1, \ldots, J\right\}$ can be modeled as having a common slope and possibly different intercepts using the linear model,

$$
Y_{i j}=\beta_{i}+\gamma x_{i j}+\epsilon_{i j}
$$

where $\left\{\epsilon_{i j}\right\}$ are independently and identically distributed $N\left(0, \sigma^{2}\right)$ random variables. Assume that no vector $\left(x_{i 1}, \ldots, x_{i J}\right)$, for $i=1, \ldots, k$, is proportional to the vector of 1 s .
a. Determine the ordinary-least-squares estimator of $\left(\beta_{1}, \ldots, \beta_{k}, \gamma\right)^{\prime}$.
b. Give an explicit expression for the size $\alpha$ likelihood-ratio test of the hypothesis,

$$
H_{0}: \beta_{1}=\cdots=\beta_{k}=0 \text { versus } H_{a}: \operatorname{not} H_{0}
$$

c. Compute the power of the test that you derived in part (b). Is the power dependent on $\gamma$ ?
3. Consider the following change-point model:

$$
Y_{i}=\mu_{i}+e_{i}, i=1, \ldots, n
$$

where

$$
\mu_{i}= \begin{cases}\beta_{1}, & \text { if } i \leq n / 2 \\ \beta_{2}, & \text { otherwise }\end{cases}
$$

Suppose we observe $Y_{1}, \ldots, Y_{n}$. For simplicity, assume that the same size if even, namely $n=2 m$ for some integer $m>0$.
a. Find the (ordinary) least squares estimator $\left(\hat{\beta}_{1}, \hat{\beta}_{2}\right)^{\prime}$ for $\left(\beta_{1}, \beta_{2}\right)^{\prime}$.

Assume that the errors satisfy:

$$
e_{i}=\epsilon_{i}-a \epsilon_{i-1}, i=1, \ldots, n
$$

where $a \in \mathbb{R}$ is a parameter controlling the dependence strength, and $\epsilon_{k}, k=0,1, \ldots, n$, are independent normal random variables with mean zero and variance $\sigma^{2}>0$.
b. For parts (b) and (c) only, assume that $a=0$. Find the joint distribution of $\left(\hat{\beta}_{1}, \hat{\beta}_{2}\right)^{\prime}$. Are $\hat{\beta}_{1}$ and $\hat{\beta}_{2}$ independent in this case? How does the variance of $\hat{\beta}_{1}$ change when $n \rightarrow \infty$ (for example whether it decreases to zero linearly in $n$, quadratically or at some other rate)?
c. Following part (b), find an unbiased estimator $\hat{\sigma}^{2}$ of $\sigma^{2}$ and devise a statistically valid test for the following hypotheses:

$$
H_{0}: \beta_{1}=\beta_{2} \text { vs } H_{a}: \operatorname{Not} H_{0}
$$

You need to specify the test statistic and its distribution under the null hypothesis.
d. Now suppose that $a=1$, find the joint distribution of $\left(\hat{\beta}_{1}, \hat{\beta}_{2}\right)^{\prime}$. Are $\hat{\beta}_{1}$ and $\hat{\beta}_{2}$ independent in this case? How does the variance of $\hat{\beta}_{1}$ change when $n \rightarrow \infty$ in this case (for example whether it decreases to zero linearly in $n$, quadratically or at some other rate)? Compare your result with the one in part (b) and comment on the effect of dependence among the errors. Is dependence always a "bad" thing?
e. Now suppose that $0<a<1$. Find the distribution of $\hat{\beta}_{1}$. How does the variance of $\hat{\beta}_{1}$ change when $n \rightarrow \infty$ (for example whether it decreases to zero linearly in $n$, quadratically or at some other rate)? Compare your result with the ones in parts (b) and (d) and comment.

