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Statistics in Evidence Based Medicine

Lecture 4: Vocabulary of Inferential Statistics Confidence Intervals

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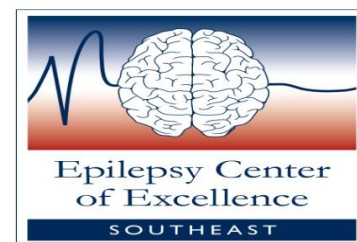
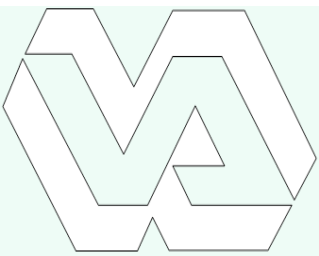
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Inferential Statistics

- The branch of statistics dealing with conclusions, generalizations, predictions, and estimations based on data from small data sets
- Answers the following questions
 - How far the computed value of a quantity of interest from a small data set be from the broader population's value?

Confidence Interval

- Is our computed statistic statistically different from the norm?

Hypothesis testing

Examples of Confidence Intervals

Table 3.1 Mean diastolic blood pressures of printers and farmers

	Number	Mean diastolic blood pressure (mmHg)	Standard deviation (mmHg)
Printers	72	88	4.5
Farmers	48	79	4.2

Q A: The sample mean of blood pressure for Printers is 88mmHg. What is the 95% C.I around this mean?

Q B: The **difference in mean blood pressures** of Printers and Farmers is 9 mmHg. What is the 95% C.I for this difference?

There is a 5% chance that the range provided by C.I will exclude the true mean blood pressure of all printers in example A and true mean difference of blood pressures between Printers and Farmers .

Examples of Hypotheses Testing

Table 3.1 Mean diastolic blood pressures of printers and farmers

	Number	Mean diastolic blood pressure (mmHg)	Standard deviation (mmHg)
Printers	72	88	4.5
Farmers	48	79	4.2

Q A: Previous studies show that the mean estimated blood pressure of all Farmers is 87mmHg. Is this sample mean (n=48) unusually low?

Null Hypothesis H_0 : $\mu=87$

Alternative Hypothesis H_A : $\mu<87$

Q B: Is there is a statistically significant difference between the blood pressure means of Printers and Farmers?

Null Hypothesis H_0 : $\mu_1 = \mu_2$

Alternative hypothesis H_A : $\mu_1 \neq \mu_2$



Confidence Intervals & Hypothesis Testing are Complimentary

- Similar terms are used in both.
- In computing confidence intervals we assume that alternative hypothesis is true while in test of significance we assume that null hypothesis is true.
- Both will lead us to the same conclusion.
- Confidence intervals give more information than a hypothesis test.



Overview

- Populations & samples
- Probability & probability distributions
- Sampling distributions
- Variation between samples: standard error
- Confidence intervals



Population vs. Sample

Population

- Aggregate of all objects under consideration
- Population parameters describe characteristics of population; generally denoted by Greek letters

Sample

- A subset of population: A well chosen sample should contain all the properties of population
- **Simple Random Sample:** Every member of population has an equal chance of being selected
- Sample characteristics (statistics) are described by Roman letters

We infer from a sample to a larger population



Symbols: Population vs. Sample

Characteristic	Population parameter	Sample statistic
mean	μ	\bar{x}
variance	σ^2	s^2
standard deviation	σ	s
proportion	π	p



Reasons for Sampling

- Samples can be studied more quickly.
- Less expensive
- Sometimes population study is impossible.
- Sample results are more accurate than population results.

Probability theory enables us to use sample statistics to estimate population parameters



Importance of a Random Sample

- Random sampling ensures that the differences from the population are solely due to chance.
- Methods of probability theory can be applied to draw inferences about population from the sample.
- Convenience sampling should ensure samples are typical of population and results can be generalized.



Probability

The probability that an event (a possible outcome of an experiment) will happen under given circumstances is the proportions of repetitions of that event in the long run.

$$\text{Pr}(\text{event}) = \frac{\text{Total number of ways event can happen}}{\text{Total number of outcomes of an experiment}}$$

Example

Experiment: A die is thrown, **Event:** A six appears

$$\text{Pr}(\text{event}) = 1/6$$



Random Variable

- The characteristic of interest in a study is called a variable.
 - Age, sex, height, blood pressure, etc
 - A variable selected through randomization process is called a random variable.
 - Values of a random variable can be summarized in a frequency distribution known as probability distribution.



Probability Distribution

- A **probability distribution** is a table or an equation that links each possible value that a random variable can assume with its probability of occurrence.

