Sampling and Estimation

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Introduction

- Now we are going to begin making inferences about the target population.
- Estimation involves assessing the value of an unknown population parameter.
 - mean, proportion, or variance.
- We use the corresponding sample estimates which are random variables characterized by some sampling distribution.
- We can then apply probability theory to our estimates.

Estimates

- Focused on two kinds of estimates:
 - Open testimates are a single number used to estimate the value of a population parameter.
 - Confidence interval estimates provide a range of values between which the value of the population parameter is believed to be.
- Due to sampling error, it is highly unlikely that our point estimate is equal to the population parameter.
- Confidence intervals also specify a probability that the interval correctly estimates the unknown parameter.
- The size of the confidence interval depends on the size of the sampling error.

Notation

Notation	Description
\bar{x}	Sample mean
$\mid \mu \mid$	Population mean
$\begin{array}{ c c } \mu \\ s^2 \end{array}$	Sample variance
σ^2	Population variance
s	Sample standard deviation
σ	Population standard deviation
ĝ	Sample proportion
π	Population proportion

Point Estimators

• The **sample mean** \bar{x} of the observations $x_1, x_2, ... x_n$ is given by

$$\bar{x} = \frac{x_1 + x_2 + \dots + x_n}{n} = \frac{\sum_{i=1}^n x_i}{n}$$
 (1)

• The **sample variance** s^2 is calculated by:

$$s^{2} = \frac{\sum (x_{i} - \bar{x})^{2}}{n - 1} = \frac{S_{xx}}{n - 1}.$$
 (2)

• The **sample standard deviation** *s* is calculated by:

$$s = \sqrt{s^2}. (3)$$

• The **population variance** σ^2 is calculated by:

$$\sigma^2 = \frac{\sum (x_i - \mu)^2}{N},\tag{4}$$

Unbiased Estimators

- A statistic is a biased estimator if it overestimates or underestimates the population parameter.
- The **Bias** of the estimator $\hat{\theta}$ is calculated by:

$$Bias(\hat{\theta}, \theta) = E[\hat{\theta} - \theta] = E[\hat{\theta}] - \theta.$$
 (5)

- A point estimator is said to be **unbiased** if the expected value of the estimator is the theoretical parameter θ ($Bias(\hat{\theta}, \theta) = 0$).
- The point estimators that we have examined have been shown to be unbiased.

Interval Estimates

- An interval estimate provides a range for a population parameter based on a sample.
- Intervals can be more informative than simple point estimates ("how plausible?")
- We generally assign a range of confidence out of 100%

Probability Interval

- A **probability interval** is any interval [A, B] such that the probability of falling between A and B is 1α .
 - General notation: $100(1-\alpha)\%$
- You can view α as the risk of incorrectly concluding that the confidence interval contains the true mean.

Confidence Intervals

- A **confidence interval** is an interval estimate that specifies the probability that the interval contains the true population parameter.
- This probability is called the **level of confidence**, denoted 1α .
- Usually expressed as a percentage:
 - Generally: 90%, 95%, or 99%
 - If $\alpha = 0.05$ then the level of confidence is 95%.
- This interval still may not include the actual population parameter.
- As the confidence level increases, so does the length of the confidence interval.

Confidence Interval Types

- Different formulas for estimating confidence intervals have been developed for different situations.
- The formulas may depend on:
 - Parameter we are trying to estimate.
 - Assumptions about the population distribution.
 - Any prior knowledge about the population variability.

Confidence Interval for the Mean with Known Standard Deviation I

- Simplest confidence interval for the mean of a population is with a known standard deviation.
- In practice the population standard deviation will not be known.
- 100(1- α) confidence interval for μ :

$$\bar{x} \pm z_{\alpha/2} \cdot \frac{\sigma}{\sqrt{n}}$$
 (6)

• Determine $z_{\alpha/2}$ in R: qnorm(p= $\alpha/2$, lower.tail=FALSE)

Confidence Interval for the Mean with Known Standard Deviation II

- $z_{\alpha/2}$ is a **quantile** of the standard normal distribution.
- 95% confidence interval for μ :

$$\left(\bar{x} - 1.96 \cdot \frac{\sigma}{\sqrt{n}}, \ \bar{x} + 1.96 \cdot \frac{\sigma}{\sqrt{n}}\right)$$
 (7)

• Can also use the functionality of the z.test() function in R.

