

# Equation Summary

Ch. 2 → 12

All finals material

## Chapter 2:

① Measure of central tendency:

1 → mean → most accurate ~ affected by outliers

2 → median → middle accuracy ~ not affected by outliers

3 → mode → least accurate ~ affected by mode in far ends.

1 → mean :  $\bar{x} = \frac{\sum x}{n}$

$$\bar{x} = \frac{\sum fx}{\sum f}$$

2 → median =  $Q_2 = n \times 50\% = n^{\text{th}}$  term

Imp → 4<sup>th</sup> term is 4<sup>th</sup> term, 4.5<sup>th</sup> term is 4<sup>th</sup> and 5<sup>th</sup> term.

3 → mode = most repeated number

\* Coefficient of variation :  $C.V = \frac{\sigma}{\mu}$

## ② Measure of spread:

1 → Range → max - min

2 → Inter-Quartile Range →  $Q_3 - Q_1$

3 → Variance →  $S^2$

4 → Standard deviation →  $S$

$$S^2 = \frac{\sum x^2}{n-1} - \frac{(\sum x)^2}{n(n-1)}$$

## ③ Quantiles, percentiles and box-and-whisker:

$$Q_1 = n \times 25\%$$

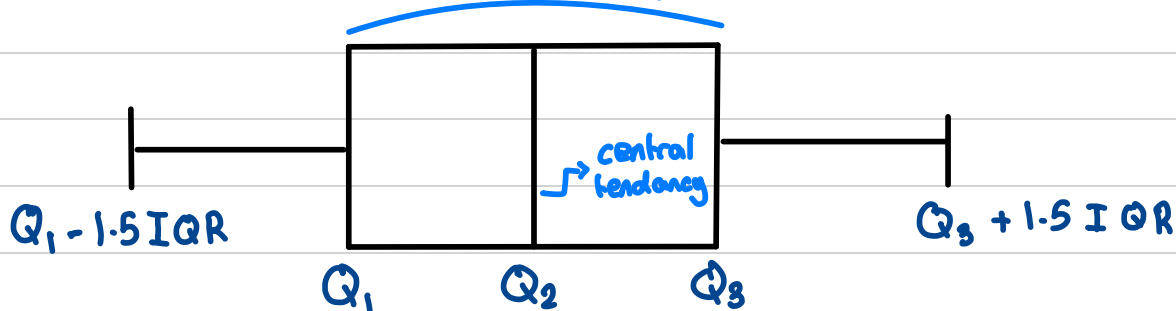
$$Q_3 = n \times 50\%$$

$$P_k = (n) \times k\%$$

If its a decimal, take next whole number.  $7.15^{\text{th}} \rightarrow 8^{\text{th}}$

If whole, take that number and next one  
 $10^{\text{th}} \rightarrow \frac{10^{\text{th}} + 11^{\text{th}}}{2}$

measure of spread



$n 25\%$

↳ decimal → next  
↳ whole → both

$n 50\%$

↳ decimal → next  
↳ whole → both

$n 75\%$

↳ decimal → next  
↳ whole → both

Outliers:

Upper extreme  $Q_3 + 1.5 IQR$

Lower extreme  $Q_1 - 1.5 IQR$

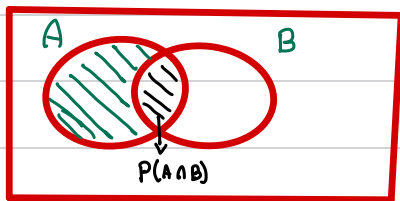
## Chapter 3:

$$\textcircled{1} P(A|B) = \frac{P(A \cap B)}{P(B)} \quad (\text{conditional probability})$$

$$\textcircled{2} P(A \cap \bar{B}) = P(A) - P(A \cap B)$$

basically, prob. of A and outside B  
because  $\bar{B} = 1 - B$ .

