

## Statistical Inference on population means and variances

The information provided in this document are for reference only

### 3 conditions

- A. Normal Distributions
- B. Parameters Known
- C. Large Sample Size

Legend for conditions: Y = Yes, N = No, -- = It does not matter

## Confidence Interval

### One Sample

Estimation of the population mean,  $\mu$

Conditions (A, B, C)	Statistic	Distribution	100(1 - $\alpha$ )% Confidence Interval
(Y, Y( $\sigma^2$ ), --)	$T = \frac{\bar{X} - \mu}{\sigma/\sqrt{n}}$	$N(0,1)$	$\bar{X} \pm z_{\alpha/2} \frac{\sigma}{\sqrt{n}}$
(N, Y( $\sigma^2$ ), Y)	$T = \frac{\bar{X} - \mu}{\sigma/\sqrt{n}}$	Approximate $N(0,1)$ (CLT)	$\bar{X} \pm z_{\alpha/2} \frac{\sigma}{\sqrt{n}}$
(Y, N, --)	$T = \frac{\bar{X} - \mu}{S/\sqrt{n}}$	$t(n-1)$	$\bar{X} \pm t_{n-1; \alpha/2} \frac{S}{\sqrt{n}}$
(N, N, Y)	$T = \frac{\bar{X} - \mu}{S/\sqrt{n}}$	Approximate $N(0,1)$ (CLT & LLN)	$\bar{X} \pm z_{\alpha/2} \frac{S}{\sqrt{n}}$

\* Y = Yes, N = No, - = It does not matter

CLT: Central Limit Theorem, LLN = Law of Large Numbers

$\Pr(Z > z_{\alpha/2}) = \alpha/2$  with  $Z \sim N(0,1)$ ,  $\Pr(T > t_{v; \alpha/2}) = \alpha/2$  with  $T \sim t(v)$

Estimation of the population variance,  $\sigma^2$

Conditions (A, B, C)	Statistic	Distribution	100(1 - $\alpha$ )% Confidence Interval
(Y, Y( $\mu$ ), --)	$T = \frac{\sum_{i=1}^n (X_i - \mu)^2}{\sigma^2}$	$\chi^2(n)$	$\frac{\sum_{i=1}^n (X_i - \mu)^2}{\chi_{n; \alpha/2}^2} < \sigma^2 < \frac{\sum_{i=1}^n (X_i - \mu)^2}{\chi_{n; 1-\alpha/2}^2}$
(Y, N, N)	$T = \frac{(n-1)S^2}{\sigma^2}$	$\chi^2(n-1)$	$\frac{(n-1)S^2}{\chi_{n-1; \alpha/2}^2} < \sigma^2 < \frac{(n-1)S^2}{\chi_{n-1; 1-\alpha/2}^2}$

$\Pr(W > \chi_{(v; \alpha)}^2) = \alpha$  with  $W \sim \chi^2(v)$

## Two Samples

### Paired Samples

Consider  $D_i = X_i - Y_i$ . Let  $\mu_D = \mu_X - \mu_Y$

Estimation of the population mean,  $\mu_D$

Conditions (A, B, C)	Statistic	Distribution	100(1 - $\alpha$ )% Confidence Interval
(Y, N, N)	$T = \frac{\bar{X}_D - \mu_D}{S_D/\sqrt{n}}$	$t(n - 1)$	$\bar{X}_D \pm t_{n-1, \alpha/2} \frac{S_D}{\sqrt{n}}$
(N, N, Y)	$T = \frac{\bar{X}_D - \mu_D}{S_D/\sqrt{n}}$	Approximate $N(0,1)$ (CLT & LLN)	$\bar{X}_D \pm z_{\alpha/2} \frac{S_D}{\sqrt{n}}$

