

Case Study 3: Dressings

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STAT 8801 Statistical Consulting

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Recap of Situation: What is the Question?

- ▶ Goal: compare a new dressing for minor operations to a standard dressing
- ▶ Experiment procedure:
 - ▶ two sites selected on each subject →
 - ▶ wounds be simulated at each site →
 - ▶ each subject has new dressing at one site and standard dressing at the other
- ▶ Measurement procedure:
 - ▶ outcomes (on a 1-5 scale): pain, infection level, healing speed, etc.
 - ▶ repeated at 7 days, 14 days, and 42 days after the surgery

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 - ▶ categorical: trajectory pattern (e.g., increasing, flat, decreasing, etc.)
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- ▶ Collect as detailed personal information as possible at baseline (e.g., gender, age, socioeconomic status, etc.)
- ▶ Make efforts to ensure no subjects dropping out the study before it ends

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- ▶ Realistic constraints: money, time, effort

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Data Analysis: How to Carry Out the Data Analysis?

Consider this question: is the new dressing more effective in reducing infection level at day 7 compared to the standard dressing?

| Subject | Trt(D7) | Infection(D7) |
|---------|---------|---------------|
| 1 | New | 2 |
| 1 | Std | 4 |
| 2 | New | 4 |
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Key: Within-subject observations (e.g., rows 1 and 2) are correlated, whereas between-subject observations (e.g., subjects 1 and 2) are independent.

Approaches:

- ▶ Simplify responses
- ▶ Control for subject id
- ▶ Model the within-subject correlation explicitly

Data Analysis: How to Carry Out the Data Analysis? (cont.)

Simplify responses: one response for each subject

- ▶ Case I: continuous outcomes:
 - ▶ Compute the difference between the infection score under new dressing and that under standard dressing for each subject

| Subject | $I_{new}(D7)$ | $I_{std}(D7)$ | δ |
|---------|---------------|---------------|----------|
| 1 | 2 | 4 | -2 |
| 2 | 4 | 4 | 0 |

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- ▶ Paired t -test: $\text{mean}(\delta)$ different from zero?

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▶ Case II: binary outcomes:

- ▶ Choose a threshold, e.g., 3

| Subject | $I_{new}(D7)$ | $I_{std}(D7)$ | Profile |
|---------|---------------|---------------|---------|
| 1 | 2 | 4 | LH |
| 2 | 4 | 4 | HH |

...

- ▶ Pearson's residual test

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 - > `aov(infection ~ factor(subject) + factor(trt))`

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> `lmer(infection ~ factor(trt) + (1|subject))`
- ▶ Similar logic extends into repeated measures analysis (e.g., if responses from all three measurement occasions are considered)
> `lmer(infection ~ factor(trt) + (1|subject) + (1|subject: day))`