

Ex. 1 Age and memory

2-factor anova with interaction

Response Y : words recalled (out of 27)

Factor α : Age, 2 levels

Factor γ : Process, 5 levels

Model

mean recall age group j ,
processing method k

$$Y_{ijk} = \mu_{jk} + \epsilon_{ijk} \quad \epsilon_{ijk} \sim N(0, \sigma^2)$$

$$= \mu + \alpha_j + \gamma_k + (\alpha\gamma)_{jk} + \epsilon_{ijk}$$

global
mean

effect
of age
 j

effect
of
process
 k

added
effect of
combination jk

$$\sum_j \alpha_j = 0 \quad \sum_k \gamma_k = 0 \quad \sum_j (\alpha\gamma)_{jk} = 0 \quad \sum_k (\alpha\gamma)_{jk} = 0$$

R: `table(words)` `hist(words)`

NB: Y is a count $[0, 27]$

$$Y_{ijk} \sim \text{Binomial}(27, p_{jk})$$

probability that
subject in age grp. j
and process k recalls
a word

$$E(Y_{ijk}) = 27 p_{jk} = \mu_{jk}$$

Normal approximation to discrete distⁿ

From basic
course, ok
when
 $p \approx 0.5$

R: `boxplot(words ~ Age)`

- larger variation among younger?
- better recall younger?

this was already
known before the
study

R: boxplot (words ~ Process)

Counting, rhyming (least active) → poor recall
reorder → increasing (as expected)

R: boxplots (words ~ Age × Process)

aggregate (.....) ← averages in each group.

Interaction?

As in earlier plot, similarities for counting rhyming, larger differences for more 'active' processes.

NB: $\epsilon_{ijk} \sim N(0, \sigma^2)$ Issues? Maybe.

R: fit model and look at residuals

→ Seem centered around 0,
somewhat lower variance
for inactive processes, but
difficult to conclude with only
10 observations per group.

Tests: 1. Interaction?

✓ Yes
Stop?

↘ No
test main
effect of
age and process.

Recall:

$$\beta = \begin{bmatrix} \mu \\ \alpha_1 \\ \gamma_1 \\ \gamma_2 \\ \gamma_3 \\ \gamma_4 \\ (\alpha\gamma)_{11} \\ (\alpha\gamma)_{12} \\ (\alpha\gamma)_{13} \\ (\alpha\gamma)_{14} \end{bmatrix}$$

Annotations for the vector β :

- β_1 points to μ
- β_2 points to the group of γ terms ($\gamma_1, \gamma_2, \gamma_3, \gamma_4$)
- β_3 points to the group of $(\alpha\gamma)$ interaction terms

$\alpha_2 = -\alpha_1$ ← younger (alphabetical)

$$\gamma_5 = -\gamma_1 - \gamma_2 - \gamma_3 - \gamma_4$$

↑ intentional
(since we re-ordered in R)

$$(\alpha\gamma)_2 = -(\alpha\gamma)_{11} \text{ etc.}$$

$$H_0: \vec{\beta}_3 = 0 \text{ vs } H_1: \vec{\beta}_3 \neq 0 \text{ test interaction}$$

R: Anova(model)

$p = 0.00027$ interaction is significant.

R: Summary(model)

① Find $\hat{\mu}_{11} = \hat{\mu} + \hat{\alpha}_1 + \hat{\gamma}_1 + (\hat{\alpha\gamma})_{11}$

$\hat{\mu}_{21} = \hat{\mu} + \hat{\alpha}_2 + \hat{\gamma}_1 + (\hat{\alpha\gamma})_{21}$

$$\hat{\mu}_{11} = 11.61 - 1.55 - 4.86 + 1.8 = 7 \text{ words}$$

Annotations for the calculation of $\hat{\mu}_{11}$:

- 11.61: mean
- 1.55: α_1 (labeled "old is negative")
- 4.86: γ_1 (labeled "counting is a very bad strategy")
- +1.8: $(\alpha\gamma)_{11}$ (labeled "'add' back the age effect")

$$\hat{\mu}_{21} = 11.61 + 1.55 - 4.86 - 1.8 = 6.5 \text{ words}$$

R: compare to mean, aggregate -

③ Find $\hat{\mu}_{14} = \hat{\mu} + \hat{\alpha}_1 + \hat{\gamma}_4 + (\hat{\alpha}\hat{\gamma})_{14}$
 $\hat{\mu}_{24} = \hat{\mu} + \hat{\alpha}_2 + \hat{\gamma}_4 + (\hat{\alpha}\hat{\gamma})_{24}$

$\hat{\mu}_{14} = 11.61 - 1.55 + 3.89 - 0.55$
 $= 13.4 \text{ words}$

← age is neg. *← imagery good.* *← perhaps even worse to be old...*

$\hat{\mu}_{24} = 11.61 + 1.55 + 3.89 + 0.55 = 17.6 \text{ words}$

Note: these predictions we also get from group means, but perhaps additional insight from model?

