

4) controlling δ_F increases $P(\text{at least one type I error})$
 and there are many procedures for
 multiple comparisons.

end exam 3 material begin Quiz 11 material

ch6 1) P213 A: two way Anova model has
 2 factors A and B and a response Y.
 A has a levels and B has b levels.

2) know p 214 Factorial crossing is

used: each combination of an A level
 and a B level is used and called a
 treatment. There are ab treatments.
 Randomly assign units into treatments

or take random samples from ab populations.

p213

3) The cell means model is

$$Y_{ijk} = \mu_{ij} + \epsilon_{ijk}, \quad i=1, \dots, a \quad j=1, \dots, b \quad k=1, \dots, m$$

for
correct
inference

$Y_{ijk} \sim f(y - \mu_{ij})$, a location family with
 location parameter μ_{ij} and variance σ^2 .

$n_{ij} \equiv m$ and sample size $n = abm$

A	1	2	\vdots	B	b
1	m	m_2	...	m	
2	m	m	...	m	
\vdots	\vdots	\vdots	...	\vdots	
a	m	m	...	m	

randomly assign
 m units per cell
 or take random
 samples of size m
 from each of the ab
 populations determined by a cell

fitted values $\hat{Y}_{ijk} = \hat{\mu}_{ij} = \bar{Y}_{ijo}$
 residuals $e_{ijk} = Y_{ijk} - \hat{Y}_{ijk}$

4] * Anova table p 215

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	Source	df	SS	MS	F	Pval
main effects	A	a-1	SSA	MSA	$F_A = MSA/MSE$	$P(F_{a-1, n-ab} > F_A)$
	B	b-1	SSB	MSB	$F_B = MSB/MSE$	$P(F_{b-1, n-ab} > F_B)$
interaction	AB	(a-1)(b-1)	SSAB	MSAB	$F_{AB} = \frac{MSAB}{MSE}$	$P(F_{(a-1)(b-1), n-ab} > F_{AB})$
	error	n-ab = ab(n-1)	SSE	MSE		

5] know If there is an AB interaction, keep both the A and B main effects

6] p215

know for final, Q10 4 Step test for AB interaction

- H_0 no interaction H_A there is an interaction
- F_{AB}
- pval
- If pval < α reject H_0 and conclude there is an interaction between A and B.
If pval $\geq \alpha$ fail to reject H_0 and conclude that there is no interaction between A and B.

7] Tests for A and B main effects make the most sense if we fail to reject the test for interaction.

8] know for final, Q10 The test for A main effects is

- $H_0 \mu_{10} = \dots = \mu_{ao} \quad H_A$ not H_0
- F_A
- pval
- If pval < α , reject H_0 , conclude the mean response depends on the level of A. If pval $\geq \alpha$, fail to reject H_0 , conclude the mean response does not depend on the level of A.

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 Q3 Know for final, Q10 test for main effects B

- i) $H_0: \mu_{01} = \dots = \mu_{0b}$ $H_A: \text{not } H_0$
- ii) F_B } from output
- iii) P_{val}
- iv) If $P_{\text{val}} < \alpha$, reject H_0 , conclude the mean response depends on the level of B.
 If $P_{\text{val}} \geq \alpha$, fail to reject H_0 , conclude the mean response does not depend on the level of B.

10] The main effects F tests are just like the F test for the fixed effects 1-way Anova model. (If there is an AB interaction, failing to reject H_0 for a main effect implies that the main effect is small given that the interaction is in the model.)

so A and B effect in a compound manner

ex) Want to improve asphalt pavement

$y = \text{tensile strength}$, $A = \text{type: Basalt or Silicious}$, $B = \text{compaction method: Static, regular, low or very low}$

Source	df	SS	MS	F	Pval
A	1	1734.0	1734.0	182.53	0.000
B	3	16243.5	5414.0	569.95	0.000
AB	3	1145.0	381.67	40.18	0.000
error	16	152.0	9.50		

a) $H_0: \text{no interaction}$ $H_A: \text{there is an interaction}$

Mnemonic
 $F_{AB} = 40.18$
 $P_{\text{val}} = 0$

reject H_0 , there is an interaction between type of asphalt and compaction method.

b) $H_0: \mu_{10} = \mu_{20}$ $H_A: \text{not } H_0$ 57.5

$$F_A = 182.53$$

$$p\text{val} = 0.0$$

reject H_0 ; mean tensile strength depends on type of asphalt

c) $H_0: \mu_{11} = \mu_{12} = \mu_{13} = \mu_{14}$ $H_A: \text{not } H_0$

$$F_B = 569.95$$

$$p\text{val} = 0.0$$

reject H_0 ; mean tensile strength depends on compaction method.

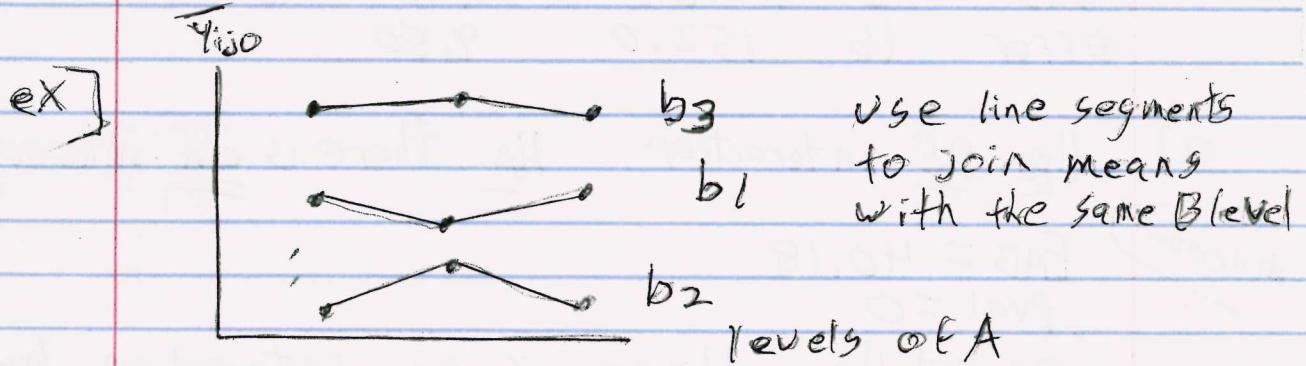
see ex 6.2.