Example 1

- Calculate the mean and a 95% confidence interval for the mean of the mpg variable in the mtcars data. Assume a known population standard deviation: $\sigma=6$.
- Calculate a 99% confidence interval for the mean.
- Hint: Rework the existing example code.

Finite Population Correction

- If our sample n is larger than 5% of the population size N we need to use a finite population correction factor (FPC).
- Otherwise, the confidence interval will be too large.
- 100(1- α) confidence interval for μ with FPC:

$$\bar{x} \pm z_{\alpha/2} \cdot \frac{\sigma}{\sqrt{n}} \cdot \sqrt{\frac{N-n}{N-1}}$$
 (8)

Example 2

- Assume the *iris* data to be our population of flowers.
- Set your seed to be 123 and draw a random sample of n=20 flowers.
- Use the population correction factor (FPC) and population standard deviation to obtain a 95% confidence interval for the *petal length* of the flowers.

Confidence Interval for the Mean with Unknown Standard Deviation I

- In practice, the population standard deviation will not be known.
- In this case we now use the quantiles from a t-distribution.
- The *t*-distribution has a similar shape to the normal distribution but has an additional *degrees of freedom* (*df*) parameter.
- ullet The t-distribution has a larger variability than the normal distribution.
 - ullet Meaning that the confidence intervals are wider than if σ is known.

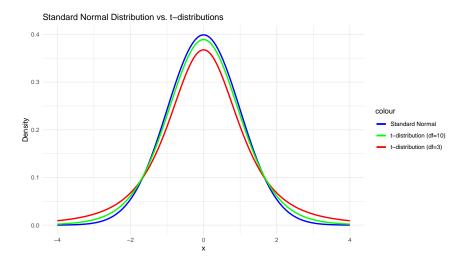
Confidence Interval for the Mean with Unknown Standard Deviation II

- Degrees of freedom (df) = n number of estimated parameters.
 - When estimating a confidence interval for the mean using a t-distribution: df = n 1.
- In this case we now use the quantiles from a t-distribution.
- 100(1- α) confidence interval for μ :

$$\bar{x} \pm t_{\alpha/2,n-1} \cdot \frac{s}{\sqrt{n}}$$
 (9)

- To find $t_{\alpha/2,n-1}$ in R: qt(p= $\alpha/2$, df=n-1, lower.tail=FALSE).
- May also use the t.test(x,conf.level = 0.95) function in R.

Distribution Plots



Example 3

- Assuming a sufficiently large population, obtain a 95% confidence interval for the petal length of the flowers in the *iris* data.
 - Hint: We do not know the population standard deviation.

Sample Proportions

- Sample proportions are a natural way to examine categorical data.
- The numerical summaries account for the individual frequencies and the relative frequencies.
 - What brand of laptops students own.
 - Own a laptop or not.
- A sample proportion reflects how many observations fall into a certain category relative to the other possible categories.
- An unbiased estimator of a population proportion π is the statistic $\hat{p} = x/n$ (sample proportion).

Confidence Intervals for Sample Proportions

- We will use quantiles of the standard normal distribution $z_{\alpha/2}$.
- 100(1- α) confidence interval for π :

$$\hat{p} \pm z_{\alpha/2} \cdot \sqrt{\frac{\hat{p}(1-\hat{p})}{n}} \tag{10}$$

- Because $\sqrt{\frac{\hat{p}(1-\hat{p})}{n}}$ is the standard error of the sampling distribution of the proportion.
- May use the prop.test(x,n,conf.level=0.95) function but the results will be slightly different.

Example 4

- Assume we have a sample from a sufficiently large population of the favourite colour of twelve (n = 12) students: red, red, blue, green, green, orange, red, orange, yellow, green, red, orange.
 - Calculate a confidence 95% confidence interval for the proportion of students that like red.

Measures of Variability

- Understanding variability is very important in the implementation of statistical theory to decision-making process.
- We have point estimates for variability (standard deviation & variance).
- It might be useful to calculate confidence intervals for the variability.
- For example: The estimated variance in food prices.

