“Inference by Eye“
Standard deviation, standard error
and confidence intervals as error bars

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MERSEY POSTGRADUATE TRAINING PROGRAMME

Workshop Series: Basic Statistics for Eye Researchers and Clinicians
“Inference by eye is the interpretation of graphically presented data” - Cumming, Finch 2005

Two goals to draw a picture
• To enhance description of data
• To visualize results of statistical inference

On first seeing the figure on left, what questions should spring to our mind?
• What is the centre?
• What are the error bars i.e. the antenas in the picture?
  • 1 Standard deviation, 2 x Standard deviation (or 1.96),
  • Standard errors, Confidence interval (2 x Standard error),
  • Range i.e. min and max
Outline

- Fundamental statistical concepts
- Mean
- Standard deviation
- Normal distribution
- Standard error
- Confidence interval for the mean
- What error bars should we use?
- References
We aim to learn about population.

- Population is theoretical concept used to describe the entire group.
- Characteristics of a population can be described by quantities called parameters.

We choose sample of subjects and collect data on the sample.

- Data are used to provide estimates of population parameters.
- We describe the sample using graphs and numerical summaries. Called descriptive statistics
- We use data to make inference about the parameters of the population. Called statistical inference.
Example – population and parameter of interest

We want to investigate the mean value of mfERG of patients with diabetic maculopathy without CSMO who are 25 to 75 years old.

We randomly chose 20 patients.


Population of interest:

   People with diabetic maculopathy without CSMO who are 25 to 75 years old

Parameter of interest:

   Mean mfERG
Population

Population of size $N$

Parameters of interest:

• Population mean:

\[ \mu = \frac{\sum_{i=1}^{N} X_i}{N} \]

• Population standard deviation:

\[ \sigma = \sqrt{\frac{\sum_{i=1}^{N} (X_i - \mu)^2}{N}} \]
Sample

Random sample of size $n$

Estimates of interest:
- Sample mean: best possible estimate of population mean
- Sample standard deviation (S or SD): an estimate of population standard deviation

\[
-x = \frac{\sum_{i=1}^{n} x_i}{n}
\]

\[
s = \sqrt{\frac{\sum_{i=1}^{n} (x_i - \bar{x})^2}{n-1}}
\]

Now we know how to calculate standard deviation but how to use it in construction of the error bars? We need to know more about its interpretation.
The Normal Distribution \( N(\mu, \sigma^2) \)

- Continuous data e.g. mfERG, can be plotted using a histogram.

- Data that are said to follow the ‘Normal Distribution’ will produce a characteristic single peaked, bell-shaped histogram symmetric about the mean.
Normal distribution with mean $\mu$ and standard deviation $\sigma$

Possible values for the variable $x$ from the population
95 % reference range for the variable $X$

$N(\mu, \sigma^2)$

(95%)

$1.96\sigma$

(2.5%)  \hspace{1cm} \mu \hspace{1cm} x \hspace{1cm} (2.5%)$
Summary of main properties of the Normal distribution

• Exactly 95% of the distribution lies between

\[ \mu - 1.96 \cdot \sigma \quad \text{and} \quad \mu + 1.96 \cdot \sigma \]

• This range defines the interval within which 95% of observations in the population lie.

• Referred to as a reference range or prediction interval

• In practice, \( \mu \) and \( \sigma \) are estimated from a sample.

We can use \( 1.96 \cdot s \) as an error bar and it will indicate that 95% of subjects are within such error bars. Or we can use \( s \) as error bar and it will indicate 68% of all subjects. These error bars get more accurate with more data, and they effectively describe the data.
Example - reference range

• In our sample of 20 patients we found mean and standard deviation of mfERG:

\[ \bar{x} = 27.67 \text{ and } s = 5.3135 \]

• A 95% reference range for the mfERG for a person with DM without CSMO and age 25-75 would therefore be:

\[ 27.67 - 1.96 \times 5.3135 \text{ to } 27.67 + 1.96 \times 5.3135 \]

i.e. 17.26 to 38.08.

Approximately 95% of mfERG of 25-75 yrs patients with DM without CSMO lie in this interval.
Different sample means are obtained from different samples.

- How good is our estimate of the population mean from a single sample?
- How large an error might we make if we infer this to be the population mean?
- Sample means ($\bar{x}$) vary about the population mean ($\mu$)
- This variability can be quantified using the Standard Error of the Sample Mean (SE)

Standard error is another measure used to construct error bars. How do we calculate it and use it properly?
Standard error of the sample mean

- Standard Error of the Sample Mean (SE) is given by

\[ SE = \frac{\sigma}{\sqrt{n}} \]  

(\( \sigma \) estimated by \( s \))

- This is a measure of the precision with which the population mean is estimated

- SE increases as variability of individual observations (\( \sigma \)) increases

- SE decreases as sample size (\( n \)) increases

SE is a measure of how variable the mean will be, if you repeat the whole study many times.
Confidence interval for population mean

We wish to make inference about population mean. One solution is to find its confidence interval.