Example: Tossing a coin

Random variable: # of times a head appears

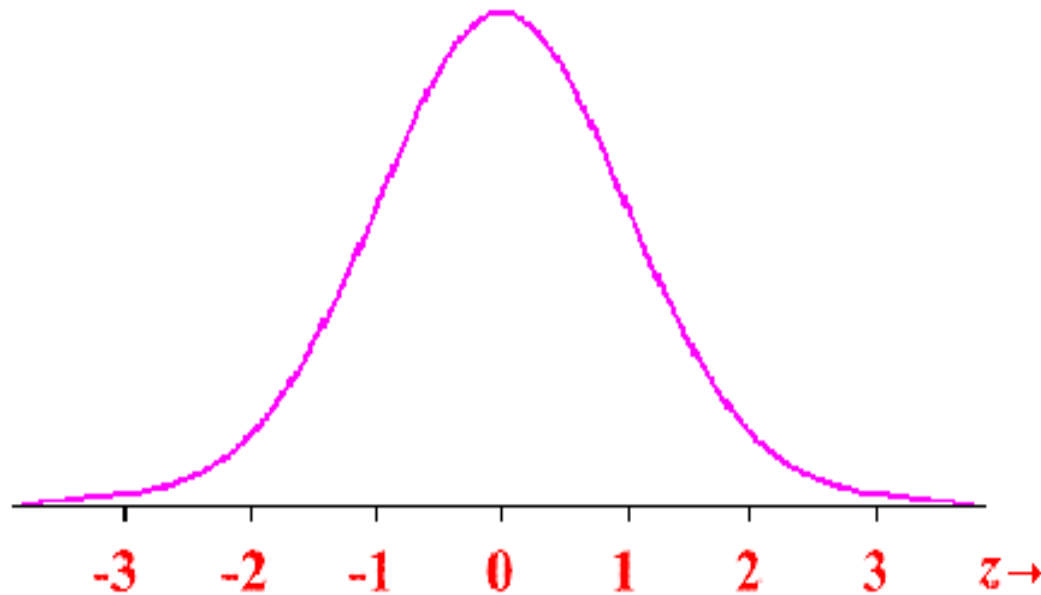
$\Pr(\text{head appears zero times}) = 1/2$

$\Pr(\text{head appears one time}) = 1/2$

- In many situations, a theoretical probability distribution is a good fit to the distribution of variable of interest.

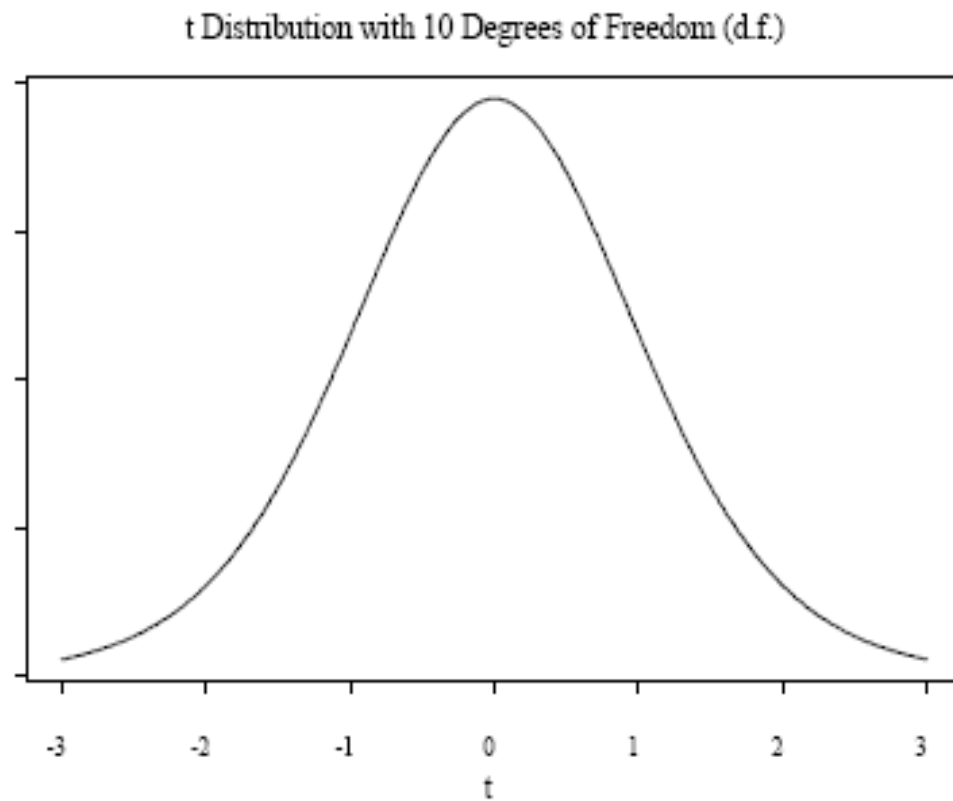
Familiar Theoretical Probability Distributions

