

1 Introduction to Analysis of Variance

1.1 Using multiple pairwise comparisons to test multi-sample hypothesis → INVALID

- inflated risk of type I error:  $P(\text{error}) = 1 - (1 - \alpha)^z$  where  $z = \text{no. of pairwise comparisons}$

1.2 Single Factor ANOVA Hypotheses:

$H_0: \mu_1 = \mu_2 = \dots = \mu_k$  ;  $k = \# \text{ experimental groups (samples)}$

$H_A: \text{group means are not all equal}$

1.3 Notation

1.3.1 individual measurements:  $X_{ij} = j^{\text{th}}$  measurement in  $i^{\text{th}}$  group  
 1<sup>st</sup> subscript = group #      2<sup>nd</sup> subscript = individual w/in group

1.3.2 sample sizes:  $n_i = \# \text{ in group } i$  (sample size of  $i^{\text{th}}$  group);  
 $N = \sum n_i \rightarrow \text{total \# in expt}$

1.3.3 means: 2 kinds of means: group means (group  $i$ ):  $\bar{X}_i$  & grand mean:  $\bar{X}$

1.3.4 Summation:  $\sum_i X_i$        $\sum_i \sum_j X_{ij} = \sum_{i=1}^k \left( \sum_{j=1}^{n_i} X_{ij} \right)$

1.4 Linear model for single-factor ANOVA:

$$X_{ij} = \mu + \tau_i + \varepsilon_{ij}$$

where:

$\mu$  is the grand mean

$\tau_i$  is the difference betw/  $\mu$  and  $i^{\text{th}}$  group mean

$\varepsilon_{ij}$  is random error within  $i^{\text{th}}$  group

assumptions: (1)  $\bar{\varepsilon}_{ij} = 0$  i.e., no bias w/in sample

(2)  $\text{var}(\varepsilon_{ij}) = \sigma^2$  i.e., equal variances among groups

(3)  $\varepsilon_{ij} \sim \text{normal}$

- reject  $H_0$  if variation in  $X_{ij}$  influenced more by  $\tau_i$  than by  $\varepsilon_{ij}$
- do not reject  $H_0$  if variation in  $X_{ij}$  influenced more by  $\varepsilon_{ij}$  than by  $\tau_i$

2 ANOVA Calculations

2.1 Partitioning variability & Sources of variation

- total variability
- within group variability (“error”)
- among group variability

Deviation of each datum from grand mean partitioned into:

(1) deviation from group mean

(2) deviation of group mean from grand mean

$$(X_{ij} - \bar{X}) = (X_{ij} - \bar{X}_i) + (\bar{X}_i - \bar{X})$$

2.2 Sums of squares & degrees of freedom:

total:      total SS =  $\sum_{i=1}^k \sum_{j=1}^{n_i} (X_{ij} - \bar{X})^2$

total DF =  $N - 1$

among-groups: among-groups SS =  $\sum_{i=1}^k n_i (\bar{X}_i - \bar{X})^2$

among-groups DF =  $k - 1$

w/in-groups: within-groups SS =  $\sum_{i=1}^k \left[ \sum_{j=1}^{n_i} (X_{ij} - \bar{X}_i)^2 \right]$

w/in-groups DF = total DF – groups DF =  $N - k$

**2 ANOVA Calculations (continued)**

2.3 *SS & DF* are additive: total *SS* = groups *SS* + error *SS*  
 total *DF* = groups *DF* + error *DF*

2.4 Mean Squared Deviation from mean: divide *SS* by *DF*

$$\text{groups MS} = \frac{\text{groups SS}}{\text{groups DF}} \qquad \text{error MS} = \frac{\text{error SS}}{\text{error DF}}$$

2.5 Summary of Calculations:

Source	Sum of squares ( <i>SS</i> )	Degrees of freedom ( <i>DF</i> )	Mean Square ( <i>MS</i> )
Total	$\sum_{i=1}^k \sum_{j=1}^{n_i} (X_{ij} - \bar{X})^2$	$N - 1$	
among groups	$\sum_{i=1}^k n_i (\bar{X}_i - \bar{X})^2$	$k - 1$	$\frac{\text{groups SS}}{\text{groups DF}}$
within groups	$\sum_{i=1}^k \left[ \sum_{j=1}^{n_i} (X_{ij} - \bar{X}_i)^2 \right]$	$N - k$	$\frac{\text{error SS}}{\text{error DF}}$

- Summary of calculations w/ machine formulas – Zar, Table 10.2

**3 Interpreting ANOVA**

3.1 ANOVA Hypothesis Test

3.1.1 If  $H_0$  true, then groups MS & error MS each estimate common pop var ( $\sigma^2$ )  
 If pop means  $\neq$ , then groups MS > error MS

∴ test for equality of means using variance ratio test w/ groups MS in numerator

$$F = \frac{\text{groups MS}}{\text{error MS}}$$

3.1.2 Critical value =  $F_{\alpha(1),(k-1),(N-k)}$

3.1.3 If  $F_{\text{calc}} \geq F_{\text{crit}}$ , then reject  $H_0 \Rightarrow$  all  $k$  pop means not equal

→ to determine which means differ, use multiple comparisons test (e.g., Tukey test)

3.2 ANOVA with Means & Variances only (not original data)

3.2.1 First, determine  $SS_i$

$$s_i^2 = (s_i)^2 = n_i (s_{\bar{X}_i})^2 = \frac{SS_i}{n_i - 1}$$

3.2.2 Next, calculate error *SS* & groups *SS*

$$\text{error SS} = \sum_{i=1}^k SS_i = \sum_{i=1}^k (n_i - 1) s_i^2$$

$$\text{groups SS} = \sum_{i=1}^k n_i \bar{X}_i^2 - \frac{\left( \sum_{i=1}^k n_i \bar{X}_i \right)^2}{\sum_{i=1}^k n_i}$$

3.2.3 Complete ANOVA using groups *SS* & error *SS* (& groups *DF* & error *DF*)

### 3 Interpreting ANOVA (continued)

#### 3.3 Two types of ANOVA

3.3.1 Fixed effects model (model I): Levels of factor (treatments) specifically chosen

Hypotheses:  $H_0: \mu_1 = \mu_2 = \dots = \mu_k$

3.3.2 Random effects model (model II): Levels of factor selected @ random

Null hypothesis:  $H_0$ : no difference in measurements among groups  
– calculations identical to fixed effects model

3.3.3 Distinction very important for ANOVA involving  $\geq 2$  factors

#### 3.4 Assumptions

##### 3.4.1 Equal sample variances

- ANOVA fairly robust to violations, provided all  $n_i$  nearly equal
- if not, then different P(type I error):
  - if larger variances w/ larger samples,  $P(\text{type I error}) < \alpha$ ;
  - if larger variances w/ smaller samples,  $P(\text{type I error}) > \alpha$ ;

##### 3.4.2 Samples from normal populations

- ANOVA fairly robust to violations (both skew & kurtosis)

### 4 Confidence Limits for Population Means

4.1 Confidence Interval calculation similar to two sample case

$1-\alpha$  confidence limits for  $\mu_i$  ( $\mu + \tau_i$ ):  $\bar{X}_i \pm t_{\alpha(2), \nu} \sqrt{\frac{s^2}{n_i}}$

– where  $s^2$  is error *MS* &  $\nu$  is error *DF* ( $= N-k$ )

4.2 CI calculation appropriate only when know that  $\mu_i$  different than other  $\mu$ 's  
=> first must perform multiple comparison test (e.g., Tukey test)