

Development Statistics

S10 ANOVA

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Distribution of sample mean

- The population follows a normal distribution

$$x_i \sim N(\mu, \sigma^2)$$

- Whether σ^2 is known or unknown

$$\bar{X} \sim N\left(\mu, \frac{\sigma^2}{n}\right)$$

Normalization

- Normalization of normal distribution

$$\bar{X} \sim N\left(\mu, \frac{\sigma^2}{n}\right) \rightarrow z = \frac{\bar{X} - \mu}{\sqrt{\frac{\sigma^2}{n}}} \sim N(0,1)$$

ANOVA

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Definition of Chi-square distribution

- The population is $N(0,1)$
- Take sample whose size is n

$$x_i \sim N(0,1)$$

- The squared sum of n samples follows Chi-square distribution with degree of freedom $n : \chi^2(n)$

$$C = \sum_{i=1}^n x_i^2 \sim \chi^2(n)$$

ANOVA

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Sample variance

- The population is $N(\mu, \sigma^2)$
- Take sample whose size is n

**Unbiased sample
Variance**

$$s^2 = \frac{\sum_i (x_i - \bar{x})^2}{n-1}$$

ANOVA

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**Practically speaking,
Chi square is ...**

$$\begin{aligned}(n-1)s^2 &= \sum_i (x_i - \bar{x})^2 \\&= \sum_i [(x_i - \mu) - (\bar{x} - \mu)]^2 \\&= \sum_i (x_i - \mu)^2 - 2(\bar{x} - \mu) \sum_i (x_i - \mu) + n(\bar{x} - \mu)^2 \\&= \sum_i (x_i - \mu)^2 - \frac{(\bar{x} - \mu)^2}{1/n}\end{aligned}$$

**Chi square
distribution**

$$C = (n-1) \frac{s^2}{\sigma^2} \sim \chi^2(n-1)$$

ANOVA

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Definition of T distribution

Standard normal Chi square (n)

$$z \sim N(0,1) \quad C \sim \chi^2(n)$$



$$t = \frac{Z}{\sqrt{\frac{C}{n}}} \sim t(n) \quad \mathbf{T (n)}$$

ANOVA

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Practically speaking,

Standard normal Chi square (n-1)

$$z = \frac{\bar{X} - \mu}{\sqrt{\frac{\sigma^2}{n}}} \sim N(0,1) \quad C = (n-1) \frac{s^2}{\sigma^2} \sim \chi^2(n-1)$$



$$t = \frac{\bar{X} - \mu}{\sqrt{\frac{s^2}{n}}} \sim t(n-1) \quad \mathbf{T (n-1)}$$

ANOVA

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F distribution

- Two independent Chi distributions
- The ratio follows F distribution

$$\underbrace{C_1 \sim \chi^2(n_1) \quad C_2 \sim \chi^2(n_2)}_{\frac{C_1 / n_1}{C_2 / n_2} \sim F(n_1, n_2)}$$

ANOVA

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Difference means among groups

- When the number of the groups = 2
 - Test of the difference of means of two populations
 - Based on T-distribution
- When the number of the groups ≥ 3
 - Analysis of Variance = ANOVA
 - Based on F-distribution

ANOVA

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Data

group 1	12.2	18.8	18.2
group 2	22.2	20.5	14.6
group 3	20.8	19.5	26.3
group 4	26.4	32.6	31.1
group 5	24.5	21.2	22.4

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Test of difference of the means of two populations

- Assuming the population variance is the same

$$z = \frac{(\bar{X}_1 - \bar{X}_2) - (\mu_1 - \mu_2)}{\sqrt{s^2 \left(\frac{1}{n_1} + \frac{1}{n_2} \right)}} \sim t(n_1 + n_2 - 2)$$

$$S^2 = \frac{\sum_{i=1}^{n_1} (X_{1i} - \bar{X}_1)^2 + \sum_{i=1}^{n_2} (X_{2i} - \bar{X}_2)^2}{n_1 + n_2 - 2}$$

- You can do a statistical test based on t-distribution

ANOVA

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Use Excel sheet !

● T-statistics

	2	3	4	5
1	-0.865	-1.957	-4.842	-2.719
2		-0.998	-3.687	-1.442
3			-2.799	-0.218
4				3.489

ANOVA

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Assumption of ANOVA

	distribution
1	$N(\mu + \alpha_1, \sigma^2)$
2	$N(\mu + \alpha_2, \sigma^2)$
3	$N(\mu + \alpha_3, \sigma^2)$
4	$N(\mu + \alpha_4, \sigma^2)$
5	$N(\mu + \alpha_5, \sigma^2)$

ANOVA

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Mean square Decomposition of total variation

- Total variation
= Intra-group variation + Inter-group variation

$$S_T = S_{\text{intra}} + S_{\text{inter}}$$

$$\begin{aligned} & \sum_i \sum_j (x_{ij} - \bar{x})^2 \\ &= \sum_i \sum_j (x_{ij} - \bar{x}_i)^2 + \sum_i n_i (\bar{x}_i - \bar{x})^2 \end{aligned}$$

ANOVA

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If $\alpha_A = \alpha_B = \alpha_C = \alpha_D = \alpha_E$

- Then, the population distribution is same

$$x_{ij} \sim N(\mu, \sigma^2)$$

- Therefore (see slide No.6)

$$\sum_j (x_{ij} - \bar{x}_i)^2 / \sigma^2 \sim \chi^2(2)$$

- Therefore

$$\sum_i \sum_j (x_{ij} - \bar{x}_i)^2 / \sigma^2 \sim \chi^2(10)$$

ANOVA

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If $\alpha_A = \alpha_B = \alpha_C = \alpha_D = \alpha_E$

- Distribution of mean of each group

$$\bar{x}_i \sim N(\mu, \sigma^2 / 3)$$

- Therefore (see slide No.6)

$$\frac{\sum_i (\bar{x}_i - \bar{x})^2}{\sigma^2 / 3} \sim \chi^2(4)$$

ANOVA

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Then

If $\alpha_A = \alpha_B = \alpha_C = \alpha_D = \alpha_E$

$$\frac{\frac{\sum_i (\bar{x}_i - \bar{x})^2 / 4}{\sigma^2 / 3}}{\frac{\sum_i \sum_j (x_{ij} - \bar{x}_i)^2 / 10}{\sigma^2}} = \frac{\sum_i 3(\bar{x}_i - \bar{x})^2 / 4}{\sum_i \sum_j (x_{ij} - \bar{x}_i)^2 / 10} \sim F(4, 10)$$

ANOVA

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One-way ANOVA

Factor	Square sum	Degree of freedom	Mean square	F statistic
Inter-group	S_{intra}	Group -1	V_{intra}	$\frac{V_{\text{intra}}}{V_{\text{Inter}}}$
Intra-group	S_{Inter}	N - group	V_{Inter}	

ANOVA

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One-way ANOVA

Factor	Square sum	Degree of freedom	Mean square	F statistic
Inter-group	314.391	4	78.598	7.079
Intra-group	111.027	10	11.103	

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