# **Triola Statistics Series Review**



## **Basic Terms**

**Statistics:** Methods for planning experiments, obtaining data, organizing, summarizing, analyzing, interpreting, and drawing conclusions based on data. **Population:** Collection of *all* elements to be studied.

**Census:** Data from *every* member of a population. **Sample:** Subcollection of members from a population.

**Parameter:** Numerical measurement of characteristic of a *population*.

**Statistic:** Numerical measurement of characteristic of a *sample*.

**Random Sample:** Every member of population has same chance of being selected.

**Simple Random Sample:** Every sample of same size *n* has the same chance of being selected.

## **Describing, Exploring, and Comparing Data**

#### **Measures of Center:**

Population mean:  $\mu$ 

Sample mean:  $\bar{x} = \frac{\sum x}{n}$ 

Mean from frequency dist.:

$$
\bar{x} = \frac{\sum (f \cdot x)}{n}
$$

Median: Middle value of data arranged in order. Mode: Most frequent data value(s).

 $\overline{2}$ 

 $Midrange:$   $\frac{maximum + minimum}{2}$ 

#### **Measures of Variation:**

Range: maximum - minimum Sample standard deviation:

$$
s = \sqrt{\frac{\sum (x - \bar{x})^2}{n - 1}} \text{ or } \sqrt{\frac{n(\sum x^2) - (\sum x)^2}{n(n - 1)}}
$$

St. dev. from frequency dist.:

$$
s = \sqrt{\frac{n[\sum (f \cdot x^2)] - [\sum (f \cdot x)]^2}{n(n-1)}}
$$

Sample variance: *s* 2

Population st. dev.:

$$
\sigma = \sqrt{\frac{\sum (x - \mu)^2}{N}}
$$

Population variance:  $\sigma^2$ 

**Distribution:** Explore using frequency distribution, histogram, dotplot, stemplot, boxplot.

**Outlier:** Value far away from almost all other values. **Time:** Consider effects of changes in data over time.

(Use time-series graphs, control charts.)

## **Probability**

**Rare Event Rule:** If, under a given assumption, the probability of a particular observed event is very small and the observed event occurs significantly less than or significantly greater than what we typically expect with that assumption, conclude that the assumption is probably not correct.

**Relative Frequency:**

 $P(A) = \frac{\text{number of times } A \text{ occurred}}{\text{number of trials}}$ 

**Classical Approach:**

 $P(A) = \frac{s}{n}$  (equally likely outcomes)

**Probability property:**  $0 \leq P(A) \leq 1$ 

**Complement of Event** *A***:**

 $P(\bar{A}) = 1 - P(A)$ 

**Addition Rule:**

**Disjoint Events:** Cannot occur together. If *A*, *B* are disjoint:  $P(A \text{ or } B) = P(A) + P(B)$ If *A*, *B* are not disjoint:  $P(A \text{ or } B) = P(A) + P(B) - P(A \text{ and } B)$ 

#### **Multiplication Rule:**

**Independent Events:** No event affects probability of other event.

If *A*, *B* are independent:

$$
P(A \text{ and } B) = P(A) \cdot P(B)
$$

If *A, B* are dependent:

$$
P(A \text{ and } B) = P(A) \cdot P(B \mid A)
$$

where  $P(B|A)$  is  $P(B)$  assuming that event *A* has already occurred.

## **Counting**

**Multiplication Counting Rule:** If an event can occur *m* ways and a second event can occur *n* ways, together they can occur  $m \cdot n$  ways.

**Factorial Rule:** *n* different items can be arranged *n*! different ways.

**Permutations** (order counts) of *r* items selected from *n* different items:

$$
{}_{n}P_{r}=\frac{n!}{(n-r)!}
$$

**Permutations** when some items are identical to others:

$$
\frac{n!}{n_1!n_2!\ldots n_k!}
$$

**Combinations** (order doesn't count) of *r* items selected from *n* different items:

$$
{}_{n}C_{r}=\frac{n!}{(n-r)!r!}
$$

## **Random Variables**

**Random Variable:** Variable that has a single numerical value, determined by chance, for each outcome.

**Probability Distribution:** Graph, table, or formula that gives the probability for each value of the random variable.

**Requirements** of random variable:

$$
1. \quad \Sigma P(x) = 1
$$

2.  $0 \le P(x) \le 1$ 

**Parameters** of random variable:

$$
\mu = \sum [x \cdot P(x)]
$$

$$
\sigma = \sqrt{\sum [x^2 \cdot P(x)] - \mu^2}
$$

**Expected value:**  $E = \sum [x \cdot P(x)]$ 

**Binomial Distribution:** Requires fixed number of independent trials with all outcomes in two categories, and constant probability.

