

Introduction to Probability and Statistics

Review of interval estimates

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Here is a list of all of the situations for which we have constructed confidence intervals. In each case the situation is that we have drawn a random sample X_1, \dots, X_n from a population whose distribution is known up to some parameter.

Normal population, mean μ unknown, variance σ^2 known

- **Unbiased estimator for μ :** $\bar{X} = \frac{X_1 + \dots + X_n}{n}$
- **Maximum likelihood estimator for μ :** \bar{X}
- **100(1 - α)% confidence intervals for μ :**
 - **Two-sided:** $\left(\bar{X} - z_{\alpha/2} \frac{\sigma}{\sqrt{n}}, \bar{X} + z_{\alpha/2} \frac{\sigma}{\sqrt{n}} \right)$
 - **One-sided lower:** $\left(-\infty, \bar{X} + z_{\alpha} \frac{\sigma}{\sqrt{n}} \right)$
 - **One-sided upper:** $\left(\bar{X} - z_{\alpha} \frac{\sigma}{\sqrt{n}}, \infty \right)$
- **All based on:** $\frac{\bar{X} - \mu}{\sigma/\sqrt{n}} = Z$, a standard normal

Normal population, mean μ unknown, variance σ^2 unknown, estimating mean

- **100(1 - α)% confidence intervals for μ :**
 - **Two-sided:** $\left(\bar{X} - t_{\alpha/2, n-1} \frac{S}{\sqrt{n}}, \bar{X} + t_{\alpha/2, n-1} \frac{S}{\sqrt{n}} \right)$
 - **One-sided lower:** $\left(-\infty, \bar{X} + t_{\alpha, n-1} \frac{S}{\sqrt{n}} \right)$
 - **One-sided upper:** $\left(\bar{X} - t_{\alpha} \frac{S}{\sqrt{n}}, \infty \right)$
- **All based on:** $\frac{\bar{X} - \mu}{S/\sqrt{n}} = t_{n-1}$, a t -distribution with $n - 1$ degrees of freedom, where $S^2 = \frac{\sum_{i=1}^n (X_i - \bar{X})^2}{n-1}$ is sample variance

Normal population, mean μ unknown, variance σ^2 unknown, estimating variance

- **Unbiased estimator for σ^2 :** $S^2 = \frac{\sum_{i=1}^n (X_i - \bar{X})^2}{n-1}$
- **Maximum likelihood estimator for σ^2 :** $\frac{\sum_{i=1}^n (X_i - \bar{X})^2}{n}$
- **100(1 - α)% confidence intervals for σ^2 :**

– **Two-sided:** $\left(\frac{(n-1)S^2}{\chi_{\alpha/2, n-1}^2}, \frac{(n-1)S^2}{\chi_{1-\alpha/2, n-1}^2} \right)$

– **One-sided lower:** $\left(0, \frac{(n-1)S^2}{\chi_{1-\alpha, n-1}^2} \right)$

– **One-sided upper:** $\left(\frac{(n-1)S^2}{\chi_{\alpha, n-1}^2}, \infty \right)$

- **All based on:** $\frac{(n-1)S^2}{\sigma^2} = \chi_{n-1}^2$, a χ^2 distribution with $n - 1$ degrees of freedom

Two normal populations, means μ_1, μ_2 unknown, variances σ_1^2, σ_2^2 known, estimating difference between means

- **100(1 - α)% confidence intervals for $\mu_1 - \mu_2$:**

– **Two-sided:** $\left(\bar{X} - \bar{Y} - z_{\alpha/2} \sqrt{\frac{\sigma_1^2}{n} + \frac{\sigma_2^2}{m}}, \bar{X} - \bar{Y} + z_{\alpha/2} \sqrt{\frac{\sigma_1^2}{n} + \frac{\sigma_2^2}{m}} \right)$, where \bar{X} is sample mean from first population (sample size n), \bar{Y} is sample mean from second population (sample size m)