So:



$$\begin{aligned} \therefore P(A \cap \bar{B}) \\ \Rightarrow P(A) - P(A \cap B) \\ \text{Lgreen} \quad \quad \quad \text{Lblack} \end{aligned}$$

$\textcircled{3}$  Total probability rule:

$$P(A) = P(A|B) \times P(B) + P(A|\bar{B}) \times P(\bar{B})$$

Given that  $A \cap B$  &  $A \cap \bar{B}$  are mutually exclusive (disjoint)

#### ④ Baye's rule:

$$P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{P(B|A) \times P(A)}{P(B)}$$

$$= \frac{P(B|A) \times P(A)}{P(B|A) \times P(A) + P(B|\bar{A}) \times P(\bar{A})}$$

#### ⑤ Total probability rule:

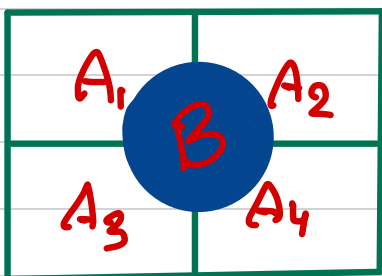
Given that events  $A_1, \dots, A_n$  are <sup>→ can't occur at same time</sup> mutually exclusive

<sup>→ at least one event MUST occur</sup>

AND exhaustive

$$P(B) = \sum_{i=1}^n P(B|A_i) P(A_i)$$

$n=4$



⑥ For any independent events:

$$P(A \cap B) = P(A) \times P(B)$$

⑦ For any two events:

$$P(A \cup B) = P(A) + P(B) - P(A \cap B)$$

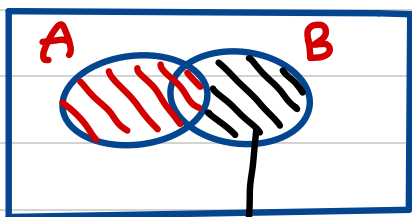
For mutually exclusive events:

$$P(A \cup B) = P(A) + P(B)$$

(because  $P(A \cap B)$  does not exist in mutually exclusive events).

⑧ For two independent events:

$$P(A \cup B) = P(A) + P(B) \times P(\bar{A})$$

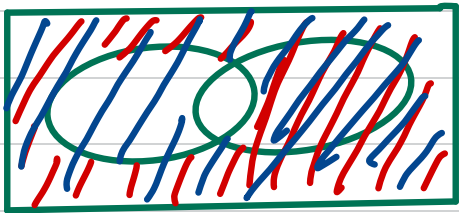


$P(B \cap \bar{A})$

$$P(A \cup B) = P(A) + P(B \cap \bar{A})$$

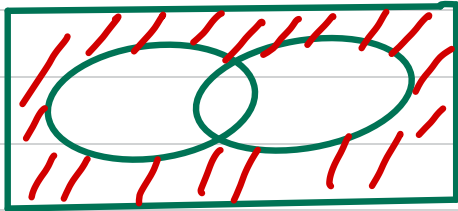
$$= P(A) + P(B) \times P(\bar{A})$$

$$\textcircled{9} P(\bar{A} \cup \bar{B}) = P(\overline{A \cap B}) = 1 - P(A \cap B)$$



Only unshaded region is  $P(A \cap B)$ .  $\therefore P(\bar{A} \cup \bar{B}) = P(\overline{A \cap B}) = 1 - P(A \cap B)$

$$\textcircled{10} P(\bar{A} \cap \bar{B}) = P(\overline{A \cup B}) = 1 - P(A \cup B)$$



$P(A \cup B)$  is not shaded  
 $\therefore P(\bar{A} \cap \bar{B}) = P(\overline{A \cup B}) = 1 - P(A \cup B)$