11) Could do a 1way Anova on the ab treatments, but this procedure loses info about A, B and the AB interaction.

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12) * Interaction Plot for two way Anova:

Put the factor with the most levels on the horiz axis. For each level of the other factor, connect \bar{Y}_{ij0} with line segments

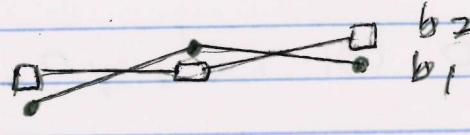


13) If there is no interaction, the "curves" will be roughly parallel.
For small $m \equiv n_{ij}$, "not crossing" may be adequate. "Crossing curves" suggests interaction unless the 2 curves are

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nearly the same. The interaction plot
is useful if the conclusions for the plot agree
with those for the F test for no interaction.

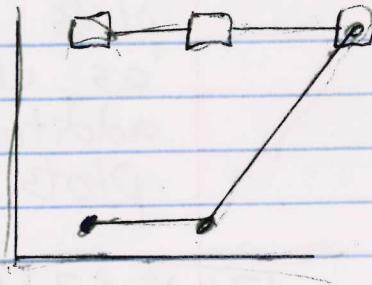
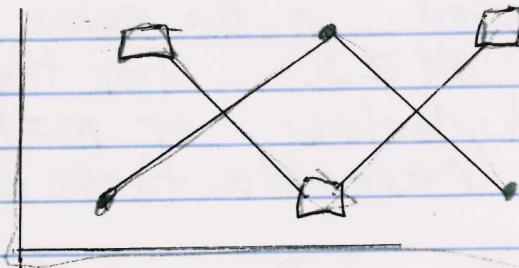
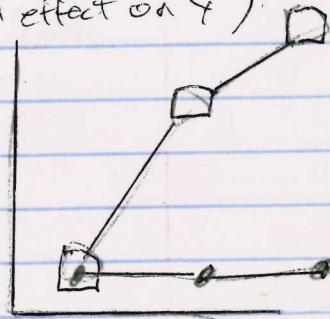
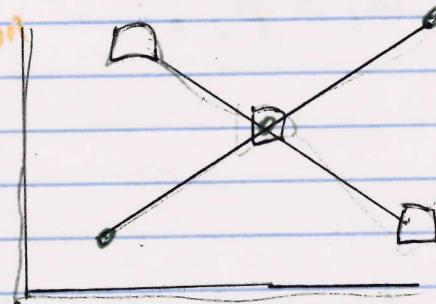
Software "fills space" so won't look like this plot. So need to look at the values on the y-axis.



If levels b₁ and b₂ give \approx the same mean response for each level of A, then the curves should be close and may cross, but there is no interaction.

(and the 2 levels have \approx the same mean effect on Y)

interaction but tests for main effects may fail to reject H₀



The above 4 plots suggest interaction if the values on the Y axis are far apart (need std errors of \bar{Y}_{ij0} 's). See ex 6.1.

14) p 218 Let $M_{ij} = \mu_{00} + \alpha_i + \beta_j + (\alpha\beta)_{ij}$.

The A main effects $\alpha_i = \mu_{i0} - \mu_{00}$

overparametrized model

$$\sum_i \alpha_i = 0 \quad \sum_j \beta_j = 0 \quad \beta_j = \mu_{j0} - \mu_{00} \quad \text{The interaction}$$

parameters $(\alpha\beta)_{ij} = \mu_{ij} - \mu_{i0} - \mu_{0j} + \mu_{00}$

(58.5)

$\sum_i (\alpha\beta)_{ij} = 0$ for $j=1, \dots, b$

$\sum_j (\alpha\beta)_{ij} = 0$ for $i=1, \dots, a$.

so $\sum_i \sum_j (\alpha\beta)_{ij} = 0$.

- 15) If there is no interaction, all $(\alpha\beta)_{ij} = 0$ and the factor effects are additive:

$$\mu_{ij} = \mu_{00} + \alpha_i + \beta_j.$$

- 16) Response, residual, and transformation plots are used in the same way as ch 5 if $m \geq 5$. For the additive model below or $m=1$, the plots often resemble those of MLR.

- 17) *P214 If A and B are factors, there are several models.

terms:

model

SAS notation

R notation

i) A, B and AB 2 way Anova

$y = A|B$

$y \sim A|B$ or $y \sim AB + A \times B$

ii) A and B additive or main effects

$y = A \text{ } B$

$y \sim A + B$

iii) A one way for A

$y = A$

$y \sim A$

iv) B one way for B

$y = B$

$y \sim B$

v) None null

?

$y \sim 1$

p
one

For the null model, the y_{itk} are iid
and $\bar{Y}_{000} = \hat{\mu}_{000}$