*n*: Fixed number of trials

 $\mu = np$ 

- *x*: Number of successes in *n* trials
- *p*: Probability of success in one trial
- *q*: Probability of failure in one trial
- *P*( $x$ ): Probability of  $x$  successes in  $n$  trials

$$
P(x) = \frac{n!}{(n-x)!x!} \cdot p^x \cdot q^{n-x}
$$

 $\sigma = \sqrt{npq}$  St. dev. (binomial)

**Poisson Distribution:** Discrete probability distribution that applies to occurrences of some event *over a specified interval*.

$$
P(x) = \frac{\mu^x \cdot e^{-\mu}}{x!}
$$
 where  $e \approx 2.71828$ 



## **normal Distribution**

Continuous random variable having bell-shaped and symmetric graph and defined by specific equation.

#### **Standard Normal Distribution:** Normal distribution with

$$
\mu = 0 \text{ and } \sigma = 1.
$$
  
Standard z score:  $z = \frac{x - \mu}{\sigma}$ 

#### **Central Limit Theorem:**

As sample size increases, sample means  $\bar{x}$  approach *normal* distribution;

 $\sigma$ 

$$
\mu_{\overline{x}} = \mu \text{ and } \sigma_{\overline{x}} = \frac{\sigma}{\sqrt{n}}
$$

$$
\overline{x} - \mu_{\overline{x}}
$$

so that  $z =$  $\sigma$  $\sqrt{n}$ 

**normal approximation to Binomial:**

Requires  $np \ge 5$  and  $nq \ge 5$ . Use  $\mu = np$  and  $\sigma = \sqrt{npq}$ .

## **Determining Sample Size**

#### **Proportion:**

$$
n = \frac{\left[z_{\alpha/2}\right]^2 \cdot 0.25}{E^2}
$$

$$
n = \frac{\left[z_{\alpha/2}\right]^2 \hat{p}\hat{q}}{E^2} \quad (\hat{p} \text{ and } \hat{q} \text{ known})
$$
  
Mean: 
$$
n = \left[\frac{z_{\alpha/2}\sigma}{E}\right]^2
$$

## **Confidence Intervals (Using One Sample)**

**Proportion:**  $\hat{p} - E < p < \hat{p} + E$ 

where 
$$
E = z_{\alpha/2} \sqrt{\frac{\hat{p}\hat{q}}{n}}
$$
 and  $\hat{p} = \frac{x}{n}$ 

**Mean:**  $\bar{x} - E < \mu < \bar{x} + E$ 

$$
E = t_{\alpha/2} \cdot \frac{s}{\sqrt{n}} \left( \sigma \text{ not known} \right)
$$

**Standard Deviation:**

$$
\sqrt{\frac{(n-1)s^2}{\chi^2_R}} < \sigma < \sqrt{\frac{(n-1)s^2}{\chi^2_L}}
$$

#### **Confidence Intervals (Using two Samples)**

#### **two Proportions:**

$$
(\hat{p}_1 - \hat{p}_2) - E < (p_1 - p_2) < (\hat{p}_1 - \hat{p}_2) + E
$$
\n
$$
\text{where } E = \frac{z_{\alpha/2}}{\sqrt{\frac{\hat{p}_1 \hat{q}_1}{n_1} + \frac{\hat{p}_2 \hat{q}_2}{n_2}}}
$$

#### **two Means (Independent):**

$$
(\bar{x}_1 - \bar{x}_2) - E < (\mu_1 - \mu_2) < (\bar{x}_1 - \bar{x}_2) + E
$$
  
\nwhere  $E = t_{\alpha/2} \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}$   
\n $df = \text{smaller of } n_1 - 1 \text{ and } n_2 - 1.$ 

