Lecture 31 Analysis of Variance (ANOVA)

BIO210 Biostatistics

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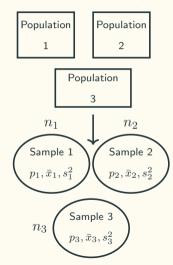
Fall, 2023

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Compare More Than Two Means

More than two-samples



Intuitive way: compare all possible pairs using two-sample independent t test:

Samples 1 vs 2: $H_0: \mu_1 = \mu_2; \ H_1: \mu_1 \neq \mu_2$

Samples 1 vs 3: $H_0: \mu_1 = \mu_3; \ H_1: \mu_1 \neq \mu_3$

Samples 2 vs 3: $H_0: \mu_2 = \mu_3; \ H_1: \mu_2 \neq \mu_3$

Good enough?

Compare More Than Two Means

What if we have 15 samples from 15 different populations?

- Intuitive way: compare all possible pairs using two-sample independent t test:
- Number of comparisons: $\binom{15}{2} = \frac{15 \times 14}{2} = 105$
- Significance level: $\alpha = 0.05$
- When we set $\alpha=0.05$, we want to tolerate a 5% of chance of making a type I error. That is, the intended number of tests of making a type I error: ≈ 5
- Assume that the means are all the same, what is the probability of making a type I error in at least one test?

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\mathbb{P}\left(\text{reject }H_0 \text{ in at least one test} \,|\, H_0 \text{ is true}\right) =1-\mathbb{P}\left(\text{not rejecting }H_0 \text{ in all tests} \,|\, H_0 \text{ is true}\right) =1-0.95^{105} =0.995
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Source of Variation - Total

Sample 1	Sample 2	Sample 3	
3	5	5	
2	3	6	
1	4	7	
$\bar{x}_1 = 2$	$\bar{x}_2 = 4$	$\bar{x}_3 = 6$	

sum of squares (SS): add up the squared distance between an observation and the mean:

$$\sum (X - \bar{X})^2$$

SST: total sum of squares

The grand mean:
$$\bar{x} = \frac{3+2+1+5+3+4+5+6+7}{9} = 4$$

SST = $(3-4)^2 + (2-4)^2 + (1-4)^2 + (5-4)^2 + (3-4)^2 + (4-4)^2 + (5-4)^2 + (6-4)^2 + (7-4)^2 = 30$

What is the *df* ?
$$df_T = 9 - 1 = 8$$

Source of Variation - Within Groups

Sample 1	Sample 2	Sample 3	
3	5	5	
2	3	6	
1	4	7	
$\bar{x}_1 = 2$	$\bar{x}_2 = 4$	$\bar{x}_3 = 6$	

sum of squares (SS): add up the squared distance between an observation and the mean:

$$\sum (X - \bar{X})^2$$

SSW: sum of squares within

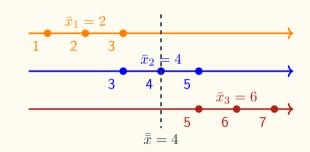
SSW =
$$(3-2)^2 + (2-2)^2 + (1-2)^2 + (5-4)^2 + (3-4)^2 + (4-4)^2 + (5-6)^2 + (6-6)^2 + (7-6)^2$$

= $df_1 \cdot s_1^2 + df_2 \cdot s_2^2 + df_3 \cdot s_3^2$
= 6

What is the
$$df$$
 ? $df_W = (3-1) + (3-1) + (3-1) = 6$

Source of Variation - Between Groups

Sample 1	Sample 2	Sample 3	
3	5	5	
2	3	6	
1	4	7	
$\bar{x}_1 = 2$	$\bar{x}_2 = 4$	$\bar{x}_3 = 6$	