### Two Independent Samples

Estimation of the difference of population means,  $\mu_1 - \mu_2$

Conditions (A, B, C)	Statistic	Distribution	100(1 - $\alpha$ )% Confidence Interval
(Y, Y( $\sigma_1^2, \sigma_2^2$ ), --)	$T = \frac{\bar{X}_1 - \bar{X}_2 - (\mu_1 - \mu_2)}{\sqrt{\sigma_1^2/n_1 + \sigma_2^2/n_2}}$	$N(0,1)$	$\bar{X}_1 - \bar{X}_2 \pm z_{\alpha/2} \sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}$
(N, Y( $\sigma_1^2, \sigma_2^2$ ), Y)	$T = \frac{\bar{X}_1 - \bar{X}_2 - (\mu_1 - \mu_2)}{\sqrt{\sigma_1^2/n_1 + \sigma_2^2/n_2}}$	Approximate $N(0,1)$	$\bar{X}_1 - \bar{X}_2 \pm z_{\alpha/2} \sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}$
(Y, N( $\sigma_1^2 = \sigma_2^2$ ), --)	$T = \frac{\bar{X}_1 - \bar{X}_2 - (\mu_1 - \mu_2)}{\sqrt{S_p^2(1/n_1 + 1/n_2)}}$	$t(n_1 + n_2 - 2)$	$\bar{X}_1 - \bar{X}_2 \pm t_{n_1+n_2-2, \alpha/2} \sqrt{S_p^2 \left( \frac{1}{n_1} + \frac{1}{n_2} \right)}$
(N, N, Y)	$T = \frac{\bar{X}_1 - \bar{X}_2 - (\mu_1 - \mu_2)}{\sqrt{S_1^2/n_1 + S_2^2/n_2}}$	Approximate $N(0,1)$ (CLT & LLN)	$\bar{X}_1 - \bar{X}_2 \pm z_{\alpha/2} \sqrt{\frac{S_1^2}{n_1} + \frac{S_2^2}{n_2}}$

Estimation of the ratio of population variances,  $\sigma_1^2/\sigma_2^2$

Conditions (A, B, C)	Statistic	Distribution	100(1 - $\alpha$ )% Confidence Interval
(Y, N, --)	$T = \frac{S_1^2/\sigma_1^2}{S_2^2/\sigma_2^2}$	$F_{n_1-1, n_2-1}$	$\frac{S_1^2}{S_2^2} \frac{1}{F_{n_1-1, n_2-1; \alpha/2}} < \frac{\sigma_1^2}{\sigma_2^2} < \frac{S_1^2}{S_2^2} F_{n_2-1, n_1-1; \alpha/2}$ <p style="text-align: center;">Or</p> $\frac{S_1^2}{S_2^2} \frac{1}{F_{n_1-1, n_2-1; 1-\alpha/2}} < \frac{\sigma_1^2}{\sigma_2^2} < \frac{S_1^2}{S_2^2} F_{n_1-1, n_2-1; 1-\alpha/2}$

$$\Pr(W > F_{v_1, v_2; \alpha}) = \alpha \text{ with } W \sim F(v_1, v_2)$$

## Hypothesis Testing

### One Sample

Hypothesis testing on the population mean.  $H_0: \mu = \mu_0$

Conditions (A, B, C)	Test Statistic	Alternative	Reject $H_0$ if
(Y, Y( $\sigma^2$ ), --)	$T = \frac{\bar{X} - \mu_0}{\sigma/\sqrt{n}}$	$H_1: \mu \neq \mu_0$	$T < -z_{\alpha/2}$ or $T > z_{\alpha/2}$
		$H_1: \mu > \mu_0$	$T > z_{\alpha}$
		$H_1: \mu < \mu_0$	$T < -z_{\alpha}$
(N, Y( $\sigma^2$ ), Y)	$T = \frac{\bar{X} - \mu_0}{\sigma/\sqrt{n}}$	$H_1: \mu \neq \mu_0$	$T < -z_{\alpha/2}$ or $T > z_{\alpha/2}$
		$H_1: \mu > \mu_0$	$T > z_{\alpha}$
		$H_1: \mu < \mu_0$	$T < -z_{\alpha}$
(Y, N, --)	$T = \frac{\bar{X} - \mu_0}{S/\sqrt{n}}$	$H_1: \mu \neq \mu_0$	$T < -t_{n-1; \alpha/2}$ or $T > t_{n-1; \alpha/2}$
		$H_1: \mu > \mu_0$	$T > t_{n-1; \alpha}$
		$H_1: \mu < \mu_0$	$T < -t_{n-1; \alpha}$
(N, N, Y)	$T = \frac{\bar{X} - \mu_0}{S/\sqrt{n}}$	$H_1: \mu \neq \mu_0$	$T < -z_{\alpha/2}$ or $T > z_{\alpha/2}$
		$H_1: \mu > \mu_0$	$T > z_{\alpha}$
		$H_1: \mu < \mu_0$	$T < -z_{\alpha}$

\* Y = Yes, N = No, - = It does not matter

Hypothesis testing on the population variance.  $H_0: \sigma^2 = \sigma_0^2$

Conditions (A, B, C)	Test Statistic	Alternative	Reject $H_0$ if
(Y, N, --)	$T = \frac{(n-1)S^2}{\sigma_0^2}$	$H_1: \sigma^2 \neq \sigma_0^2$	$T < \chi_{(n-1; 1-\alpha/2)}^2$ or $T > \chi_{(n-1; \alpha/2)}^2$
		$H_1: \sigma^2 > \sigma_0^2$	$T > \chi_{(n-1; \alpha)}^2$
		$H_1: \sigma^2 < \sigma_0^2$	$T < \chi_{(n-1; 1-\alpha)}^2$