#### **Alternative Cases for Two Independent Means:**

If  $\sigma_1$ ,  $\sigma_2$  unknown but assumed equal, use pooled variance  $s_p^2$ :

$$
E = t_{\alpha/2} \sqrt{\frac{s_p^2}{n_1} + \frac{s_p^2}{n_2}}
$$

whe

$$
s_p^2 = \frac{(n_1 - 1)s_1^2 + (n_2 - 1)s_2^2}{(n_1 - 1) + (n_2 - 1)}
$$
  
wn  $\sigma_1$  and  $\sigma_2$ :  $E = z_{\alpha/2} \sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}$ 

$$
\text{Known } \sigma_1 \text{ and } \sigma_2: E = z_{\alpha/2} \sqrt{\frac{1}{n_1} + \frac{z^2}{n_2}}
$$
\n
$$
\text{Matched Pairs:}
$$

$$
\overline{d} - E < \mu_d < \overline{d} + E
$$
\n
$$
\text{where } E = t_{\alpha/2} \frac{s_d}{\sqrt{n}} \text{ and } df = n - 1
$$

## **Hypothesis testing**

**Hypothesis Test:** Procedure for testing claim about a population characteristic.

**Null Hypothesis**  $H_0$ **: Statement that value of popu**lation parameter is *equal to* some claimed value.

**alternative Hypothesis** *H***1:** Statement that population parameter has a value that somehow differs from value in the null hypothesis.

**Critical Region:** All values of test statistic leading to rejection of null hypothesis.

 $\alpha$  = **Significance Level:** Probability that test statistic falls in critical region, assuming null hypothesis is true.

**Type I Error:** Rejecting null hypothesis when it is true. Probability of type I error is significance level  $\alpha$ .

**Type II Error:** Failing to reject null hypothesis when it is false. Probability of type II error is denoted by  $\beta$ .

**Power of test:** Probability of rejecting a false null hypothesis.

#### **Procedure**

- 1. Identify original claim, then state null hypothesis (with equality) and alternative hypothesis (without equality).
- 2. Select significance level  $\alpha$ .
- 3. Evaluate test statistic.
- 4. Proceed with critical value method or *P*-value method:

## **Common Critical** *z* **Values**

#### **COnfIDenCe InterVal**



#### **HyPOtHeSIS teSt: rIgHt-taIleD**



#### **HyPOtHeSIS teSt: left-taIleD**



#### **HyPOtHeSIS teSt: twO-taIleD**



#### **Critical Value Method of Testing Hypotheses:**

Uses decision criterion of rejecting null hypothesis only if test statistic falls within critical region bounded by critical value.

**Critical Value:** Any value separating critical region from values of test statistic that do not lead to rejection of null hypothesis.

*P***-value Method of Testing Hypotheses:** Uses decision criterion of rejecting null hypothesis only if *P*-value  $\leq \alpha$  (where  $\alpha$  = significance level).

*P***-value:** Probability of getting value of test statistic *at least as extreme* as the one found from sample data, assuming that null hypothesis is true.

**Left-Tailed Test:** *P*-value = area to *left* of test statistic **Right-Tailed Test:** *P*-value = area to *right* of test statistic

**Two-Tailed Test:** *P*-value = *twice* the area in tail beyond test statistic

## **Choosing Between** *t* **and** *z* **for Inferences about Mean**

- $\sigma$  unknown and normally distributed population: use *t*
- $\sigma$  unknown and  $n > 30$ : use *t*
- $\sigma$  known and normally distributed population: use *z*
- $\sigma$  known and  $n > 30$ ; use  $\tau$

If none of the above apply, use nonparametric method or bootstrapping.

## **wording of Conclusion**



## **Hypothesis Testing (One Sample)**

**One Proportion:** Requires simple random sample,  $np \geq 5$  and  $nq \geq 5$ , and conditions for binomial distribution.