Chi-square (χ^2) Distribution

- The sampling distribution of s^2 is **not** normally distributed.
- Instead, the chi-square (χ^2) distribution is used.
- The chi-square (χ^2) distribution also relies on degrees of freedom (df) and it is **not** symmetric.
- 100(1- α) confidence interval for the variance σ^2 :

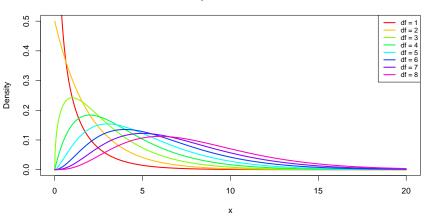
$$\left[\frac{(n-1)s^2}{\chi^2_{n-1,\alpha/2}}, \frac{(n-1)s^2}{\chi^2_{n-1,1-\alpha/2}}\right] \tag{11}$$

Chi-square (χ^2) Distribution in R

- Use the pchisq(q, df, ncp = 0, lower.tail = TRUE) function to get the chi-square quantiles.
 - Not symmetric so you need to change to lower.tail = FALSE for the upper limit.
- We can also use the VarCI(x,method = "classic",conf.level = 0.95) function from the DescTools package.

Chi-square Plots





Example 5

 Assuming a sufficiently large population, obtain a 95% confidence interval for the variance of the petal length of the flowers in the *iris* data.

Applications of Confidence Intervals

- Confidence intervals may be used to help with the decision-making process.
- Example: Government regulator mandates that water tanks must have 64 litres.
 - Take a sample of n=30 tanks and they have an average fill of $\bar{x}=63.82$
 - Is the equipment under filling the tanks?
 - 95% confidence interval: (63.43, 64.21)
- With 95% certainty we can say that the equipment fills are not different than 64 litres.

Sample Sizes for Mean Confidence Intervals

• We can see using algebra that we can determine the size of n for any given half interval length E.

$$n = \left(\frac{z_{\alpha/2} \cdot \sigma}{E}\right)^2 \tag{12}$$

- Be sure to round up to the nearest whole number.
- We can use R to obtain *n* using the *EnvStats* package:
 - ciNormN(half.width, sigma.hat = 1, conf.level = 0.95)
- This only works with random samples, we need other methods for more robust experimental designs.

Example 6

• How many observations n, drawn from a random sample will we need to obtain a half confidence interval of 0.25 from a sample with $\sigma = 3.3$?

Differences Between Means, Independent Samples I

 A confidence interval for independent samples with unequal variances is:

$$ar{x}_1 - ar{x}_2 \; \pm \; \Big(t_{lpha/2,df^*}\Big) \sqrt{rac{s_1^2}{n_1} \, + \, rac{s_2^2}{n_2}}$$

where:

$$df^* = rac{\left[rac{s_1^2}{n_1} + rac{s_2^2}{n_2}
ight]^2}{\left[rac{\left(s_1^2/n_1
ight)^2}{n_1 - 1}
ight] + \left[rac{\left(s_2^2/n_2
ight)^2}{n_2 - 1}
ight]}$$

Differences Between Means, Independent Samples II

A confidence interval for independent samples with equal variances is:

$$ar{x}_1 - ar{x}_2 \pm (t_{lpha/2,n_1+n_2-2}) s_p \sqrt{rac{1}{n_1} + rac{1}{n_2}}$$

where:

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

Differences Between Means, Paired Samples

• A confidence interval for the difference in paired samples:

$$\overline{D}\pmig(t_{n-1,\,lpha/2}ig)s_D/\sqrt{n}$$

where:

$$s_D = \sqrt{rac{\displaystyle\sum_{i=1}^n ig(D_i - \overline{D}ig)^2}{n-1}}$$

ullet $ar{D}$ is a point estimate for the mean difference between the populations.

Differences Between Proportions

 A confidence interval for differences between proportions of two populations:

$$\hat{p}_1 - \hat{p}_2 \pm z_{lpha/2} \sqrt{rac{\hat{p}_1(1-\hat{p}_1)}{n_1} + rac{\hat{p}_2(1-\hat{p}_2)}{n_2}}$$