$$\textcircled{11} \text{ Sensitivity} = P(T^+ | D)$$

$$\textcircled{12} \text{ Specificity} = P(T^- | \bar{D})$$

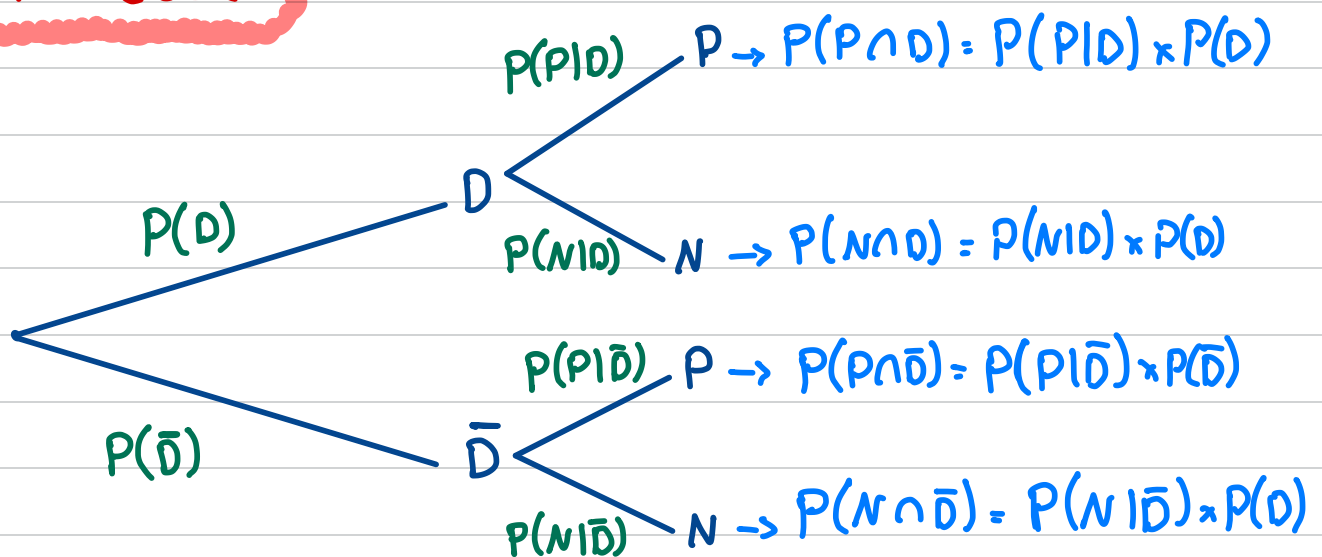
$$\textcircled{13} PV^+ = P(D | T^+)$$

$$\textcircled{14} PV^- = P(\bar{D} | T^-)$$

$$\textcircled{15} P(T^+ | D) = 1 - P(T^- | D)$$

$$\textcircled{16} P(T^- | \bar{D}) = 1 - P(T^+ | \bar{D})$$

Cheat code:



Remember that first branch is D/ $\bar{D}$  and second branch is P/N where its probability is conditional.

# Chapter 4:

## ① Conditions for Binomial distribution:

→ A sample of independent trials  $n$ .

→ Only two possible outcomes, success  $p$  or failure  $q$ .

→  $p$  and  $q$  are constant for each trial.

## ② $X \sim B(n, p) \Rightarrow P(X=a)$

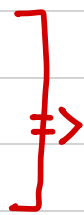
$$\therefore P(X=a) = {}^n C_a \times p^a \times q^{n-a}$$

$$(q = 1 - p)$$

## ③ Variance and standard deviation:

$$\text{Var}(x) = \sigma^2$$

$$\text{Sd}(x) = \sigma$$



$$\sigma^2 = npq$$

$$\sigma = \sqrt{npq}$$

## ④ Discrete Random variable:

$x$	1	2	3
$P(X=x)$	$a$	$b$	$c$

$$1 \rightarrow a + b + c = 1$$

$$2 \rightarrow E(x) = 1 \times a + 2 \times b + 3 \times c$$

$$3 \rightarrow \text{Var}(x) = 1^2 \times a + 2^2 \times b + 3^2 \times c$$



## Chapter 5:

$$\textcircled{1} \quad X \sim N(\mu, \sigma^2)$$

$$\Rightarrow \boxed{Z = \frac{X - \mu}{\sigma}} \quad \left\{ \begin{array}{l} P(X > a) = P\left(Z > \frac{a - \mu}{\sigma}\right) \end{array} \right.$$

$$\textcircled{2} \quad P(Z > -a) = P(Z < a)$$

$$P(Z < -a) = P(Z > a)$$

$$P(a < Z < b) = P(Z < b) - P(Z < a)$$

# Chapter 6:

Questions: They ask what's the probability of a sample mean or a sample proportion to be greater than a.  $\rightarrow \bar{x}$   
 $\hookrightarrow \hat{p}$

Or calculate confidence interval for sample mean or sample proportion.

① Standard error of the mean:

$$\text{Var}(\bar{x}) = \frac{\sigma^2}{n}$$

$$\text{S.d}(\bar{x}) = \frac{\sigma}{\sqrt{n}}$$

$\frac{\sigma}{\sqrt{n}}$  is estimated by  $\frac{S}{\sqrt{n}}$

**Note:**  $\text{Var}(\bar{x})$  &  $\text{sd}(\bar{x})$  are NOT the same as  $S^2$  &  $S$ , they are not sample variance and s.d but they are s.d and variance of the set of sample means.

② Central limit theorem:  $\bar{x} \sim N(\mu, (\sigma/\sqrt{n})^2)$   $\therefore P(\bar{x} > a)$   
 $\downarrow$   
 $P(z > \frac{a - \mu}{\sigma/\sqrt{n}})$

1  $\rightarrow$   $\sigma$  is known:  $z = \frac{\bar{x} - \mu}{\sigma/\sqrt{n}}$

2  $\rightarrow$   $\sigma$  is unknown. AND  $n \leq 30$ :  $t = \frac{\bar{x} - \mu}{S/\sqrt{n}}$  degree of freedom =  $n - 1$   
Using  $S$ .