Test statistic: 
$$
z = \frac{\hat{p} - p}{\sqrt{\frac{pq}{n}}}
$$
 where  $\hat{p} = \frac{x}{n}$ 

**One Mean:** Requires simple random sample and either  $n > 30$  or normally distributed population.

Test statistic

$$
t = \frac{\bar{x} - \mu_{\bar{x}}}{\frac{s}{\sqrt{n}}}
$$
 (for  $\sigma$  not known)  
where df =  $n - 1$ 

**One Standard Dev. or Variance:** Requires simple random sample and normally distributed population.

Test statistic: 
$$
\chi^2 = \frac{(n-1)s^2}{\sigma^2}
$$
  
where df = n - 1

## **Hypothesis Testing (Two Proportions or Two Independent Means)**

**Two Proportions:** Requires two independent simple random samples and  $np \geq 5$  and  $nq \geq 5$  for each. Test statistic:

$$
z = \frac{(\hat{p}_1 - \hat{p}_2) - (p_1 - p_2)}{\sqrt{\frac{\overline{p}\overline{q}}{n_1} + \frac{\overline{p}\overline{q}}{n_2}}}
$$
  
where  $\overline{p} = \frac{x_1 + x_2}{n_1 + n_2}$  and  $\overline{q} = 1 - \overline{p}$   
and  $\hat{p}_1 = \frac{x_1}{n_1}$  and  $\hat{p}_2 = \frac{x_2}{n_2}$ 

#### **Two Means (independent samples):**

Requires two independent simple random samples with both populations normally distributed or  $n_1 > 30$  and  $n_2 > 30$ .

The population standard deviations  $\sigma_1$  and  $\sigma_2$  are usually unknown.

Recommendation: do not assume that  $\sigma_1 = \sigma_2$ .

Test statistic (unknown  $\sigma_1$  and  $\sigma_2$  and not assuming  $\sigma_1 = \sigma_2$ :

$$
t = \frac{(\bar{x}_1 - \bar{x}_2) - (\mu_1 - \mu_2)}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}}
$$

df = smaller of  $n_1 - 1$  and  $n_2 - 1$ 

#### **Hypothesis Testing (Alternative Cases for Two Means with Independent Samples)**

Requires two independent simple random samples and either of these two conditions:

Both populations normally distributed or  $n_1 > 30$ and  $\overline{n_2} > 30$ .

#### Alternative case when  $\sigma_1$  and  $\sigma_2$  are not **known, but it is assumed that**  $\sigma_1 = \sigma_2$ **:**

Pool variances and use test statistic

$$
t = \frac{(\bar{x}_1 - \bar{x}_2) - (\mu_1 - \mu_2)}{\sqrt{\frac{s_p^2}{n_1} + \frac{s_p^2}{n_2}}}
$$

where

$$
s_p^2 = \frac{(n_1 - 1)s_1^2 + (n_2 - 1)s_2^2}{(n_1 - 1) + (n_2 - 1)}
$$

and df =  $n_1 + n_2 - 2$ 

Alternative case when  $\sigma_1$  and  $\sigma_2$  are both **known values:**

$$
z = \frac{(\bar{x}_1 - \bar{x}_2) - (\mu_1 - \mu_2)}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}}
$$

## **Matched Pairs**

Requires simple random samples of matched pairs and either the number of matched pairs is  $n > 30$  or the pairs have differences from a population with a distribution that is approximately normal.

- *d*: Individual difference between values in a single matched pair
- $\mu_d$ : Population mean difference for all matched pairs
- *d*: Mean of all *sample* differences *d*
- *sd*: Standard deviation of all *sample* differences *d*
- *n*: Number of *pairs* of data

Test statistic: 
$$
t = \frac{\overline{d} - \mu_d}{\frac{s_d}{\sqrt{n}}}
$$
 where df =  $n - 1$ 

## **Hypothesis Testing (Two Variances or Two Standard Deviations)**

Requires independent simple random samples from populations with normal distributions.

- *s*1 2 : *larger* of the two sample variances
- *n*1: size of the sample with the *larger* variance
- $\sigma_1^2$ : variance of the population with the *larger* sample variance

Test statistic:  $F = \frac{s_1^2}{2}$  $\frac{s_1}{s_2^2}$  where  $s_1^2$  is the *larger* of the

two sample variances and numerator df  $= n_1 - 1$ and denominator df =  $n_2$  - 1

## **Correlation**

**Scatterplot:** Graph of paired (*x*, *y*) sample data. **linear Correlation Coefficient** *r:* Measures

strength of *linear* association between the two variables.

**Property of** *r***:**  $-1 \le r \le 1$ 

**Correlation Requirements:** Bivariate normal distribution (for any fixed value of *x*, the values of *y* are normally distributed, and for any fixed value of *y*, the values of *x* are normally distributed).

