

1. 15 points

Note:

- Linearity does not necessarily imply identifiability, i.e. each parameter has a unique solution. See part (c) as an example
- "Linear model" means linear in terms of parameters β , but not necessarily linear in explanatory variables X_i

- (a) Linear model.
- (b) Linear model.
- (c) Linear model, since it is equivalent to $Y_i - \log(X_{1i}) = (\beta_0 + \log(\beta_1)) + \beta_2 X_{2i} + \epsilon_i$. In this scenario, neither β_0 nor β_1 are identifiable.
- (d) Nonlinear model.
- (e) Intrinsically linear model. $\log((1-x)/x)$ transformation to both side. (This is the logistic regression)
- (f) Nonlinear model.
- (g) Intrinsically linear model. Log transformations of the response variable and random error are needed.

2. 5 points

(Properties of P_X : see notes page 872)

With $P_X = X(X^T X)^{-1} X^T$, $P_1 = \mathbf{1}(\mathbf{1}^T \mathbf{1})^{-1} \mathbf{1}^T$, and $P_X \mathbf{u} = \mathbf{u}$ for any vector \mathbf{u} in the space spanned by the columns of X ,

$$\begin{aligned}
 (P_X - P_1)(P_X - P_1) &= P_X P_X - P_X P_1 - P_1 P_X + P_1 P_1 \\
 &= X(X^T X)^{-1} X^T X(X^T X)^{-1} X^T - P_X \mathbf{1}(\mathbf{1}^T \mathbf{1})^{-1} \mathbf{1}^T - \\
 &\quad \mathbf{1}(\mathbf{1}^T \mathbf{1})^{-1} \mathbf{1}^T P_X + \mathbf{1}(\mathbf{1}^T \mathbf{1})^{-1} \mathbf{1}^T \mathbf{1}(\mathbf{1}^T \mathbf{1})^{-1} \mathbf{1}^T \\
 &= X(X^T X)^{-1} X^T - \mathbf{1}(\mathbf{1}^T \mathbf{1})^{-1} \mathbf{1}^T - \mathbf{1}(\mathbf{1}^T \mathbf{1})^{-1} (P_X \mathbf{1})^T + \mathbf{1}(\mathbf{1}^T \mathbf{1})^{-1} \mathbf{1}^T \\
 &= X(X^T X)^{-1} X^T - \mathbf{1}(\mathbf{1}^T \mathbf{1})^{-1} \mathbf{1}^T \\
 &= P_X - P_1.
 \end{aligned}$$

So $(P_X - P_1)$ is idempotent.

3. 5 points To show B satisfy the definition of a generalized inverse of A, you must check if $ABA = A$.

$$\begin{aligned}
ABA &= A(BA) = \begin{bmatrix} 1 \\ 2 \\ 5 \\ -2 \end{bmatrix} \cdot ([1, 0, 0, 0] \cdot \begin{bmatrix} 1 \\ 2 \\ 5 \\ -2 \end{bmatrix}) \\
&= \begin{bmatrix} 1 \\ 2 \\ 5 \\ -2 \end{bmatrix} \cdot 1 \\
&= A.
\end{aligned}$$

B is a generalized inverse matrix for A by definition.

4. 10 points

If there exists matrix A such that $c^T\beta = AEY$, denote $A = [a_1, a_2, a_3, \dots,]$

(a)

$$\begin{aligned}
c^T\beta &= \alpha_1 - \frac{1}{2}(\alpha_2 + \alpha_3) \\
&= (\mu + \alpha_1) - \frac{1}{2}[(\mu + \alpha_2) + (\mu + \alpha_3)] \\
&= EY_{11} - \frac{1}{2}(EY_{21} + EY_{31}) \\
&= \left[1, 0, -\frac{1}{2}, 0, 0, 0, -\frac{1}{2}, 0\right] EY.
\end{aligned}$$

$A = [1, 0, \frac{1}{2}, 0, 0, 0, \frac{1}{2}, 0]$ is one non-random matrix so that AEY is the specified linear function of the model parameters. Or any A such that $a_1 + a_2 = 1, a_3 + a_4 + a_5 + a_6 = a_7 + a_8 = -\frac{1}{2}$ is correct.

(b)

$$\begin{aligned}
c^T\beta &= 3\mu + \alpha_1 + 2\alpha_2 \\
&= (\mu + \alpha_1) + 2(\mu + \alpha_2) \\
&= EY_{11} + 2EY_{21} \\
&= [1, 0, 2, 0, 0, 0, 0, 0]EY.
\end{aligned}$$

One non-random matrix A can be $[1, 0, 2, 0, 0, 0, 0, 0]$ or any A such that $a_1 + a_2 = 1, a_3 + a_4 + a_5 + a_6 = 2, a_7 + a_8 = 0$ is correct.