Standard normal z distribution



Familiar Theoretical Probability Distributions

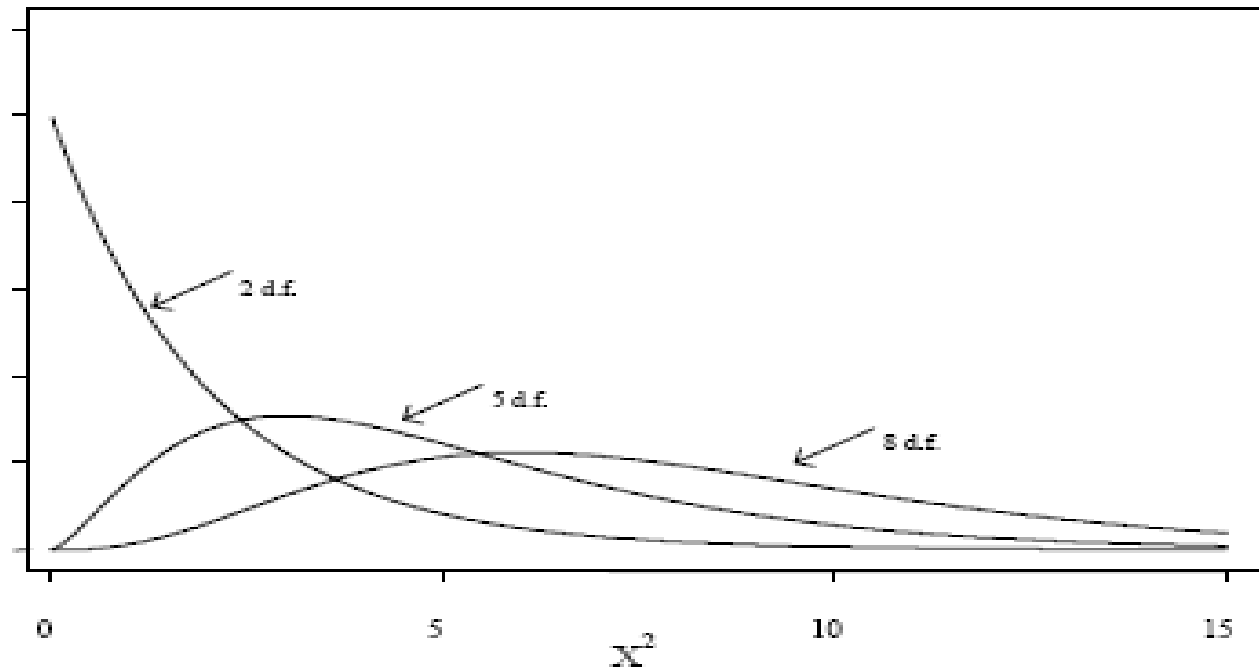
t distribution



Familiar Theoretical Probability Distributions

chi-square distribution

Chi-Square Distributions





Sampling Distribution

- Suppose all possible samples of size n are drawn from a population.
- A statistic of interest such as mean from each sample is calculated.
- The probability distribution of this series of statistics is called a sampling distribution.

The distribution of data in a sample may be similar or different from sampling distributions of statistic of interest.



Similarity of Sampling Distributions with Known Theoretical Distributions

- When the data are normal: Sampling distribution of mean is approximately normal and variance follows a chi square distribution.
- Data not normally distributed: Sampling distribution of mean still normal for large ($n \geq 30$) sample size and sampling distribution of mean follows a **t** distribution when sample size is small ($n < 30$).
- Sampling distribution of ratio of two variances assumes F distribution.



Standard Error: Variation between Samples

- Standard error of a statistic is the standard deviation of sampling distribution of the statistic. The standard error of the sample mean is $\frac{s}{\sqrt{n}}$.
- Standard error (SE) is a measure of the variability that a constant (population parameter) would be expected to show during sampling. It provides a measure of how far from the true value the estimate is likely to be.
- A population parameter usually lies within statistic ± 2 SE.



Standard Error: Estimated Standard Deviation of a Statistic

$$\text{SE for mean} = \frac{s}{\sqrt{n}}$$

$$\text{SE for proportion or percentage} = \sqrt{\frac{p(100 - p)}{n}}$$

$$\text{SE for difference between two means} = \sqrt{SE_1^2 + SE_2^2}$$

$$\text{SE for difference between two percentages} = \sqrt{\frac{p_1(100 - p_1)}{n_1} + \frac{p_2(100 - p_2)}{n_2}}$$

SE is used in computation of confidence intervals and hypothesis testing.



Confidence Intervals

An interval of values bounded by confidence limits within which the true value of a population parameter is stated to lie with a specified probability

$$CI = \text{statistic} \pm \text{critical value} \times \text{standard error}$$

critical value is a number related to the desired level of confidence. The typical values for level of confidence are 90%, 95% & 99%.