– **One-sided lower:** $\left(-\infty, \bar{X} - \bar{Y} + z_{\alpha} \sqrt{\frac{\sigma_1^2}{n} + \frac{\sigma_2^2}{m}} \right)$

– **One-sided upper:** $\left(\bar{X} - \bar{Y} - z_{\alpha} \sqrt{\frac{\sigma_1^2}{n} + \frac{\sigma_2^2}{m}}, \infty \right)$

- **All based on:** $\frac{(\bar{X} - \bar{Y}) - (\mu_1 - \mu_2)}{\sqrt{\frac{\sigma_1^2}{n} + \frac{\sigma_2^2}{m}}} = Z$, a standard normal

Two normal populations, means μ_1, μ_2 unknown, variances σ_1^2, σ_2^2 unknown, assumed equal to σ^2 , estimating difference between means

- **Pooled estimator for common variance:** $S_p^2 = \frac{(n-1)S_1^2 + (m-1)S_2^2}{n+m-2}$
- **100(1 - α)% confidence intervals for $\mu_1 - \mu_2$:**
 - **Two-sided:** $\left(\bar{X} - \bar{Y} - t_{\alpha/2, n+m-2} \sqrt{\frac{S_p^2}{n} + \frac{S_p^2}{m}}, \bar{X} - \bar{Y} + t_{\alpha/2, n+m-2} \sqrt{\frac{S_p^2}{n} + \frac{S_p^2}{m}} \right)$
 - **One-sided lower:** $\left(-\infty, \bar{X} - \bar{Y} + t_{\alpha, n+m-2} \sqrt{\frac{S_p^2}{n} + \frac{S_p^2}{m}} \right)$
 - **One-sided upper:** $\left(\bar{X} - \bar{Y} - t_{\alpha, n+m-2} \sqrt{\frac{S_p^2}{n} + \frac{S_p^2}{m}}, \infty \right)$
- **All based on:** $\frac{(\bar{X} - \bar{Y}) - (\mu_1 - \mu_2)}{\sqrt{\frac{S_p^2}{n} + \frac{S_p^2}{m}}} = t_{n+m-2}$, a t -distribution with $n + m - 2$ degrees of freedom

Bernoulli population, parameter p unknown

Here X_1, \dots, X_n is a sample of size n from a population in which an unknown proportion p have a certain characteristic, and each X_i is 1 if sample i has the characteristic, and 0 otherwise.

- **Approximate 100(1 - α)% confidence intervals for p , if sample size is large ($n \geq 30$):**
 - Using $\hat{p} = \frac{X_1 + \dots + X_n}{n}$ as an estimate for p in the variance:
 - * **Two-sided:** $\left(\hat{p} - z_{\alpha/2} \sqrt{\frac{\hat{p}(1-\hat{p})}{n}}, \hat{p} + z_{\alpha/2} \sqrt{\frac{\hat{p}(1-\hat{p})}{n}} \right)$
 - * **One-sided lower:** $\left(0, \hat{p} - z_{\alpha} \sqrt{\frac{\hat{p}(1-\hat{p})}{n}} \right)$
 - * **One-sided upper:** $\left(\hat{p} - z_{\alpha} \sqrt{\frac{\hat{p}(1-\hat{p})}{n}}, 1 \right)$
 - Using worst-case estimate $p = 1/2$ in the variance:
 - * **Two-sided:** $\left(\hat{p} - z_{\alpha/2} \sqrt{\frac{1}{4n}}, \hat{p} + z_{\alpha/2} \sqrt{\frac{1}{4n}} \right)$
 - * **One-sided lower:** $\left(0, \hat{p} - z_{\alpha} \sqrt{\frac{1}{4n}} \right)$
 - * **One-sided upper:** $\left(\hat{p} - z_{\alpha} \sqrt{\frac{1}{4n}}, 1 \right)$
- **All based on:** $\frac{\bar{X} - np}{\sqrt{np(1-p)}} \approx Z$, a standard normal (Central Limit Theorem), where $X = X_1 + \dots + X_n$.