3 →  $\sigma$  is unknown. AND  $n > 30$ :  
Using  $S$ .

$$Z = \frac{\bar{x} - \mu}{S/\sqrt{n}}$$

③ Confidence Interval of the mean:

$$\mu = \bar{x} \pm E$$

E:

1 →  $\sigma$  is known:

$$E = Z_{\frac{\alpha}{2}} * \frac{\sigma}{\sqrt{n}}$$

2 →  $\sigma$  is unknown AND  $n \leq 30$ :  
Using  $S$

$$E = t_{\frac{\alpha}{2}} * \frac{S}{\sqrt{n}}$$

3 →  $\sigma$  is unknown AND  $n > 30$ :  
Using  $S$

$$E = Z_{\frac{\alpha}{2}} * \frac{S}{\sqrt{n}}$$

Notes:

↳ To find  $Z_{\frac{\alpha}{2}}$ . Use C.I % for D. Or  $\frac{\alpha}{2}$  value for B.

↳ To find  $t_{\frac{\alpha}{2}}$ :  $1 - \text{C.I. \%} \Rightarrow \alpha \Rightarrow \frac{\alpha}{2} \Rightarrow \frac{\alpha}{2} + \text{C.I. \%} \Rightarrow t_{\frac{\alpha}{2}}$

Ex: 95% C.I  $\Rightarrow$  5%  $\alpha \Rightarrow$  2.5%  $\frac{\alpha}{2} \Rightarrow$  97.5%  $\rightarrow$  look t-dist.  $\rightarrow t_{\frac{\alpha}{2}} = 2.776$   
d.f. = 4 (example)

↳ The bigger the C.I the more the error (smaller rejection region)

↳ The bigger the  $n$ , the less the error.

④ Finding **sample size**  $n$ :

$$n = \left( Z_{\frac{\alpha}{2}} * \frac{\sigma}{E} \right)^2 \rightarrow \text{equation derived from E.}$$

$\hookrightarrow$  or  $t_{\frac{\alpha}{2}}$  depending on conditions

⑤ Point estimation of  $p$ :

$$\hat{p} \sim N\left(p, \sqrt{\frac{pq}{n}}\right)$$

$\hat{p} \Rightarrow$  sample proportion.

$p \Rightarrow$  population proportion.

$$\therefore Z = \frac{\hat{p} - p}{\sqrt{\frac{pq}{n}}}$$

(always Z-dist)

⑥ Standard Error:

S.E.:  $\sqrt{\frac{pq}{n}}$

Estimated by

$$\sqrt{\frac{\hat{p}\hat{q}}{n}}$$

$\hookrightarrow$  SE point estimate

⑦ Interval Estimate of  $p$ : (confidence interval)

$\rightarrow$  For C.I., we use  $\sqrt{\frac{\hat{p}\hat{q}}{n}}$ . For C.B.T. we use  $\sqrt{\frac{pq}{n}}$ .   
  $\rightarrow$  point estimator

$$p = \hat{p} \pm E$$

⋮

$$E = Z_{\frac{\alpha}{2}} * \sqrt{\frac{\hat{p}\hat{q}}{n}}$$

# Chapter 7:

Questions: Testing for one sample mean or population

① Type 1 & Type 2 errors:

\* Type 1 error  $\Rightarrow \alpha \Rightarrow$  Reject True  $H_0$

Mnemonic:  
 $\alpha$  RT  
(art)  $\alpha$  reject true  $H_0$

\* Type 2 error  $\Rightarrow \beta \Rightarrow$  Accept False  $H_0$

$\beta$  AF  
(Baf)  $\beta$  accept false  $H_0$

② Testing of the hypothesis:

1  $\rightarrow H_0: \theta = \theta_0$  VS \*  $H_1: \theta > \theta_0 \Rightarrow$  Right tailed

$\rightarrow$  numerical value

\*  $H_1: \theta < \theta_0 \Rightarrow$  Left tailed

\*  $H_1: \theta \neq \theta_0 \Rightarrow$  Two tailed

2  $\rightarrow$  Test Statistic

3  $\rightarrow$  p-value method OR rejection region method

### ③ Test statistic for $\mu$ :

1  $\rightarrow$   $\sigma$  is known: 
$$z = \frac{\bar{x} - \mu}{\sigma/\sqrt{n}}$$

