

STAT 510

EXAM 2 SOLUTIONS

SPRING 2018

POINTS POSSIBLE: 100

1 a) 8

1 b) 10

1 c) 8

2 a) 10

2 b) 8

2 c) 8

3 a) 10

3 b) 8

4 a) 10

4 b) 10

4 c) 10

$$\begin{aligned}
 |a) \text{ } \text{Cov}(y_{111}, y_{122}) &= \text{Cov}(M_{11} + e_{111}, M_{11} + e_{122}) \\
 &= \text{Cov}(M_{11}, M_{11}) \\
 &= \text{Var}(M_{11}) \\
 &= \sigma_m^2
 \end{aligned}$$

$$\text{Var}(y_{iijk}) = \text{Var}(M_{ii} + e_{iijk}) = \sigma_m^2 + \sigma_e^2$$

$$\text{Thus, } \text{Corr}(y_{111}, y_{122}) = \frac{\sigma_m^2}{\sigma_m^2 + \sigma_e^2}$$

|b) Let M = MATTERLESS TYPE
 L = LIGHT
 S = SOUND

SOURCE	DF
M	2
man(M)	15 ← WHOLE-PLOT ERROR
L	1
S	1
L×S	1
M×L	2
M×S	2
M×L×S	2
SPLIT-PLOT ERROR	45 ← $L \times \text{man}(M) + S \times \text{man}(M) + L \times S \times \text{man}(M)$
C. TOTAL	71

$$1c) X = \underbrace{I}_{3 \times 3} \otimes \underbrace{\frac{1}{6 \times 1}} \otimes \underbrace{I}_{4 \times 4}$$

$$Z = \underbrace{I}_{18 \times 18} \otimes \underbrace{\frac{1}{4 \times 1}}$$

$$2a) Y' (P_{[1,A]} - P_1) Y = \| \hat{Y}_{[1,A]} - \hat{Y}_1 \|^2$$

AVERAGE OF Y VALUES FOR A = 1 IS $\frac{2 \times 3.0 + 8 \times 5.0}{10} = 4.6$

AVERAGE OF Y VALUES FOR A = 2 IS $\frac{6 \times 7.0 + 4 \times 3.0}{10} = 5.4$

AVERAGE OF ALL Y VALUES IS $\frac{4.6 + 5.4}{2} = 5.0$

$$\text{THUS, } \| \hat{Y}_{[1,A]} - \hat{Y}_1 \|^2 = 10(4.6 - 5.0)^2 + 10(5.4 - 5.0)^2 \\ = 10(.4)^2 + 10(.4)^2 \\ = 20(.4)^2 = 20(.16) \\ = 3.2$$

2b)

		B	
		1	2
A	1	μ_{11}	μ_{12}
	2	μ_{21}	μ_{22}

$\bar{M}_{1\cdot}$
 $\bar{M}_{2\cdot}$

LSMEAN For A2

IS BLUE OF

$$\bar{M}_{2\cdot} = \frac{\mu_{21} + \mu_{22}}{2}$$

$$\hat{\bar{M}}_{2\cdot} = \frac{\hat{\mu}_{21} + \hat{\mu}_{22}}{2} = \frac{\bar{Y}_{21\cdot} + \bar{Y}_{22\cdot}}{2}$$

$$= \frac{7.0 + 3.0}{2} = 5.0$$

$$2c) \text{ LET } X = \begin{bmatrix} \frac{1}{2x_1} & 0 & 0 & 0 \\ 0 & \frac{1}{6x_1} & 0 & 0 \\ 0 & 0 & \frac{1}{6x_1} & 0 \\ 0 & 0 & 0 & \frac{1}{4x_1} \end{bmatrix} \quad \beta_r = \begin{bmatrix} \mu_{11} \\ \mu_{12} \\ \mu_{21} \\ \mu_{22} \end{bmatrix}$$

$$(X'X)^{-1} = \begin{bmatrix} \frac{1}{2} & 0 & 0 & 0 \\ 0 & \frac{1}{8} & 0 & 0 \\ 0 & 0 & \frac{1}{6} & 0 \\ 0 & 0 & 0 & \frac{1}{4} \end{bmatrix} \quad C'_1 = \begin{bmatrix} 1 & 1 & -1 & -1 \end{bmatrix}$$