• Sample mean ($\bar{x}$) is an estimate of the unknown population mean ($\mu$)

• A confidence interval for the population mean is a range of values which we are confident (to some degree) includes the true value of the population mean
An approximate 95% confidence interval for the population mean is given by:

- estimate of the **mean** \((1.96 \times \text{SE of the mean})\)

  \[\bar{x} - 1.96 \frac{s}{\sqrt{n}}, \pm \bar{x} + 1.96 \frac{s}{\sqrt{n}}\]

This formula is exact if data are Normal, otherwise it is approximate for large enough \(n\) (due Central Limit Theorem).

We can plot error bars as \(1.96 \cdot \text{SE}\) and it gives a 95% range of values for true mean. We can be 95% confident that this range contains the true mean. With increasing sample size \(n\) this error bar gets smaller.
Central limit theorem

Sample means ($\bar{x}$) follow a Normal distribution if:

- the original observations ($x$) have a Normal distribution,

or

- the sample size is large enough (Central Limit Theorem)
Central Limit Theorem: Whatever is the distribution of the variable in the population, the distribution of the sample mean will be nearly Normal as long as the samples are large enough.
Example – confidence interval

• The mean and standard deviation of the mfERG of a random sample of 20 patients were 27.67 and 5.3135.

• The standard error of the sample mean is estimated by:

\[
\frac{5.3135}{\sqrt{20}} = 1.1881
\]

• The 95% confidence interval for the population mean is

\[(27.67 - 1.96 \times 1.1881, \ 27.67 + 1.96 \times 1.1881)\]

• i.e. (25.3, 30.0). We are 95% confident that the true population mean lies within this interval.
We say we are 95% confident that true mean is in the CI.

It means that the procedure that we used does produce a CI that contains the true mean value in 95% independent replications.

Figure: The 95% confidence intervals for the population mean for 20 independent replications of a study. Each sample has size $n=36$.

Cumming and Finch, 2005
Which error bars to use in graphs

Now we know how to construct 4 types of error bars:

- Range i.e. min to max
- Standard deviation: s
- Standard error: SE
- Confidence interval: CI

Which measure is the appropriate one when graphing data?
It depends on the question or aspect of data analysis to be illustrated.
There are 2 situations: descriptive and inferential.
We need *descriptive* error bars if we want to:

- Indicate the spread in the population.
- We have an observation for a new patient and we need to know if it is a typical value.
- We want to *describe* the data at hand.

Can we use error bars:

- Range i.e. min to max? **Yes**
- Mean+/-S? **Yes**
- Mean+/-SE? **No**
- Mean +/−1.96 SE? **No**

Holopigian and Bach, 2010
Inferential error bars

We need *inferential* error bars if our goal is:

- Indicate plausible values of the population mean i.e. to do *inference* about it.
- Compare two groups of patients i.e. to do *inference* if the two population means are different.

Can we use error bars:
- Range i.e. Min to max? **No**
- Mean+-S ? **No**
- Mean+-SE ? **Yes**
- Mean +-1.96 SE ? **Yes**

Holopigian and Bach, 2010
Example – plots of error bars

Which plot we use to see if value \( mfERG = 20 \) is typical? LEFT plot, and it is typical.

Which plot we use to illustrate a claim that the mean \( mfERG \) is less than 40? RIGHT
Summary

- Concepts
  - Population and sample
  - Population mean and sample mean
  - Descriptive statistics and inferential statistics

- They are 2 types of error bars
  - Descriptive error bars
    - Range, standard deviation
  - Inference error bars
    - Standard error and confidence interval

- When showing error bars, always describe in the figure legends what they are (e.g. s, or 1.96 SE).
Books

• Practical statistics for medical research by Douglas G. Altman
• Medical Statistics from Scratch by David Bowers

Journals’ with series on how to do statistics in clinical research

• American Journal of Ophthalmology has **Series on Statistics**
• British Medical Journal has series **Statistics Notes**

Papers on “inference by eye”

• Holopigian and Bach, A primer on common statistical errors in clinical ophthalmology, Doc. Ophthalmology, 2010
• Cumming G, Finch S, Inference by eye: confidence intervals and how to read pictures of data, Am Psychol 60, 2005
Thank you for your attention

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Slides can be found on:
http://pcwww.liv.ac.uk/~czanner/

Suggestions for topics for future workshops?