## **linear Correlation Coefficient:**

$$
r = \frac{n\Sigma xy - (\Sigma x)(\Sigma y)}{\sqrt{n(\Sigma x^2) - (\Sigma x)^2} \sqrt{n(\Sigma y^2) - (\Sigma y)^2}}
$$

$$
\text{or } r = \frac{\Sigma (z_x z_y)}{n-1}
$$

**Explained Variation:**  $r^2$  is the proportion of the variation in *y* that is explained by the linear association between *x* and *y*.

#### **Hypothesis test**

1. Using *r* as test statistic: If  $|r| \geq$  critical value (from table), then there is sufficient evidence to support a claim of linear correlation.

If  $|r| <$  critical value, there is not sufficient evidence to support a claim of linear correlation. 2. Using *t* as test statistic:

$$
t = \frac{r}{\sqrt{\frac{1 - r^2}{n - 2}}}
$$
 with df = n - 2

triolates the stats  $s$  -dependence of  $\frac{1}{2}$  26/11/16 1:02 PM is a 26/11/16  $\frac{1}{2}$  PM is a 26/11/16  $\frac{1}{2}$ 

## **Triola Statistics Series Review**

**Regression**

**Regression Equation:**  $\hat{v} = b_0 + b_1x$ 

- *x*: Independent (predictor, explanatory) variable
- $\hat{y}$ : Dependent (response) variable
- *b*<sub>1</sub>: Slope of regression line

$$
b_1 = r \frac{s_y}{s_x}
$$
 or  $b_1 = \frac{n \sum(xy) - (\sum x)(\sum y)}{n(\sum x^2) - (\sum x)^2}$ 

*b*<sub>0</sub>: *y*-intercept of regression line

 $b_0 = \overline{y} - b_1 \overline{x}$ 

**Predicting value of** *y*: If *no* linear correlation, best predicted *y* value is  $\overline{y}$ ; if there *is* a linear correlation, the best predicted *y* value is found by substituting *x* value into regression equation.

**Marginal Change:** Amount a variable changes when the other variable changes by one unit.

Slope  $b_1$  is the marginal change in *y* when *x* changes by one unit.

**Influential Point:** Strongly affects graph of regression line.

**Residual:** Difference between an observed sample *y* value and the value  $\hat{y}$  that is predicted using the regression equation.

Residual = 
$$
y - \hat{y}
$$
.

**least-Squares Property:** The sum of squares of the residuals is the smallest sum possible.

## **Correlation/Regression: Variation and Prediction Intervals**

Total Deviation:  $y - \overline{y}$ Explained Deviation:  $\hat{v} - \overline{v}$ Unexplained Deviation:  $y - \hat{y}$  $\sum (y - \bar{y})^2 = \sum (\hat{y} - \bar{y})^2 + \sum (y - \hat{y})^2$ 

Coefficient of Determination:

 $r^2 = \frac{\text{explained variation}}{}$ total variation

Standard Error of Estimate:

$$
s_e = \sqrt{\frac{\sum (y - \hat{y})^2}{n - 2}}
$$

**Prediction Interval for an Individual** *y:*

 $\hat{y} - E < y < \hat{y} + E$  where  $x_0$  is given,

$$
E = t_{\alpha/2} s_e \sqrt{1 + \frac{1}{n} + \frac{n(x_0 - \bar{x})^2}{n(\sum x^2) - (\sum x)^2}}
$$
  
and  $t_{\alpha/2}$  has df =  $n - 2$ .

## **Multiple Regression**

**Procedure:** *Obtain results using computer or calculator.*

#### **Multiple Regression Equation:**

$$
\hat{y} = b_0 + b_1 x_1 + b_2 x_2 + \ldots + b_k x_k
$$

- *n*: Sample size
- *k*: Number of *independent* (*x*) variables

**Important factors to consider:**

*P*-value (from computer display) and adjusted  $R^2$ , where

$$
adj. R2 = 1 - \frac{(n-1)}{[n - (k+1)]} (1 - R2)
$$

#### **goodness-of-Fit**

**goodness-of-Fit Test:** Test the null hypothesis that an observed frequency distribution fits claimed distribution.

- *O*: Observed frequency
- *E*: Expected frequency
- *k*: Number of different categories
- Total number of trials

