

BIOSTATISTICS

Definitions

Parametric Tests	Non-Parametric Tests (Distribution-free)
Assume population is normally, homogenous, & independently distributed.	Don't require this assumption. Observation is independent.
- \bar{x} - SD - t -test - ANOVA - Pearson Coefficient - Kolmogorov Smirnov test (Fischer's exact test & Mantel-Haenszel test are extensions of χ^2)	- Median - Mode - χ^2 - Spearman Coefficient - Mann-Whitney U test - Wilcoxon's signed rank test - Kruskal Wallis test - Friedman test.

Accuracy	Closure of value to true value.
Precision (Kappa; $\kappa = 0-1$) $\kappa = 0 \rightarrow$ no agreement, >0 & $<1 \rightarrow$ good agreement, $1 \rightarrow$ perfect agreement	Reproducibility of result (i.e, every time you repeat the test, it gives the same result)
If true value is unknown \rightarrow precision more imp than accuracy	
Bias	Systematic difference from true value.
Statistical Errors = failure of statistical test	
Type I error (α) = Probability (P value)	Rejection of correct hypothesis (false +ve).
Chance of type I error = P value	
Type II error (β)	Acceptance of incorrect hypothesis (false -ve).
Depends on statistical power of study. \uparrow statistical power \rightarrow \downarrow Type II error Power of study = $1 - \beta$ <u>Power depends on:</u> - Significance level. - Sample size. - Accuracy of measurements.	

Glossary p. 98 (Cont. variables)

FPGEE Secrets p. 73 (independent vs. depend...)

Glossary p. 99 (Discrete “)

FPGEE Secrets p. 72, 74

Important Statistical Values & Tests

n: no. P: probability

Name	Symbol	Equation/Definition
Mean	\bar{x}	$\frac{\sum x}{n}$ also, $n \times P$
Median	-	Mean of the two middle values
Mode	-	Most common occurring value
Frequency of Distribution	-	Mean – Median (Mode) <u>In graph:</u> If result: +ve → +ve skewed -ve → -ve skewed 0 → normal distribution
Range of Set	-	Highest value – Lowest value
Binomial Distribution	-	$n \times P$
Standard Deviation	SD	$\sqrt{\frac{\sum (x - \bar{x})^2}{n - 1}}$
$\bar{x} \pm SD$		
Degree of Freedom	df	$n - 1$
Degree of Freedom in Chi-square table	df	$(R - 1) \times (C - 1)$ R: no. of rows C: no. of columns
Variance	-	$(SD)^2$
Relative Standard Deviation	%RSD	$\frac{SD}{\bar{x}} \times 100$
Standard Deviation of Mean	S_m	$\frac{SD}{\sqrt{n}}$
Pearson Coefficient	r	$\frac{3 (\text{Mean} - \text{Median})}{SD}$

