

1 a) See slide 1 of our course notes.

b) $X(X'X)^{-1}X'y$

c) $X'X\underline{b} = X'y$

d) Yes. $X(X'X)^{-1}X'y = X(X'X)^{-1}X'X\underline{b}$
 $= P_X X \underline{b}$
 $= X \underline{b}.$

2. Let A be any symmetric and idempotent matrix. Then

$$\begin{aligned} A &= A A^T A = A (A A)^T A \\ &= A (A'^T A)^T A' = P_A. \end{aligned}$$

Thus, A is the orthogonal projection matrix that projects onto $C(A)$.

3a) $\begin{bmatrix} 1 & 1 & 0 & -1 \\ 1 & 1 & 0 & -1 \\ 1 & 1 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 1 & 1 & 0 & 1 \\ 1 & 1 & 0 & 1 \\ 1 & 0 & 1 & -1 \\ 1 & 0 & 1 & -1 \\ 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 1 \\ 1 & 0 & 1 & 1 \end{bmatrix}$

b) A quantity is estimable if and only if it can be written as a linear combination of $E(Y) = X\beta$. or, equivalently, if there is a linear function of Y whose expectation is equal to the quantity.

i) $\mu + s_1 = E(Y_{121})$. Thus, $\mu + s_1$ is estimable.

ii) $\mu + s_1 + 10\beta = E(Y_{121} + 10(Y_{131} - Y_{121})) = E(10Y_{131} - 9Y_{121})$
 Thus, $\mu + s_1 + 10\beta$ is estimable.

3 b) (continued)

(ii) $s_1 - s_2 = E(Y_{121} - Y_{221}).$

Thus, $s_1 - s_2$ is estimable.

(v) In order for μ to be estimable, we need to find $\underline{\alpha}$ so that $\underline{\alpha}' X \beta = \mu$, i.e

$$\underline{\alpha}' X = [1, 0, 0, 0],$$

$$\underline{\alpha}' X = [1, 0, 0, 0] \Rightarrow \sum_{m=1}^{12} \alpha_m = 1, \text{ and}$$

$$\sum_{m=1}^6 \alpha_m = 0, \text{ and}$$

$$\sum_{m=7}^{12} \alpha_m = 0.$$

Because $\sum_{m=1}^6 \alpha_m = 0$ and $\sum_{m=7}^{12} \alpha_m = 0 \Rightarrow \sum_{m=1}^{12} \alpha_m = 0,$

there does not exist $\underline{\alpha}$ such that $\underline{\alpha}' X \beta = \mu$. Therefore, μ is not estimable.

There are many other ways to complete
3 b). Using the method of HW2, Problem 7
would work like this.

Suppose $X\underline{d} = \underline{0}$. Then

$$d_1 + d_2 - d_4 = 0$$

$$d_1 + d_2 = 0 \quad d_1 = -d_2 = -d_3$$

$$d_1 + d_2 + d_4 = 0 \iff \text{and}$$

$$d_1 + d_3 - d_4 = 0 \quad d_4 = 0$$

$$d_1 + d_3 = 0$$

$$d_1 + d_3 + d_4 = 0$$

which means that $X\underline{d} = \underline{0} \iff$

\underline{d} is of the form $[d, -d, -d, 0]'$, $d \in \mathbb{R}$.

Now note that

i) $M + S_1 = [1, 1, 0, 0] \beta$

$$[1, 1, 0, 0] \begin{bmatrix} d \\ -d \\ -d \\ 0 \end{bmatrix} = 0 \quad \forall d \in \mathbb{R}.$$

Thus, $M + S_1$ is estimable.

ii) $M + S_1 + 10\beta = [1, 1, 0, 10] \beta$

$$[1, 1, 0, 10] \begin{bmatrix} d \\ -d \\ -d \\ 0 \end{bmatrix} = 0 \quad \forall d \in \mathbb{R}$$

Thus, $M + S_1 + 10\beta$ is estimable.

iii) $S_1 - S_2 = [0, 1, -1, 0] \beta$

$$[0, 1, -1, 0] \begin{bmatrix} d \\ -d \\ -d \\ 0 \end{bmatrix} = 0 \quad \forall d \in \mathbb{R}$$

Thus, $S_1 - S_2$ is estimable.

iv) $M = [1, 0, 0, 0] \beta \quad [1, 0, 0] \begin{bmatrix} d \\ -d \\ -d \\ 0 \end{bmatrix} = d \neq 0 \quad \forall d \in \mathbb{R}$

Thus, M is NOT ESTIMABLE.