**Requirements:** Random data of frequency counts, and  $E \geq 5$  for each category.

Test is *right-tailed* with test statistic:

$$
\chi^2 = \sum \frac{(O - E)^2}{E}
$$

where  $df = k - 1$ 

## **Contingency Tables**

**Contingency Table:** Two-way frequency table with row variable and column variable.

**Requirements:** Random data of frequency counts and  $E \ge 5$  for each cell (where *E* is expected frequency).

**Test of Independence:** Test null hypothesis of no association between row variable and column variable.

**Test of Homogeneity:** Test that different populations have same proportions of some characteristic.

Test (independence or homogeneity) is right-tailed with test statistic:

$$
\chi^2 = \sum \frac{(O - E)^2}{E}
$$

where df =  $(r - 1)(c - 1)$ 

**Expected frequency (of cell):**

$$
E = \frac{(\text{row total}) (\text{column total})}{(\text{grand total})}
$$

## **McNemar's Test**

Use for a  $2 \times 2$  frequency table from matched pairs. **Requirement:**  $b + c \ge 10$ , where *b* and *c* are frequencies from discordant pairs.

Test is *right-tailed* with test statistic

$$
\chi^2 = \frac{(|b - c| - 1)^2}{b + c}
$$

where  $df = 1$ 

## **One-way AnOVA**

**Procedure:** *Obtain P-value using computer or calculator.*

**AnOVA (analysis of variance):** Method of testing equality of three or more population means (by analyzing sample variances).

**Requirements:** Populations have approximately normal distributions, populations have same variance  $\sigma^2$ , samples are simple random samples, and samples are independent.

**Decision criterion** using significance level  $\alpha$ :



Test statistic:

 $F = \frac{\text{variance between samples}}{}$ variance within samples

## **Two-way AnOVA**

**Two-way AnOVA** (analysis of variance) uses *two* factors: row factor and column factor.

**Requirements:** Data in each cell are from a normally distributed population; populations have same variance; sample data are from simple random samples; samples are independent; data are categorized two ways.

**Balanced Design:** All cells have the same number of sample values.

**Procedure:** Use computer or calculator to obtain *P*-values, then use the *P*-values to test for

- 1. **Interaction effect** between the two variables. (Stop here if there appears to be an interaction effect.)
- 2. **Effect from row variable**
- 3. **Effect from column variable**

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#### **nonparametric Methods**

**nonparametric (distribution-free) tests:** Do not require assumptions about the population distributions.

**Rank:** Number assigned to a sample value according to its order in the *sorted* list. Lowest value has rank 1, 2nd lowest has rank 2, and so on.

**Sign Test:** Uses plus/minus signs instead of original data values. Used to test claims involving matched pairs, nominal data, or claims about median.

- *n*: Total number of (nonzero) signs
- *x*: Number of the less frequent sign

Test statistic if  $n \leq 25$ : *x* 

Test statistic if  $n > 25$ :

$$
z = \frac{(x + 0.5) - (n/2)}{\sqrt{n}/2}
$$

Reject  $H_0$  if test statistic  $\leq$  critical values.

**Wilcoxon Signed-Ranks Test:** Uses ranks of data consisting of *matched pairs;* based on ranks of differences between pairs of values. Used to test null hypothesis that the matched pairs have differences with a median equal to zero.

*T*: Smaller of two rank sums (sum of absolute values of negative ranks; sum of positive ranks)

Test statistic if  $n \leq 30 \cdot T$ 

Test statistic if  $n > 30$ :

$$
z = \frac{T - \frac{n(n+1)}{4}}{\sqrt{\frac{n(n+1)(2n+1)}{24}}}
$$

**Wilcoxon Rank-Sum Test: Uses ranks from** two *independent* samples. Used to test null hypothesis that two independent samples are from populations with same median. Requires two independent random samples, each with more than 10 values.

Sum of ranks for Sample 1

$$
n_1
$$
: Size of Sample 1

*n*2: Size of Sample 2

Test statistic: 
$$
z = \frac{R - \mu_R}{\sigma}
$$

where 
$$
\mu_R = \frac{n_1(n_1 + n_2 + 1)}{2}
$$

 $\sigma$ <sub>*p*</sub>

and 
$$
\sigma_R = \sqrt{\frac{n_1 n_2 (n_1 + n_2 + 1)}{12}}
$$

**Kruskal-Wallis Test:** Used to test null hypothesis that three or more independent samples are from populations with same median. Requires at least five observations in each independent sample of randomly selected values.

