

I. Nonparametric ANOVA – Kruskal-Wallis test (\cong ANOVA by ranks)

- A. Application:
 - whenever parametric single-factor ANOVA appropriate,
 - when parametric assumptions violated: non-normal populations; \neq variances
- B. Assumption: sampled populations have same dispersions & shapes
- C. Hypotheses: H_0 : measurements the same in all k populations
 H_A : measurements not the same in all k populations
- D. Procedure: Rank data – same as with Mann-Whitney two-sample test
- E. Test statistic:

$$H = \frac{12}{N(N+1)} \sum_{i=1}^k \frac{R_i^2}{n_i} - 3(N+1) \quad n_i = \# \text{ in group } i; \quad R_i = \text{sum ranks in } i$$

- F. Compare w/ critical values, Table B.13
- G. If $k > 5$ or n_i large (> 6 or 7), approximate critical values of H: χ^2 w/ $k-1$ DF
- H. If tied ranks, H an underestimate. Correction factor:

$$C = 1 - \frac{\sum t}{N^3 - N}$$

$$H_c = H/C$$

$$\sum t = \sum_{i=1}^m (t_i^3 - t_i) \quad t_i = \# \text{ ties in } i^{\text{th}} \text{ group}, \quad m = \# \text{ groups of tied ranks}$$

$H_c \approx H$ when t_i 's $\ll N$

II. Test for Homogeneity of Variances – Bartlett’s test

- A. Hypotheses:
 - $H_0: \sigma_1^2 = \sigma_2^2 = \dots = \sigma_k^2$
 - H_A : variances are not all equal

B. Test Statistic:

$$B = (\ln s_p^2) \left(\sum_{i=1}^k v_i \right) - \sum_{i=1}^k v_i \ln s_i^2 \quad v_i = n_i - 1, \quad s_p^2 = \text{pooled variance} = \sum SS_i / \sum v_i$$

- C. $B \approx$ Chi-squared; critical value: $\chi_{\alpha, k-1}^2$, where k = number of samples

E. More accurate w/ correction factor:

$$C = 1 + \frac{1}{3(k-1)} \left(\sum \frac{1}{v_i} - \frac{1}{\sum v_i} \right)$$

Corrected statistic: $B_c = B/C$ Compare B_c with $\chi_{\alpha, k-1}^2$

- F. Test badly affected by deviations from normality
 Not recommended to test assumptions prior to doing single-factor ANOVA,
 – esp. since ANOVA relatively robust to unequal variances.