3c) THE X MATRIX IN 3(a) HAS RANK 3.
 ANY ONE OF THE FIRST THREE COLUMNS CAN BE
 WRITTEN AS A LINEAR COMBINATION OF THE
 OTHER TWO. R WOULD DISCARD THE SECOND
 COLUMN OF THE MATRIX, BUT CALCULATIONS
 ARE MUCH EASIER IF COLUMNS OF THE
 MODEL MATRIX ARE ORTHOGONAL. THUS, I
 WILL REMOVE THE FIRST COLUMN AND USE

$$X = \begin{bmatrix} 1 & 0 & -1 \\ 1 & 0 & -1 \\ 1 & 0 & 0 \\ 1 & 0 & 0 \\ 1 & 0 & 1 \\ 1 & 0 & 1 \\ 0 & 1 & -1 \\ 0 & 1 & -1 \\ 0 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 1 & 1 \\ 0 & 1 & 1 \end{bmatrix}$$

AS THE MODEL MATRIX.

OUR MODEL SPECIFICATION IN
 THE PROBLEM STATEMENT

$$\text{IMPLIES THAT } E(\gamma) = \begin{bmatrix} M+S_1-\beta \\ M+S_1-\beta \\ M+S_1 \\ M+S_1 \\ M+S_1+\beta \\ M+S_1+\beta \\ M+S_2-\beta \\ M+S_2-\beta \\ M+S_2 \\ M+S_2 \\ M+S_2+\beta \\ M+S_2+\beta \end{bmatrix}$$

IT IS EASY TO SEE THAT MY X MATRIX
 MULTIPLIED ON THE RIGHT BY $\begin{bmatrix} M+S_1 \\ M+S_2 \\ \beta \end{bmatrix}$ WILL GIVE $E(\gamma)$.

$$\text{THUS, } (X'X)^{-1}X'Y = \hat{\beta} = \begin{bmatrix} \hat{M+S_1} \\ \hat{M+S_2} \\ \hat{\beta} \end{bmatrix}$$

$$\begin{bmatrix} M+S_1-\beta \\ M+S_1-\beta \\ M+S_1 \\ M+S_1 \\ M+S_1+\beta \\ M+S_1+\beta \\ M+S_2-\beta \\ M+S_2-\beta \\ M+S_2 \\ M+S_2 \\ M+S_2+\beta \\ M+S_2+\beta \end{bmatrix}$$

3d) The BLUE of $E(Y_{111})$ is

$$\underbrace{\begin{bmatrix} 1 & 0 & -1 \end{bmatrix}}_{\text{First row of } X} \hat{\beta}, \text{ where } \hat{\beta} = (\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'\mathbf{y}.$$

$$\mathbf{X}'\mathbf{X} = \begin{bmatrix} 6 & 0 & 0 \\ 0 & 6 & 0 \\ 0 & 0 & 8 \end{bmatrix} \quad (\mathbf{X}'\mathbf{X})^{-1} = \begin{bmatrix} \frac{1}{6} & 0 & 0 \\ 0 & \frac{1}{6} & 0 \\ 0 & 0 & \frac{1}{8} \end{bmatrix}$$

$$\mathbf{X}'\mathbf{y} = \begin{bmatrix} \bar{Y}_{1..} \\ \bar{Y}_{2..} \\ \bar{Y}_{3..} - \bar{Y}_{1..} \end{bmatrix} \quad (\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'\mathbf{y} = \begin{bmatrix} \bar{Y}_{1..} \\ \bar{Y}_{2..} \\ \frac{1}{2}(\bar{Y}_{3..} - \bar{Y}_{1..}) \end{bmatrix}$$

Therefore, $\widehat{E(Y_{111})} = \bar{Y}_{1..} - \frac{1}{2}(\bar{Y}_{3..} - \bar{Y}_{1..})$.

3e) $\widehat{\mu + s_1} = \bar{Y}_{1..}$

$$\widehat{\mu + s_1 + 10\beta} = \bar{Y}_{1..} + 5(\bar{Y}_{3..} - \bar{Y}_{1..})$$

$$\widehat{s_1 - s_2} = \bar{Y}_{1..} - \bar{Y}_{2..}$$

4a) The F-test at the very bottom of the output tests whether all model coefficients other than the intercept are simultaneously equal to ϕ . Thus, this is a test of one mean for all observations vs. one mean for each combination of source and protein amount (the cell means model).

$$F = 39.84$$

$$df = 5, 6$$

$$p\text{-value} = 0.0001587$$

There is strong evidence of a difference among treatment means.

b) The design matrix used by R is

μ	s	x	x^2	$sinx$	$sinx^2$
1	0	-1	1	0	0
1	0	-1	1	0	0
1	0	0	0	0	0
1	0	0	0	0	0
1	0	1	1	0	0
1	0	1	1	0	0
1	1	-1	1	-1	1
1	1	-1	1	-1	1
1	1	0	0	0	0
1	1	0	0	0	0
1	1	1	1	1	1
1	1	1	1	1	1