2  $\rightarrow$   $\sigma$  is unknown. AND  $n \leq 30$ :  
Using  $S$ . 
$$t = \frac{\bar{x} - \mu}{S/\sqrt{n}}$$
 degree of freedom =  $n - 1$

3  $\rightarrow$   $\sigma$  is unknown. AND  $n > 30$ :  
Using  $S$ . 
$$z = \frac{\bar{x} - \mu}{S/\sqrt{n}}$$

### ④ Test statistic for $p$ :

$$\therefore z = \frac{\hat{p} - p}{\sqrt{\frac{pq}{n}}}$$

$$\therefore z_{\text{corr}} = \frac{|\hat{p} - p| - \frac{1}{2n}}{\sqrt{\frac{pq}{n}}}$$

## ⑤ p-value method:

Use test stat to find p-value

\*  $p\text{-value} > \alpha$  Accept  $H_0$  & Reject  $H_1$

\*  $p\text{-value} < \alpha$  Reject  $H_0$  & Accept  $H_1$



\* Reject  $H_0 \Rightarrow$  Due statistically significant

## ⑥ Rejection region method:

Use  $\alpha$  to find critical value. Use test statistic to find z-value. If z-value in rejection region then reject  $H_0$  else accept  $H_0$ .

# Chapter 8:

Questions: Two samples. Test hypothesis. For means only.

① Two Dependent Samples: (same  $n$ )

1 →  $d = \text{after} - \text{before}$

2 →  $\bar{d} = \frac{\sum d}{n}$

$$d.f. = n - 1$$

3 →  $S_d^2 = \frac{\sum d^2}{n-1} - \frac{(\sum d)^2}{n(n-1)}$

Test Statistic :

$$t = \frac{\bar{d}}{S_d / \sqrt{n}} \quad n \leq 30$$

$$z = \frac{\bar{d}}{S_d / \sqrt{n}} \quad n > 30$$

note that test stat is similar to one sample mean, but  $d = 0$  so we don't write it.

C.I ⇒

$$d = \bar{d} \pm E \quad \vdots \quad E = t_{\frac{\alpha}{2}} * \frac{S_d}{\sqrt{n}}$$



② Two independent samples:

Test Statistic:

$$t = \frac{\bar{X}_1 - \bar{X}_2}{S_p \cdot \sqrt{\frac{1}{n} + \frac{1}{m}}}$$

degree of freedom =  $n + m - 2$   
✈ 🇺🇸 🇩🇪

C.I :  $\mu_1 - \mu_2 = (\bar{X}_1 - \bar{X}_2) \pm E$

$$E = t_{\frac{\alpha}{2}} * S_p * \sqrt{\frac{1}{n} + \frac{1}{m}}$$

Pooled Variance,  $S_p \Rightarrow \sqrt{\frac{S_1^2(n-1) + S_2^2(m-1)}{n + m - 2}}$

# Chapter 10:

Questions: Testing a hypothesis for two-sample proportion.  $\rightarrow \hat{P}_1, \hat{P}_2$

① Normal Theory method,  $z$ :

Conditions:  $* n p^* q^* > 5$  &  $m p^* q^* > 5$

\* Sample size is large, samples discrete and independent

①  $H_0: P_1 - P_2 = 0$  Vs  $H_1: P_1 - P_2 \neq 0$

$\rightarrow$  two-tailed  
(can also be right-tailed or left-tailed but  $\chi^2$  is always right-tailed)

② Test stat<sub>1</sub>:

$\hookrightarrow$  no correction

$$Z = \frac{(\hat{P}_1 - \hat{P}_2) - (P_1 - P_2)}{\sqrt{p^* q^*} * \sqrt{\frac{1}{n} + \frac{1}{m}}}$$

$\downarrow$   $q^* = 1 - p^*$

$\downarrow$  Is always zero.