Summary of Confidence Interval Formulas

Type of Confidence Interval	Formula
Mean, standard deviation known	$\bar{x} \pm z_{\alpha/2} (\alpha/\sqrt{n})$
Mean, standard deviation unknown	$\bar{x} \pm t_{\alpha/2,n-1} (s/\sqrt{n})$
Proportion	$\hat{\rho} \pm z_{\alpha/2} \sqrt{\frac{\hat{\rho} (1 - \hat{\rho})}{n}}$
Population total	$N\overline{x} \pm t_{\alpha/2, n-1}N\frac{s}{\sqrt{n}}\sqrt{\frac{N-n}{N-1}}$
Difference between means, independent samples, equal variances	$\overline{x}_1 - \overline{x}_2 \pm (t_{\alpha/2, n_1 + n_2 - 2}) S_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}$
	$s_p = \sqrt{\frac{(n_1 - 1)s_1^2 + (n_2 - 1)s_2^2}{n_1 + n_2 - 2}}$
Difference between means, independent samples, unequal variances	$\bar{x}_1 - \bar{x}_2 \pm (t_{\alpha/2, df^*}) \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}$
	$df^{\star} = \frac{\left[\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}\right]^2}{\left[\frac{(s_1^2/n_1)^2}{n_1 - 1}\right] + \left[\frac{(s_2^2/n_2)^2}{n_2 - 1}\right]}$
Difference between means, paired samples	$\overline{D} \pm (t_{n-1, \alpha/2})s_D/\sqrt{n}$
Differences between proportions	$\hat{\rho}_1 - \hat{\rho}_2 \pm z_{\alpha/2} \sqrt{ - \frac{\hat{p}_1 (1 - \hat{p}_1)}{n_1} } + \frac{\hat{p}_2 (1 - \hat{p}_2)}{n_2} $
Variance	$\left[\frac{(n-1)s^2}{\chi^2_{n-1,\;\alpha/2}},\;\frac{(n-1)s^2}{\chi^2_{n-1,\;1-\alpha/2}}\right]$

Figure: Source: (1)

Using R

- Confidence interval for the difference of two means (DescTools) package:
 - MeanDiffCI(x, y, method = "classic", conf.level = 0.95, paired = FALSE)
 - Set paired = TRUE for paired samples
- Confidence interval for the difference of two proportions using the functionality of the prop.test() function:
 - prop.test(x=c(\hat{p}_1 , \hat{p}_2), n=c(n_1 , n_2), correct=FALSE)

Example 7

• Calculate a 90% confidence interval for the differences in petal lengths between the *setosa* and *versicolor* flowers from the *iris* dataset.

Example 8

• In Kamloops, 25% of $(n_1 = 52)$ people interviewed said they prefer winter. In Kelowna, 35% of $(n_2 = 60)$ people interviewed said they prefer winter. Construct a 95% confidence interval for the differences in proportions.

- Given the following situations, which method should you select to obtain a confidence interval:
 - Confidence interval for the variance of annual incomes in Canada.
 - Confidence interval for the mean of a sample of bird weights with a population standard deviation $\sigma = 8.23$.
 - Confidence interval for the proportion of students that have Apple laptops.
 - Confidence interval for the mean generated by a Sample with a mean of $\bar{x} = 3.4$ and sample standard deviation s = 4.4.
 - Confidence interval for the mean heights of ADSC1000 students obtained from a sample n = 10. *Hint:* N = 35.

- Assuming a sufficiently large population, obtain a 95% confidence interval for the sepal length of the flowers in the iris data.
- Assuming a sufficiently large population, obtain a 95% confidence interval for the variance of the sepal length of the flowers in the *iris* data.

- Assume we have a sample from a sufficiently large population of the favourite colour of twelve (n = 12) students: red, red, blue, green, green, orange, red, orange, yellow, green, red, orange.
 - Calculate a confidence 95% confidence interval for the proportion of students that like orange.

• How many observations n, drawn from a random sample will we need to obtain a half confidence interval of 0.66 from a sample with $\sigma = 2.35$?

- Before taking additional lessons, students' mathematics grades were as follows: 65, 55, 78, 80, 55, 45, 65, 66, 76, 45.
- After additional training the **same** students' mathematics grades were as follows: 75, 64, 81, 90, 62, 56, 71, 70, 79, 55.
- Construct a 95% confidence interval for the differences in mathematics scores.

References & Resources

- Evans, J. R., Olson, D. L., & Olson, D. L. (2007). Statistics, data analysis, and decision modeling. Upper Saddle River, NJ: Pearson/Prentice Hall.
- ② Devore, J. L., Berk, K. N., & Carlton, M. A. (2012). *Modern mathematical statistics with applications (Second Edition)*. New York: Springer.

- BSDA
- VarCI()
- Chi-squared Distribution
- The Student t Distribution
- ciNormN()