(c) If there exists a matrix A such that $c^T\beta = AEY$, then

$$\begin{aligned}
\alpha_2 + \alpha_3 &= AEY = (a_1 + a_2)(\mu + \alpha_1) + (a_3 + a_4 + a_5 + a_6)(\mu + \alpha_2) + (a_7 + a_8)(\mu + \alpha_3) \\
&= (a_1 + a_2 + a_3 + a_4 + a_5 + a_6 + a_7 + a_8)\mu + (a_1 + a_2)\alpha_1 + (a_3 + a_4 + a_5 + a_6)\alpha_2 + \\
&\quad (a_7 + a_8)\alpha_3,
\end{aligned}$$

the above equation requires $(a_1 + a_2 + a_3 + a_4 + a_5 + a_6 + a_7 + a_8) = 0$, $(a_1 + a_2) = 0$, $(a_3 + a_4 + a_5 + a_6) = 1$ and $(a_7 + a_8) = 1$, no solutions exist that can satisfy all the above three equations.

(d) If there exists a matrix A such that $c^T \beta = AEY$, then

$$\begin{aligned} 3\mu - \alpha_1 - \alpha_2 - \alpha_3 &= AEY = (a_1 + a_2)(\mu + \alpha_1) + (a_3 + a_4 + a_5 + a_6)(\mu + \alpha_2) + (a_7 + a_8)(\mu + \alpha_3) \\ &= (a_1 + a_2 + a_3 + a_4 + a_5 + a_6 + a_7 + a_8)\mu + (a_1 + a_2)\alpha_1 + \\ &\quad (a_3 + a_4 + a_5 + a_6)\alpha_2 + (a_7 + a_8)\alpha_3, \end{aligned}$$

the above equation requires $(a_1 + a_2 + a_3 + a_4 + a_5 + a_6 + a_7 + a_8) = 3$, $(a_1 + a_2) = -1$, $(a_3 + a_4 + a_5 + a_6) = -1$ and $(a_7 + a_8) = -1$, no solutions exist that can satisfy all the above three equations.

5. 15 points

- (a) μ is not estimable. No individual components of the parameter vector β is estimable as they are not identifiable with not full-ranked X .
- (b) $\alpha_1 - \alpha_2$ is not estimable. We need to find a linear combination of expected values $EY_{ijl} = \mu + \alpha_i + \tau_j + (\alpha\tau)_{ij}$ that equals to $\alpha_1 - \alpha_2$, which is not possible because of the interaction terms $(\alpha\tau)_{ij}$.
- (c) $(\alpha\tau)_{12}$ is not estimable with the same reason as part (a).
- (d) $(\alpha\tau)_{11} - (\alpha\tau)_{12}$ is not estimable. We need to find a linear combination of expected values $EY_{ijl} = \mu + \alpha_i + \tau_j + (\alpha\tau)_{ij}$ that equals to $(\alpha\tau)_{11} - (\alpha\tau)_{12}$, which is not possible because of the terms τ_j .
- (e) $(\alpha\tau)_{11} - (\alpha\tau)_{12} - (\alpha\tau)_{21} - (\alpha\tau)_{22}$ is not estimable.
We want to find a, b, c, d such that $E[aY_{11k} + bY_{12k} + cY_{21k} + dY_{22k}] = (\alpha\tau)_{11} - (\alpha\tau)_{12} - (\alpha\tau)_{21} - (\alpha\tau)_{22}$, thus $a = 1, b = c = d = -1$ has to hold. But then the coefficient for μ is $1 - 1 - 1 - 1 = -2 \neq 0$, so such set of a, b, c, d doesn't exist.
- (f) $(\alpha\tau)_{11} - (\alpha\tau)_{12} - (\alpha\tau)_{21} + (\alpha\tau)_{22}$ is estimable.

$$\begin{aligned} (EY_{11k} - EY_{12k}) - (EY_{21k} - EY_{22k}) &= [(\mu + \alpha_1 + \tau_1 + (\alpha\tau)_{11}) - (\mu + \alpha_1 + \tau_2 + (\alpha\tau)_{12})] - \\ &\quad [(\mu + \alpha_2 + \tau_1 + (\alpha\tau)_{21}) - (\mu + \alpha_2 + \tau_2 + (\alpha\tau)_{22})] \\ &= (\alpha\tau)_{11} - (\alpha\tau)_{12} - (\alpha\tau)_{21} + (\alpha\tau)_{22}. \end{aligned}$$