- older age negative, but only for 'active' processes.
 ↳ From book: perhaps they just don't try hard enough.

Discuss: Figure vs. model summary ...

Exam 2018, problem 3

response y

factor β , 4 levels

15 x 4 observations

$$y_{ij} = \beta_j + \varepsilon_{ij} \quad \varepsilon_{ij} \sim N(0, \sigma^2) \quad j=1, \dots, 4$$

↑
note no grand mean,
each β_j is the level mean.

a)

Aim $Y = X\beta + \varepsilon$

$$\beta = \begin{pmatrix} \beta_1 \\ \beta_2 \\ \beta_3 \\ \beta_4 \end{pmatrix}$$

← β_0 , but we count
factor levels from 1,
and no intercept (β_0)
in this model.

For X , assume now 2 observations (not 15)

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

NB third
method
for coding
 X !

$$(X^T X) \stackrel{\text{here}}{=} \begin{bmatrix} 2 & 0 & 0 & 0 \\ 0 & 2 & 0 & 0 \\ 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 2 \end{bmatrix} \quad (X^T X)^{-1} = \frac{1}{2} I$$

in exercise, replace 2 with 15.

$$b) \hat{\beta}_3 = 1.0902858$$

$$\hat{\beta}_4 = 0.1752633$$

$$SSE = 43.04524$$

$$H_0: \beta_3 = \beta_4 \quad H_1: \beta_3 \neq \beta_4$$

$$\text{Recall: } F = \frac{(C\hat{\beta})^T (C(X^T X)^{-1} C^T)^{-1} (C\hat{\beta}) / r}{SSE / (n-p)}$$

\nwarrow r x p - matrix \swarrow restrictions
 \uparrow length of $\vec{\beta}$

$$\text{Here: } \beta_3 = \beta_4 \rightarrow \beta_3 - \beta_4 = 0$$

$$\rightarrow C\beta = \begin{bmatrix} 0 & 0 & 1 & -1 \end{bmatrix} \beta = 0 \quad r=1.$$

$$p=4$$

$$n-p=56$$

$$C\hat{\beta} = \hat{\beta}_3 - \hat{\beta}_4 = 0.9150225$$

$$C(X^T X)^{-1} C^T = \frac{1}{15} \begin{bmatrix} 0 & 0 & -1 & 1 \end{bmatrix} I \begin{bmatrix} 0 \\ 0 \\ -1 \\ 1 \end{bmatrix} = \frac{2}{15}$$

$$F = \frac{(0.9150225)^2 \cdot \frac{15}{2} / 1}{43.04524 / 56} = 8.169$$

critical value F-dist, $F_{1,56}$, $\alpha=0.05$

$\approx 4 \rightarrow \text{reject } H_0 \Rightarrow \beta_3 \neq \beta_4$

c) Bonferroni $m=6$ tests

$$\alpha_{loc} = \frac{0.05}{6} = 0.0083$$

Reject $\beta_2 = \beta_3$ and $\beta_3 = \beta_4$.

Exam 2010

Response y : blood pressure reduction

Factor α : treatment, 3 levels

↑
one level is
"no intervention"

$$Y_{ij} = \mu_j + \varepsilon_{ij}$$

$$\text{Test: } \begin{cases} H_0: \mu_H = \mu_{No} \\ (*) H_0: \mu_H = \mu_{No} \end{cases}$$

$$(**) H_0: \mu_H = \mu_M$$

} comparing interventions
to a baseline

↓
use dummy
coding

(*) Dummy coding:

$$\beta = \begin{bmatrix} \mu_{No} \\ \mu_H - \mu_{No} \\ \mu_M - \mu_{No} \end{bmatrix}$$

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

$$\beta_1 - \beta_2 = \mu_H - \mu_{No} - (\mu_M - \mu_{No}) = \mu_H - \mu_M$$

$$\text{Test } \beta_1 = \beta_2$$

$$C\beta = 0$$

$$C = [0 \ 1 \ -1]$$