$$C'_2 = \begin{bmatrix} 1 & -1 & 1 & -1 \end{bmatrix}$$

$$C'_1 (X'X)^{-1} C'_2 = \left[\frac{1}{2}, \frac{1}{8}, -\frac{1}{6}, -\frac{1}{4} \right] \begin{bmatrix} \frac{1}{2} \\ -\frac{1}{8} \\ \frac{1}{6} \\ -\frac{1}{4} \end{bmatrix}$$

$$= \frac{1}{2} - \frac{1}{8} - \frac{1}{6} + \frac{1}{4}$$

$$= \frac{12 - 3 - 4 + 6}{24} = \frac{11}{24} \neq 0$$

NOT ORTHOGONAL CONTRASTS

$$3a) (\underline{I} + a \underline{1} \underline{1}')' = \underline{I}' + a(\underline{1} \underline{1}')' = \underline{I} + a \underline{1} \underline{1}',$$

So $\underline{I} + a \underline{1} \underline{1}'$ is symmetric. Now suppose $x \in \mathbb{R}^n \setminus \{0\}$.

$$\text{Then } \underline{x}' \vee \underline{x} = \underline{x}' (\underline{I} + a \underline{1} \underline{1}') \underline{x}$$

$$= \underline{x}' \underline{x} + a \underline{x}' \underline{1} \underline{1}' \underline{x}$$

$$= \sum_{i=1}^n x_i^2 + a \left(\sum_{i=1}^n x_i \right)^2 \quad \begin{array}{l} (\geq 0 \text{ BECAUSE}) \\ a > 0 \end{array}$$

$$\geq \sum_{i=1}^n x_i^2 > 0$$

(BECAUSE $\underline{x} \neq 0 \Rightarrow x_i^2 > 0$ For At Least ONE $i \in \{1, \dots, n\}$.)

$$3b) \nabla X = \left[\underline{\underline{I}}_{n \times n} + a \underline{1}_n \underline{1}_n' \right] X$$

$$= X + a \underline{1}_n \underline{1}_n' X$$

$$= X + a X \underline{1}_p \underline{1}_n' X \quad (\underline{1}_n = X \underline{1}_p)$$

$$= X \left[\underline{\underline{I}}_{p \times p} + a \underline{1}_p \underline{1}_n' X \right]$$

$$= X Q, \text{ WHERE } Q = \left[\underline{\underline{I}}_{p \times p} + a \underline{1}_p \underline{1}_n' X \right]$$

\therefore OLS of estimable $C\beta$ is BLUE
of $C\beta$.

4 a)

$$\frac{2S_1 + 2S_2 + S_3 + 2S_4 + S_5 + 8e's - 2S_1 - 2S_2 - S_3 - S_4 - 2S_5 - 8e's}{8} = \frac{S_4 - S_5 + 8e's - 8e's}{8}$$

THESE DIFFERENT FROM THESE

VARIANCE OF ABOVE IS

$$\frac{1}{8^2} \left[2\sigma_s^2 + 16\sigma_e^2 \right] = \frac{\sigma_s^2}{32} + \frac{\sigma_e^2}{4}$$

4 b) SUBJECT

1

2

3

4

5

VARIANCE OF DIFF. OF TRT. AVERAGES

$$V\text{AR}(\bar{e}_{11} - \bar{e}_{12}) = \sigma_e^2 \left[\frac{1}{2} + \frac{1}{2} \right] = \sigma_e^2$$

$$V\text{AR}(\bar{e}_{21} - \bar{e}_{22}) = \sigma_e^2 \left[\frac{1}{2} + \frac{1}{2} \right] = \sigma_e^2$$

$$V\text{AR}(e_{311} - e_{321}) = 2\sigma_e^2$$

$$V\text{AR}(\bar{e}_{41} - e_{421}) = \sigma_e^2 \left[\frac{1}{2} + 1 \right]$$

$$V\text{AR}(e_{511} - \bar{e}_{521}) = \sigma_e^2 \left[1 + \frac{1}{2} \right]$$