- *N*: Total number of observations
- *R*1: Sum of ranks for Sample 1
- *n*<sub>1</sub>: Number of values in Sample 1

Test statistic:

$$
H = \frac{12}{N(N+1)} \left( \frac{R_1^2}{n_1} + \frac{R_2^2}{n_2} + \ldots + \frac{R_k^2}{n_k} \right) - 3(N+1)
$$

where test is right-tailed using  $\chi^2$  distribution with  $df = k - 1$  and  $k =$  number of samples

**Rank Correlation:** Uses ranks to test for correlation from random paired data.

- *rs*: Rank correlation coefficient
- 
- *d*: Difference between ranks for two values within a pair

Test statistic (if no ties in ranks):

$$
r_s = 1 - \frac{6\sum d^2}{n(n^2 - 1)}
$$

Test statistic (with ties among ranks):

$$
r_s = \frac{n\Sigma xy - (\Sigma x)(\Sigma y)}{\sqrt{n(\Sigma x^2) - (\Sigma x)^2} \sqrt{n(\Sigma y^2) - (\Sigma y)^2}}
$$

Critical values for  $n > 30$ :

$$
r_s = \frac{\pm z}{\sqrt{n-1}}
$$

**Runs Test for Randomness:** Used to determine whether sequence of sample data is in random order.

**Run:** Sequence of data having same characteristic.

- $n_1$ : Total number of sample elements having one common characteristic
- *n*<sub>2</sub>: Total number of sample elements having the other characteristic
- *G*: Total number of runs

If  $\alpha = 0.05$ ,  $n_1 \leq 20$ , and  $n_2 \leq 20$ , test statistic is *G*. Otherwise,

Test statistic: 
$$
z = \frac{G - \mu_G}{\sigma_G}
$$

where 
$$
\mu_G = \frac{2n_1n_2}{n_1 + n_2} + 1
$$
 and  
\n
$$
\sigma_G = \sqrt{\frac{(2n_1n_2)(2n_1n_2 - n_1 - n_2)}{(n_1 + n_2)^2(n_1 + n_2 - 1)}}
$$

## **Statistical Process Control**

**Process Data:** Data arranged according to some time sequence.

**Run Chart:** Sequential plot of *individual* data values over time

**Control Chart** of a process characteristic: Sequential plot over time of the characteristic values, including a centerline, lower control limit (LCL), and upper control limit (UCL).

*R* **Chart (range chart):** Control chart of sample ranges.



*x* **Chart:** Control chart of sample means.

UCL:  $\overline{\overline{x}} + A_2\overline{R}$ Centerline:  $\overline{\overline{x}}$ LCL:  $\bar{\bar{x}} - A_2\bar{R}$ 

$$
p
$$
 Chart: Control chart to monitor the proportion  $p$  of some attribute.

UCL: 
$$
\bar{p} + 3\sqrt{\frac{\bar{p}\bar{q}}{n}}
$$
  
Centerline:  $\bar{p}$ 

$$
\text{LCL:} \quad \overline{p} - 3\sqrt{\frac{\overline{p}}{n}} \overline{q}
$$

where  $\bar{p} = \frac{\text{total number of defects}}{1 - \frac{p}{\sigma}}$ total number of items

#### **Statistically Stable Process (or within**

**statistical control):** Process with only natural variation and no patterns, cycles, or unusual points

#### **Out-of-Control Criteria:**

- 1. There is a pattern, trend, or cycle that is not random.
- 2. There is a point above the upper control limit or below the lower control limit.
- 3. There are eight consecutive points all above or all below the centerline.

#### **COntROl ChaRt COnStantS**



 $\tau$  triolastic stats  $s$  of  $\tau$  or  $\tau$  or  $\tau$  . In  $\tau$  is  $\tau$ 

- -
	- *n*: Number of pairs of sample data











Area to the *Right* of the Critical Value





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#### TABLE A-5 F Distribution ( $\alpha = 0.025$  in the right tail)