The mean for Source 2 , Protein Amount 3 is given by either of the last two rows times $\hat{\beta}$. Thus, the estimate is the sum of $\hat{\beta}_1 = 17.5 + 5 + 6.25 - .25 + 3.75 + .75$
 $= 33$

4b) This can also be seen by noting that the model is

$$Y_{ijk} = \beta_0 + \beta_1 S(i) + \beta_2 X_j + \beta_3 X_j^2 + \beta_4 S(i)*X_j + \beta_5 S(i)*X_j^2 + \varepsilon_{ijk}$$

Where $S(i) = \begin{cases} 1 & \text{if } i=2 \\ 0 & \text{if } i=1 \end{cases}$ and $X_j = j-2$.

Thus, for $i=2, j=3$, the mean is

$$\begin{aligned} & \beta_0 + \beta_1(1) + \beta_2(3-2) + \beta_3(3-2)^2 + \beta_4(1)(3-2) + \beta_5(1)(3-2)^2 \\ &= \beta_0 + \beta_1 + \beta_2 + \beta_3 + \beta_4 + \beta_5. \end{aligned}$$

c) The standard error of any cell mean

is $\sqrt{\hat{V}\text{ar}(\bar{Y}_{ij\cdot})} = \sqrt{\hat{\sigma}^2/2}$

$$= \sqrt{MSE/2}$$

$$= \sqrt{3.25/2} \quad \leftarrow$$

Also, the estimated intercept is an estimate of M_{11} , which has the same standard error as \hat{M}_{23} . R gives 1.2748 as SE for intercept 10

4d) From part (b), we can see that
the average of the source 1 means

is $[1, 0, 0, \frac{2}{3}, 0, 0] \beta$



Average of 1st 6 rows of X.

The average of the source 2 means

is $[1, 1, 0, \frac{2}{3}, 0, \frac{2}{3}] \beta$.



Average of the last 6 rows of X

Thus, $C = [0, 1, 0, 0, 0, \frac{2}{3}]$ will do.



Difference of vectors above.

$$4 \text{ e)} (19.5 + 0.37 + 28.13 + 0.04) / 9$$

$$4 \text{ f)} F = \frac{(SSE_{\text{Reduced}} - SSE_{\text{Full}}) / (dfe_{\text{Reduced}} - dfe_{\text{Full}})}{MSE_{\text{Full}}}$$
$$= \frac{(0.37 + 28.13 + .04) / (9 - 6)}{3.25}$$

$$df = 3, 6$$

$$5. \quad W \sim \chi_m^2 (\underline{\sigma}^2) \Rightarrow W = (\underline{Z} + \underline{\delta})'(\underline{Z} + \underline{\delta})$$

where $\underline{Z} \sim N(\underline{0}, I)$ and $\underline{\delta} = \begin{bmatrix} \underline{\delta} \\ \vdots \\ \underline{\delta} \end{bmatrix}_{m \times 1}$.

$$a) \quad E(W) = E(\underline{Z}'\underline{Z} + 2\underline{\delta}'\underline{Z} + \underline{\delta}'\underline{\delta})$$

$$= E(\underline{Z}'\underline{Z}) + 2\underline{\delta}'E(\underline{Z}) + \underline{\delta}'\underline{\delta}$$

$$= E(\chi_m^2) + 2\underline{\delta}'\underline{\delta} + \underline{\delta}'\underline{\delta}$$

$$= m + \underline{\delta}'\underline{\delta}$$

$$b) \quad \text{Var}(W) = \text{Var}(\underline{Z}'\underline{Z} + 2\underline{\delta}'\underline{Z} + \underline{\delta}'\underline{\delta})$$

$$= \text{Var}\left(\sum_{i=1}^m z_i^2 + 2\underline{\delta}'\underline{z}_1\right)$$

$$= \text{Var}\left(\sum_{i=1}^m z_i^2\right) + 4\underline{\delta}'\text{Var}(\underline{z}_1) + 0$$

(because $\text{cov}(z_1^2 + \dots + z_m^2, z_1) = \text{cov}(z_1^2, z_1) = 0$)

$$= \text{Var}(\chi_m^2) + 4\underline{\delta}'\underline{\delta} = 2m + 4\underline{\delta}'\underline{\delta}.$$

Point values were set as follows:

1. a) 4

b) 4

c) 4

d) 5

2. 6

3. a) 4

b) 4×4

c) 4

d) 6

e) 6

4. a) 5

b) 5

c) 5

d) 5

e) 5

f) 6

5. a) 5

b) 5