$\Pr(W > \chi_{(v; \alpha)}^2) = \alpha$  with  $W \sim \chi^2(v)$

## Two Samples

### Paired Samples

Consider  $D_i = X_i - Y_i$ . Let  $\mu_D = \mu_X - \mu_Y$

Hypothesis testing on the population mean,  $H_0: \mu_D = 0$

Conditions (A, B, C)	Test Statistic	Alternative	Reject $H_0$ if
(Y, N, --)	$T = \frac{\bar{X}_D - 0}{S_D/\sqrt{n}}$	$H_1: \mu_D \neq 0$	$T < -t_{n-1; \alpha/2}$ or $T > t_{n-1; \alpha/2}$
		$H_1: \mu_D > 0$	$T > t_{n-1; \alpha}$
		$H_1: \mu_D < 0$	$T < -t_{n-1; \alpha}$
(N, N, Y)	$T = \frac{\bar{X}_D - 0}{S_D/\sqrt{n}}$	$H_1: \mu_D \neq 0$	$T < -z_{\alpha/2}$ or $T > z_{\alpha/2}$
		$H_1: \mu_D > 0$	$T > z_{\alpha}$
		$H_1: \mu_D < 0$	$T < -z_{\alpha}$

## Two Independent Samples

Hypothesis testing on the difference of population means,  $H_0: \mu_1 - \mu_2 = 0$

Conditions (A, B, C)	Test Statistic	Alternative	Reject $H_0$ if
$(Y, Y(\sigma_1^2, \sigma_2^2), --)$	$T = \frac{\bar{X}_1 - \bar{X}_2 - 0}{\sqrt{\sigma_1^2/n_1 + \sigma_2^2/n_2}}$	$H_1: \mu_1 - \mu_2 \neq 0$	$T < -z_{\alpha/2}$ or $T > z_{\alpha/2}$
		$H_1: \mu_1 - \mu_2 > 0$	$T > z_{\alpha}$
		$H_1: \mu_1 - \mu_2 < 0$	$T < -z_{\alpha}$
$(N, Y(\sigma_1^2, \sigma_2^2), Y)$	$T = \frac{\bar{X}_1 - \bar{X}_2 - 0}{\sqrt{\sigma_1^2/n_1 + \sigma_2^2/n_2}}$	$H_1: \mu_1 - \mu_2 \neq 0$	$T < -z_{\alpha/2}$ or $T > z_{\alpha/2}$
		$H_1: \mu_1 - \mu_2 > 0$	$T > z_{\alpha}$
		$H_1: \mu_1 - \mu_2 < 0$	$T < -z_{\alpha}$
$(Y, N(\sigma_1^2 = \sigma_2^2), --)$	$T = \frac{\bar{X}_1 - \bar{X}_2 - 0}{\sqrt{S_p^2(1/n_1 + 1/n_2)}}$	$H_1: \mu_1 - \mu_2 \neq 0$	$T < -t_{n_1+n_2-2; \alpha/2}$ or $T > t_{n_1+n_2-2; \alpha/2}$
		$H_1: \mu_1 - \mu_2 > 0$	$T > t_{n_1+n_2-2; \alpha}$
		$H_1: \mu_1 - \mu_2 < 0$	$T < -t_{n_1+n_2-2; \alpha}$
$(N, N, Y)$	$T = \frac{\bar{X}_1 - \bar{X}_2 - 0}{\sqrt{S_1^2/n_1 + S_2^2/n_2}}$	$H_1: \mu_1 - \mu_2 \neq 0$	$T < -z_{\alpha/2}$ or $T > z_{\alpha/2}$
		$H_1: \mu_1 - \mu_2 > 0$	$T > z_{\alpha}$
		$H_1: \mu_1 - \mu_2 < 0$	$T < -z_{\alpha}$

Hypothesis testing on the ratio of population variances,  $H_0: \sigma_1^2/\sigma_2^2 = 1$

Conditions (A, B, C)	Test Statistic	Alternative	Reject $H_0$ if
$(Y, N, --)$	$T = \frac{S_1^2}{S_2^2}$	$H_1: \sigma_1^2/\sigma_2^2 \neq 1$	$F < F_{n_1-1, n_2-1; 1-\alpha/2}$ or $F > F_{n_1-1, n_2-1; \alpha/2}$
		$H_1: \sigma_1^2/\sigma_2^2 > 1$	$F > F_{n_1-1, n_2-1; \alpha}$
		$H_1: \sigma_1^2/\sigma_2^2 < 1$	$F < F_{n_1-1, n_2-1; 1-\alpha} (= 1/F_{n_2-1, n_1-1; \alpha})$

$\Pr(W > F_{v_1, v_2; \alpha}) = \alpha$  with  $W \sim F(v_1, v_2)$