Example of a Confidence Interval

Mean diastolic blood pressure of a group
of 72 Printers = 88mmHg

Standard deviation = 4.5mmHg

Standard error=0.53mmHg

Critical value at 95% confidence level =1.96

CI =mean 1.96 0.53 =(86.96, 89.04)

Question: How to interpret this confidence interval?



Interpretation of Confidence Interval

- There is only a 5% chance that the range 86.96 to 89.04 mmHg excludes the population mean.
- In the long run if we take many samples and compute 95% confidence intervals for each sample, then 95% of these intervals will include population mean.
- In practice it is hard to determine whether our C.I. is one of 95% (that include population mean) or the 5% (that must exclude population mean).

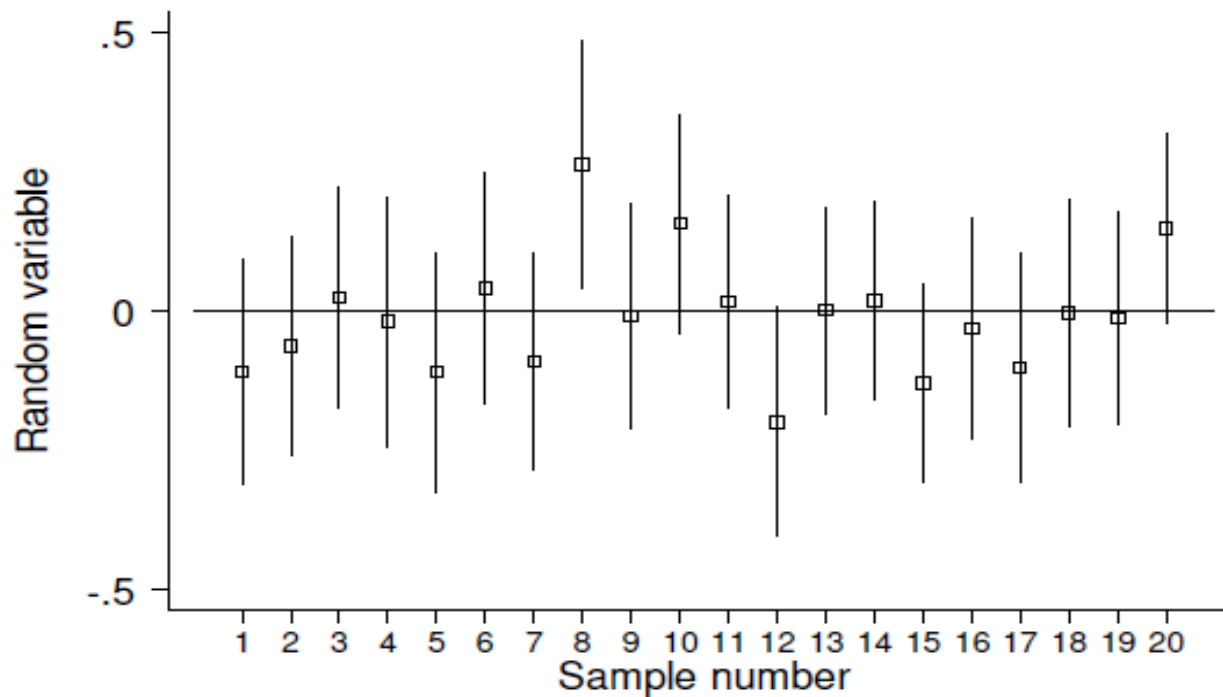


How **NOT** to Interpret Confidence Interval

- There is 95% probability that the mean of another sample lies in the range 86.96 to 89.04 mmHg.
- There is a 95% probability that the true mean lies within the range (86.96,89.04).
- If the study is repeated with same sample size, the new result falls in the confidence interval (86.96, 89.04) 95% of the time.

Another Example for Confidence Intervals

Figure 2. Simulation of 20 samples of 100 observations from a Standard Normal distribution, mean = 0 and SD = 1.0, SE = 0.10.





Reference Range for a Mean

Suppose data obtained are normally distributed.

The 95% reference range is given by

sample mean \pm 1.96 standard deviation of the sample

Mean diastolic BP of Printers was 88 mmHg with SD 4.5mmHg. Ref. Range is from 79.2 mmHg to 96.8mmHg.

Only 1 in 20 of printers would be expected to have a diastolic BP below 79 or above 97 mmHg.



Things to Remember

- Reference range and confidence interval are different.
- If a study is repeated with the same sample size, the new results DON'T fall in the confidence interval 95% of the time.
- To avoid confusion with negative numbers, it is better to quote a confidence interval as a to b or (a , b) instead of (a - b)



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Questions/Comments

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Next lecture's highlights

Hypothesis testing, p value, Type 1 & type 2 errors, power, sample size calculation