THUS, VARIANCE OF AVERAGE OF DIFFERENCES IS

$$\left(\frac{1}{5} \right)^2 \left[\sigma_e^2 + \sigma_e^2 + 2\sigma_e^2 + 1.5\sigma_e^2 + 1.5\sigma_e^2 \right] = \frac{7\sigma_e^2}{25}$$

4c) WE KNOW THAT INVERSE VARIANCE WEIGHTING IS THE BEST WAY TO COMBINE MULTIPLE INDEPENDENT UNBIASED ESTIMATORS. SO, WE SHOULD WEIGHT THE DIFFERENCES OF TREATMENT AVERAGES AS FOLLOWS

$$\frac{1 \times 8 + 1 \times 15 + \frac{1}{2} \times 7 + \frac{1}{1.5} \times 1 + \frac{1}{1.5} \times (-2)}{1 + 1 + \frac{1}{2} + \frac{1}{1.5} + \frac{1}{1.5}}$$

$$= \frac{48 + 90 + 21 + 4 - 8}{6 + 6 + 3 + 4 + 4}$$

$$= \frac{155}{23}$$

THE KEY HERE IS THAT THE SUBJECT-SPECIFIC DIFFERENCES ARE INDEPENDENT OF EACH OTHER. MANY STUDENTS TRIED TO APPLY INVERSE VARIANCE WEIGHTING TO THE ANSWERS FROM (a) AND (b), BUT THESE ESTIMATORS ARE NOT INDEPENDENT. FURTHERMORE, THE BEST WAY TO COMBINE THEM WOULD DEPEND ON THE VALUES OF UNKNOWN VARIANCE COMPONENTS. THE ANSWER ABOVE TO (c) IMPROVES ON THE ANSWER IN (b) NO MATTER WHAT THE VALUES OF σ_s^2 AND σ_e^2 ARE. SEE THE NEXT TWO PAGES FOR SOME MORE COMMENTS ON THIS PROBLEM.

IF $\hat{\Theta}_1, \dots, \hat{\Theta}_n$ ARE INDEPENDENT UNBIASED ESTIMATORS OF A PARAMETER Θ WITH VARIANCES V_1, \dots, V_n , RESPECTIVELY, THEN

$$\text{VAR} \left(\sum_{i=1}^n \frac{1}{V_i} \hat{\Theta}_i / \sum_{i=1}^n \frac{1}{V_i} \right)$$

$$= \left(\frac{1}{\sum_{i=1}^n \frac{1}{V_i}} \right)^2 \sum_{i=1}^n \left(\frac{1}{V_i} \right)^2 V_i$$

$$= \frac{1}{\left(\sum_{i=1}^n \frac{1}{V_i} \right)^2} \sum_{i=1}^n \frac{1}{V_i} = \frac{1}{\sum_{i=1}^n \frac{1}{V_i}}$$

THUS, VARIANCE OF ESTIMATOR IN 4(c) IS

$$\sigma_e^2 \frac{1}{1+1+\frac{1}{2}+\frac{1}{1.5}+\frac{1}{1.5}} = \frac{6\sigma_e^2}{6+6+3+4+4}$$

$$= \frac{6\sigma_e^2}{23}.$$

THIS IS INDEED LESS THAN VARIANCE OF $\frac{7\sigma_e^2}{25}$
IN PART (b).

HOWEVER, THE ESTIMATOR DEFINED IN PART (c)
IS NOT THE BLUE. THIS CAN BE
SEEN BY NOTING THAT THE VARIANCE
OF THE ESTIMATOR IN PART (a)

$$\frac{\sigma_s^2}{32} + \frac{\sigma_e^2}{4} \text{ COULD BE LESS THAN}$$

$\frac{6\sigma_e^2}{23}$ IF σ_s^2 IS SUFFICIENTLY
SMALL. AS σ_s^2 APPROACHES 0, THE
ESTIMATOR IN PART (a) BECOMES
MORE SIMILAR TO THE BLUE. AT $\sigma_s^2 = 0$,
THE PART (a) ESTIMATOR IS THE BLUE.