} With SD
↓
Precision

<p>Chi-square (Cross breaks) To improve accuracy of P value, apply “Yates’ continuity correction”</p>	<p>χ^2</p>	$\sum \frac{([E - O] - 0.5)^2}{E}$ <p>E: Expected O: Observed If $\chi^2 = \text{zero} \rightarrow$ null hypothesis $\uparrow \chi^2 \rightarrow$ more significant</p>
<p>Risk Ratio (Relative Risk) In Cohort studies</p>	<p>-</p>	<p>Risk = $\frac{\text{no.of events}}{\text{no.of ppl exposed to that event}}$</p> <p>Risk ratio = $\frac{\text{risk in treatment (exposed) gp}}{\text{risk in control (unexposed) gp}}$</p> <hr/> <p>Results: 1 = no risk > 1 = exposure \uparrow risk < 1 = exposure \downarrow risk</p> <p>*If CI (confidence interval) of risk ratio includes 1 \rightarrow statistically insignificant (& vice versa)</p>
<p>Odds Ratio In case-control studies</p>	<p>-</p>	<p>Odd = $\frac{\text{no.of times the event happen}}{\text{no.of times the event not happen}}$</p> <p>Odd ratio = $\frac{\text{odds of being exposed to risk factor}}{\text{odds in control gp}}$</p> <hr/> <p>Results: 1 = no diff in risk b/w gps > 1 = risk of event \uparrow in exposure < 1 = risk of event \downarrow in exposure</p> <p>*If CI (confidence interval) of odd ratio includes 1 \rightarrow statistically insignificant (& vice versa)</p>
<p>Absolute Risk Reduction</p>	<p>ARR</p>	<p>ARR= improvement (event) rate in ttt gp (%) – improvement (event) rate in control gp (%)</p>
<p>Number Needed to Treat</p>	<p>NNT</p>	$\text{NNT} = \frac{100}{\text{ARR}}$ <p>i.e, NNT of pts should be ttted for <u>1</u> to get benefit. (\downarrow NNT \rightarrow the better)</p>
<p>Relative Risk Reduction</p>	<p>RRR</p>	<p>% of \downarrow of risk (disease) from control gp to ttt gp</p> $\text{RRR} = \frac{\% \text{ of unimproved (diseased) pts in ctrl gp} - \% \text{ of unimproved pts in ttt gp}}{\% \text{ of unimproved pts in ttt gp}} \times 100$
<p>Number Needed to Harm</p>	<p>NNH</p>	<p>NNH= $\frac{100}{\% \text{ of pts had SE in ttt gp} - \% \text{ of pts had SE in ctrl gp}}$</p>

<p>Hazard Ratio (Cox Regression Model; Proportional Hazards Survival Model) -Estimate of life expectancy -Describes relationship b/w event (usually death) & variables (e.g. smoking).</p>	<p>HR</p>	<p>HR= $\frac{\text{hazard of event in gp 1}}{\text{hazard of event in gp 2}}$</p> <p>Hazard: chance of something harmful happening If HR= 1 → risk is same b/w 2 gps If HR= 2 → risk is double in gp 1 than gp 2</p>
<p>Correlation Coefficient +ve R^2 = as one variable ↑ the other variable is also ↑ -ve R^2 = as one variable ↓ the other variable is also ↓</p>	<p>R²</p>	<p>R² = (r)² → × 100 = % i.e, % of variation in (y) axis is related to variation in (x) axis * The closest to 1 → the strongest the correlation, whether +ve or -ve (r)</p>
<p>*Pearson Correlation Coefficient is used if normal distribution, otherwise Spearman Correlation Coefficient is used.</p>		
<p>Regression</p>	<p>-</p>	<p style="text-align: center;">$y = a + bx$</p> <p>b: regression coefficient</p>

Types of regression:

Logistic regression:

Used where each case in the sample can only belong to one of two groups (e.g. having disease or not) with the outcome as the probability that a case belongs to one group only .

Poisson regression:

Used to study waiting times or time between rare events.

Difference between Correlation and Regression:

- Correlation measures the **strength** of the association b/w variables.
- Regression **quantifies** the association.

Kaplan-Meier:

Survival analysis (life tables).

Cox proportional hazards regression model:

Used in survival analysis where the outcome is time until a certain event.

Sensitivity, Specificity, & Predictive value:

		Disease	
		Present	Absent
Test result	Positive	A	B (false +ve)
	Negative	C (false -ve)	D

	Equation	Definition
All diseased	Sensitivity = $\frac{A}{A+C}$	How often the test will be +ve if the pt really have the disease
All healthy	Specificity = $\frac{D}{D+B}$	How often the test will be -ve if the pt is really healthy
All +ve	Positive Predictive Value (PPV) = $\frac{A}{A+B}$	If the result is +ve, what is the likelihood that the pt really have the disease
All -ve	Negative Predictive Value (NPV) = $\frac{D}{D+C}$	If the result is -ve, what is the likelihood that the pt is really healthy

- **Perfect test if all = 1**
- ↓ the value → the test less useful
- All × 100 = %

Likelihood ratio (LR): $LR = \frac{\text{Sensitivity}}{1-\text{Specificiy}}$

if the test is +ve, how much more likely the pt is to have the disease than not having it.

Pharmacy Management MCQs Book:

Questions: 250, 264, 307, 324

