

# Confidence Intervals for The Difference of Means of Two Independent Populations

In all cases we assume:

1. The data are

$Y_{1,1}, Y_{1,2}, \dots, Y_{1,n_1}$ , where  $Y_{1,j} = \mu_1 + \epsilon_{1,j}$ , (population 1)

$Y_{2,1}, Y_{2,2}, \dots, Y_{2,n_2}$ , where  $Y_{2,j} = \mu_2 + \epsilon_{2,j}$ , (population 2).

2. The two populations are independent.

## Case 1: Known Variances

### Assumptions

1. Either
  - (a)  $n_1$  and  $n_2$  are large, or
  - (b) The  $\epsilon_{1,j}$  are from a  $N(0, \sigma_1^2)$  population, and the  $\epsilon_{2,j}$  are from a  $N(0, \sigma_2^2)$  population.
2.  $\sigma_1^2$  and  $\sigma_2^2$  are known.

### Formulas

A level  $L$  confidence interval for  $\mu_1 - \mu_2$  is  $\left( \bar{Y}_1 - \bar{Y}_2 - \sigma(\bar{Y}_1 - \bar{Y}_2)z_{\frac{1+L}{2}}, \bar{Y}_1 - \bar{Y}_2 + \sigma(\bar{Y}_1 - \bar{Y}_2)z_{\frac{1+L}{2}} \right)$ ,

where  $\sigma(\bar{Y}_1 - \bar{Y}_2) = \sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}$ , and  $z_{\frac{1+L}{2}}$  may be obtained from a table of normal quantiles ([click here](#)).

## Case 2: Unknown Variances, $n_1$ and $n_2$ Large

This case is treated exactly as Case 1 except that  $\sigma(\bar{Y}_1 - \bar{Y}_2)$  is replaced by  $\hat{\sigma}(\bar{Y}_1 - \bar{Y}_2) = \sqrt{\frac{S_1^2}{n_1} + \frac{S_2^2}{n_2}}$ , where  $S_1$  and  $S_2$  are the sample standard deviations computed from the data from populations 1 and 2 respectively.

## Case 3: Variances Unknown, but Assumed to Be Equal, $n_1$ and $n_2$ Not Both Large

### Assumptions

1. The  $\epsilon_{1,j}$  are from a  $N(0, \sigma_1^2)$  population, and the  $\epsilon_{2,j}$  are from a  $N(0, \sigma_2^2)$  population.
2.  $\sigma_1^2$  and  $\sigma_2^2$  are unknown, but are assumed to be equal.

### Formulas

A level  $L$  confidence interval for  $\mu_1 - \mu_2$  is  $\left( \bar{Y}_1 - \bar{Y}_2 - \hat{\sigma}_p(\bar{Y}_1 - \bar{Y}_2)t_{\nu, \frac{1+L}{2}}, \bar{Y}_1 - \bar{Y}_2 + \hat{\sigma}_p(\bar{Y}_1 - \bar{Y}_2)t_{\nu, \frac{1+L}{2}} \right)$ ,

where  $\hat{\sigma}_p(\bar{Y}_1 - \bar{Y}_2) = \sqrt{S_p^2 \left( \frac{1}{n_1} + \frac{1}{n_2} \right)}$ , where  $S_p^2 = \frac{(n_1-1)S_1^2 + (n_2-1)S_2^2}{n_1+n_2-2}$ ,  $S_1$  and  $S_2$  are the sample standard deviations computed from the data from populations 1 and 2 respectively,  $\nu = n_1 + n_2 - 2$ , and  $t_{\nu, \frac{1+L}{2}}$  may be obtained from a table of quantiles of the  $t$  distribution ([click here](#)).

## Case 4: Variances Unknown, and Not Assumed to Be Equal, $n_1$ and $n_2$ Not Both Large

### Assumptions

1. The  $\epsilon_{1,j}$  are from a  $N(0, \sigma_1^2)$  population, and the  $\epsilon_{2,j}$  are from a  $N(0, \sigma_2^2)$  population.
2.  $\sigma_1^2$  and  $\sigma_2^2$  are unknown, and are not assumed to be equal.

### Formulas

A level  $L$  confidence interval for  $\mu_1 - \mu_2$  is  $\left( \bar{Y}_1 - \bar{Y}_2 - \hat{\sigma}(\bar{Y}_1 - \bar{Y}_2)t_{\nu, \frac{1+L}{2}}, \bar{Y}_1 - \bar{Y}_2 + \hat{\sigma}(\bar{Y}_1 - \bar{Y}_2)t_{\nu, \frac{1+L}{2}} \right)$ ,

where  $\hat{\sigma}(\bar{Y}_1 - \bar{Y}_2) = \sqrt{\frac{S_1^2}{n_1} + \frac{S_2^2}{n_2}}$ ,  $S_1$  and  $S_2$  are the sample standard deviations computed from the data from populations 1 and 2 respectively, the degrees of freedom  $\nu$  is taken as the largest integer less than or equal to  $\left[ \frac{S_1^2}{n_1} + \frac{S_2^2}{n_2} \right]^2 / \left[ \frac{(S_1^2/n_1)^2}{n_1-1} + \frac{(S_2^2/n_2)^2}{n_2-1} \right]$ , and  $t_{\nu, \frac{1+L}{2}}$  may be obtained from a table of quantiles of the  $t$  distribution ([click here](#)).

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