

populations have the same mean, why not simply pair them off and test two at a time by testing  $H_0: \mu_1 = \mu_2$ ,  $H_0: \mu_2 = \mu_3$ , and  $H_0: \mu_1 = \mu_3$ ? For the data in Table 12-1, the approach of testing equality of two means at a time requires three different hypothesis tests. If we use a 0.05 significance level for each of those three hypothesis tests, the actual overall confidence level could be as low as  $0.95^3$  (or 0.857). In general, as we increase the number of individual tests of significance, we increase the risk of finding a difference by chance alone (instead of a real difference in the means). The risk of a type I error—finding a difference in one of the pairs when no such difference actually exists—is far too high. The method of analysis of variance helps us avoid that particular pitfall (rejecting a true null hypothesis) by using *one test* for equality of several means, instead of several tests that each compare two means at a time.

**CAUTION** When testing for equality of three or more populations, use analysis of variance. (Using multiple hypothesis tests with two samples at a time could wreak havoc with the significance level.)

## Part 2: Calculations and Identifying Means That Are Different

### Calculations with Equal Sample Sizes $n$

Table 12-2 can be very helpful in understanding the methods of ANOVA. In Table 12-2, compare Data Set A to Data Set B to see that Data Set A is the same as Data Set B with this notable exception: the Sample 1 values each differ by 10. If the data sets all have the same sample size (as in  $n = 4$  for Table 12-2), the following calculations aren't too difficult, as shown on the next page.

**Table 12-2** Effect of a Mean on the  $F$  Test Statistic

| A                          |   |                   | B                          |  |                   |
|----------------------------|---|-------------------|----------------------------|--|-------------------|
| add 10                     |   |                   |                            |  |                   |
| Sample 1                   | Sample 2  | Sample 3          | Sample 1                   | Sample 2   | Sample 3          |
| 7                          | 6   | 4                 | 17                         | 6  | 4                 |
| 3                          | 5   | 7                 | 13                         | 5  | 7                 |
| 6                          | 5   | 6                 | 16                         | 5  | 6                 |
| 6                          | 8   | 7                 | 16                         | 8  | 7                 |
| ↓                          | ↓   | ↓                 | ↓                          | ↓  | ↓                 |
| $n_1 = 4$                  | $n_2 = 4$   | $n_3 = 4$         | $n_1 = 4$                  | $n_2 = 4$  | $n_3 = 4$         |
| $\bar{x}_1 = 5.5$          | $\bar{x}_2 = 6.0$   | $\bar{x}_3 = 6.0$ | $\bar{x}_1 = 15.5$         | $\bar{x}_2 = 6.0$  | $\bar{x}_3 = 6.0$ |
| $s_1^2 = 3.0$              | $s_2^2 = 2.0$   | $s_3^2 = 2.0$     | $s_1^2 = 3.0$              | $s_2^2 = 2.0$  | $s_3^2 = 2.0$     |
| Variance between samples   | $ns_{\bar{x}}^2 = 4(0.0833) = 0.3332$                               |                   | Variance between samples   | $ns_{\bar{x}}^2 = 4(30.0833) = 120.3332$                               |                   |
| Variance within samples    | $s_p^2 = \frac{3.0 + 2.0 + 2.0}{3} = 2.3333$                        |                   | Variance within samples    | $s_p^2 = \frac{3.0 + 2.0 + 2.0}{3} = 2.3333$                           |                   |
| F test statistic           | $F = \frac{ns_{\bar{x}}^2}{s_p^2} = \frac{0.3332}{2.3333} = 0.1428$ |                   | F test statistic           | $F = \frac{ns_{\bar{x}}^2}{s_p^2} = \frac{120.3332}{2.3333} = 51.5721$ |                   |
| P-value (found from Excel) | P-value = 0.8688  |                   | P-value (found from Excel) | P-value = 0.0000118  |                   |