SSB: sum of squares between

$$SSB = (2-4)^{2} + (2-4)^{2} + (2-4)^{2} + (4-4)^{2} + (4-4)^{2} + (4-4)^{2} + (4-4)^{2} + (6-4)^{2} + (6-4)^{2} + (6-4)^{2} + (6-4)^{2}$$

$$= n_{1} \cdot (\bar{x}_{1} - \bar{x})^{2} + n_{2} \cdot (\bar{x}_{2} - \bar{x})^{2} + n_{3} \cdot (\bar{x}_{3} - \bar{x})^{2}$$

$$= 24$$

What is the df ? $df_B = 3 - 1 = 2$

Summary of The Source of Variation

Sample 1	Sample 2	Sample 3	
3	5	5	
2	3	6	
1	4	7	
$\bar{x}_1 = 2$	$\bar{x}_2 = 4$	$\bar{x}_3 = 6$	

Source of Variation	SS (sum of squares)	df	MS (mean square)
Between	24	2	12
Within	6	6	1
Total	30	8	

Variance-like

Multiple Samples From Multiple Populations

Population 1	Sample 1 (n_1, \bar{x}_1, s_1^2) Sample 2 (n_2, \bar{x}_2, s_2^2)
Population 2	Sample 2 $(n_2, ar{x}_2, s_2^2)$
Population 3	Sample 3 (n_3, \bar{x}_3, s_3^2)
:	:
Population k	Sample k (n_k, \bar{x}_k, s_k^2)

$$\bar{\bar{x}} = \frac{\sum_{i=1}^{k} \sum_{j=1}^{n_i} x_{ij}}{\sum_{i=1}^{k} n_i} = \frac{\sum_{i=1}^{k} n_i \bar{x}_i}{n}$$

The ANOVA Table

Source of Variation	SS	df	MS
Between	$SSB = \sum_{i=1}^{k} n_i (\bar{x}_i - \bar{\bar{x}})^2$	k-1	$MSB = \frac{SSB}{k-1}$
Within	SSW = $\sum_{i=1}^{k} \sum_{j=1}^{n_i} (x_{ij} - \bar{x}_i)^2 = \sum_{i=1}^{k} df_i s_i^2$	n-k	$MSW = \frac{SSW}{n-k}$
Total	$SST = \sum_{i=1}^{k} \sum_{j=1}^{n_i} (x_{ij} - \bar{\bar{x}})^2 = SSB + SSW$	n-1	

The F-test

$$\begin{cases} H_0: & \mu_1=\mu_2=\dots=\mu_k\\ H_1: & \text{not all equal} \end{cases} \Leftrightarrow \begin{cases} H_0: & \text{The main variation is from SSW}\\ H_1: & \text{The main variation is from SSB} \end{cases}$$

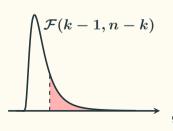
Under the null hypothesis:

$${
m \frac{SSB}{\sigma^2}}\sim \chi^2(k-1)$$
 and ${
m \frac{SSW}{\sigma^2}}\sim \chi^2(n-k)$, where σ^2 is the common variance

The test statistic:

$$F = \frac{\frac{\text{SSB}}{(k-1)\sigma^2}}{\frac{\text{SSW}}{(n-k)\sigma^2}} = \frac{\text{MSB}}{\text{MSW}} \sim \mathcal{F}(k-1,n-k)$$

$$p\text{-value: } \mathbb{P}\left(\text{data} \mid H_0 \text{ is true}\right) = \mathbb{P}\left(F_{k-1,n-k} \geqslant \frac{\text{MSB}}{\text{MSW}}\right)$$



Summary of an ANOVA result

Source of Variation	SS	df	MS	F	$p ext{-}value$
Between	$SSB = \sum_{i=1}^{k} n_i (\bar{x}_i - \bar{\bar{x}})^2$	k-1	$MSB = \frac{SSB}{k-1}$		
Within	$SSW = \sum_{i=1}^{k} df_i s_i^2$	n-k	$MSW = \frac{SSW}{n-k}$	$\frac{\text{MSB}}{\text{MSW}}$	$\mathbb{P}\left(F \geqslant \frac{\text{MSB}}{\text{MSW}}\right)$
Total	SST = SSB + SSW	n-1			