

Recap

- a paired t-test
- Bayesian inference: model selection  
 Bootstrap inference: create samples by allowing for replacement
- ANOVA (Analysis of variance) ← alternative to t-test especially in the case of more than two levels in your modifications.

$$y_{ij} = \alpha + \beta_i + e_{ij}$$

we have  $g$  - groups of levels  $i = 1, 2, \dots, g$   
 and at each ' $i$ ' level we have  $n_i$  samples  
 $j = 1, 2, \dots, n_i$

- $\beta_i \neq 0$   $y_{ij}$  follows a multiple mean model (at least one group level means are different)

$$\beta_i = 0 \Rightarrow \text{all means are equal (a single global mean)}$$

$g=4$ 
{

$B_1$	[ $y_{1,1}, y_{1,2}, \dots, y_{1,n_1}$ ]	# samples in group
$B_2$	"	"
$B_3$	"	"
$B_4$	"	"

$$\alpha \rightarrow \text{mean of } \sum_{i=1}^g n_i \text{ samples} = \bar{y}_{\cdot,\cdot}$$

$$\beta_i \rightarrow \text{mean of } \sum_{j=1}^{n_i} (y_{i,j} - \bar{y}_{\cdot,\cdot})$$

- single mean model (vs) multiple means model

$$\mu_i = \alpha + \beta_i \leftarrow \text{m.m.m}$$

$$\mu = \alpha \leftarrow \text{s.m.m}$$

- all means equal is Null hypothesis → single mean  
 at least one  $\mu_i$  is diff. Alter. hyp → multiple mean

$$y_{ij} = \alpha + \beta_i + e_{ij}$$

$$\mu_i + e_{ij} \sim N(\mu_i, \sigma^2)$$

$$\downarrow$$

$$N(0, \sigma^2)$$

$$= \mu + e_{ij} \sim N(\mu, \sigma^2)$$

- Occam's Razor - prefer simple model consistent w/ the data only move to complex when necessary (Bayesian model selection)

$$y_{i,j} = \alpha + \beta_i + e_{ij}$$

$$x_{i,j} = y_{i,j} - \bar{y}_{\cdot,\cdot} \quad (\text{single mean model})$$

↑  
global mean of the data

$$x_{i,j} = y_{i,j} - \bar{y}_{i,\cdot} \quad (\text{multiple mean model})$$

- least squares approach: sum of squares of residuals (SS)

$$x_{ij} = y_{ij} - \bar{y}_{\cdot,\cdot}$$

$$= (\bar{y}_{i,\cdot} - \bar{y}_{\cdot,\cdot}) + (y_{ij} - \bar{y}_{i,\cdot})$$

$$= \beta_i + e_{ij}$$

$$\sum_i^g \sum_j^{n_i} (y_{ij} - \bar{y}_{\cdot,\cdot})^2 = \sum_i^g \sum_j^{n_i} \beta_i^2 + \sum_i^g \sum_j^{n_i} e_{ij}^2 + 2 \sum_i^g \sum_j^{n_i} \beta_i e_{ij}$$

$$SS_{\text{Total}} = SS_{\text{level}} + SS_{\text{error}}$$

$$SS_{\text{level}} = \sum_j^{n_i} n_i \beta_i^2$$

- Total samples of 40 ( $g=4, n_i=10$ ) → fixed  $SS_{\text{Total}}$

smaller  $SS_{\text{level}}$  → single mean model

$$\beta_i = \bar{y}_{i,\cdot} - \bar{y}_{\cdot,\cdot}$$

(between groups) explained variance by the group/level

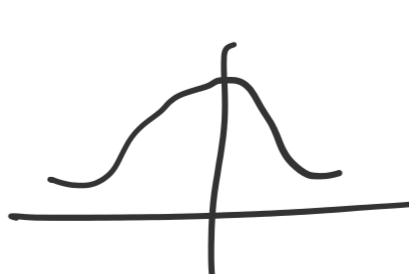

(within group)  $e_{ij}$  → unexplained by the grouping

$$MS_{\text{level}} = \frac{SS_{\text{level}}}{g-1} \quad MS_{\text{error}} = \frac{SS_{\text{error}}}{N-g}$$

$$f_0 = \frac{MS_{\text{level}}}{MS_{\text{error}}} \quad N = \sum_i^g n_i$$

$f_0 > 1$  multiple mean model

$f_0 \approx 1$  variation in each group = var from random sampling (single mean model)

F-distribution (  →  )

Normal chi<sup>2</sup>

$$f = \frac{\text{var}(\text{chi}^2)}{\text{var}(\text{chi}^2)}$$

$$Pr(f > f_0, f \approx 1) = p > 0.05$$

↑ ↑  
 How extreme the value  $f_0$  is under the Null model      single mean model Null