Test Stat<sub>2</sub>:

$\hookrightarrow$  with correction

$$Z_{\text{corr}} = \frac{|\hat{P}_1 - \hat{P}_2| - \left(\frac{1}{2n} + \frac{1}{2m}\right)}{\sqrt{p^* q^*} * \sqrt{\frac{1}{n} + \frac{1}{m}}}$$

pooled proportion,  $p^* = \frac{x + y}{n + m}$

$$\hat{P}_1 = \frac{x}{n}$$

$$\hat{P}_2 = \frac{y}{m}$$

③ Either p-value method or rejection region method

## ② Contingency Table:

$$* \chi^2 = \sum \frac{(O - E)^2}{E}$$

$$* \chi^2_{\text{corr}} = \sum \frac{(|O - E| - 0.5)^2}{E}$$

note that we mainly use  $\chi^2_{\text{corr}}$  for  $2 \times 2$ .

$$\chi^2 = \frac{(O_{11} - E_{11})^2}{E} + \frac{(O_{12} - E_{12})^2}{E} + \frac{(O_{21} - E_{21})^2}{E} + \frac{(O_{22} - E_{22})^2}{E}$$

Observed (given):

	1	2	
A	x	y	$R_1$ $x + y$
B	z	u	$R_2$ $z + u$

$$x + z$$

$C_1$

$$y + u$$

$C_2$

$$(x + y) + (z + u)$$

or  
 $(x + z) + (y + u)$

$T$

Expected

	1	2	
A	$\frac{C_1 \times R_1}{T}$	$\frac{C_2 \times R_1}{T}$	
B	$\frac{C_1 \times R_2}{T}$	$\frac{C_2 \times R_2}{T}$	

$$\therefore E = \frac{R \times C}{T}$$

Degree of freedom:  $(r-1) \times (c-1)$

$\uparrow$  no of rows       $\uparrow$  no of columns

Notes:

\*  $H_0$ : independent variables. Two variables NOT associated

\*  $H_1$ : dependent variables. Two variables ARE associated.

\* For  $2 \times 2$  table, all E must be greater than 5

\* For  $R \times C$  table,

Always the expected value is greater than 1.  
 For every 5 cells, only one value of E less than 5 is allowed.

Ex:

5	9	121
1	17	2

Doesn't Satisfy  
 $\leftarrow$

5	9	121
7	17	2

Satisfy  
 $\leftarrow$

\* Always take right tailed test

### ③ Chi-squared goodness of fit test:

Probability  $\times$  Observed total value = Expected value

1  $\rightarrow$  Continuity correction

2  $\rightarrow$  Find probability for cell

3  $\rightarrow$  multiply probability with  $O_t$ , get expected

4  $\rightarrow$  find expected for all cells.

5  $\rightarrow$  Chi-squared test. Accept or reject  $H_0$ .

$\rightarrow$  don't forget continuity correction

$$* P(x < a) \times O_t$$

Degree of freedom for Chi-squared goodness-of-fit:

$$d.f. = g - k - 1$$

$g$ : number of groups / categories

$\hookrightarrow$  If  $E < 5$  then for this category we join it with another category and it counts as one.

$k$ : number of estimated parameters

$\hookrightarrow$  point estimates  $\Rightarrow \bar{x}, s, \hat{p}$

# Chapter 11:

Questions: When they talk about **Correlation / correlation coefficient**

① Covariance: (not very important)

$$\text{Cov}(x, y) = E((x - \mu_x)(y - \mu_y)) = E(xy) - \mu_x \mu_y$$

② **Correlation coefficient**

population correlation coefficient =  $\rho$

$$\rho = \frac{\text{Cov}(x, y)}{\sigma_x \sigma_y} \quad (\text{equation not very important})$$

Sample correlation coefficient =  $r$

**Formula sheet**

$$r = \frac{L_{xy}}{\sqrt{L_{xx} L_{yy}}} = \frac{S_{xy}}{S_x \cdot S_y}$$

$r$   
↓  
 $\rho$  point estimator

$$L_{xx} = \sum x^2 - \frac{(\sum x)^2}{n}$$

$$L_{yy} = \sum y^2 - \frac{(\sum y)^2}{n}$$

$$L_{xy} = \sum xy - \frac{(\sum x)(\sum y)}{n}$$

**Formula sheet**

\*  $-1 \leq r \leq 1$  → closer to 1/-1 then stronger correlation

↳ 1 ⇒ +ve correlation  
-1 ⇒ -ve correlation

### ③ Statistical inference for hypothesis testing: Part 1

$$H_0: \rho = 0 \quad \text{Vs} \quad H_1: \rho \neq 0$$

Test Stat = 
$$t = \frac{r \sqrt{n-2}}{\sqrt{1-r^2}}$$
 In formula sheet.

### ④ Statistical inference for hypothesis testing: Part 2

$$H_0: \rho = \rho_0 \quad \text{Vs} \quad \rho \neq \rho_0$$

$\rho_0$  A non-zero number

Test stat:

1  $\rightarrow z = \frac{1}{2} \ln\left(\frac{1+r}{1-r}\right)$  Formula sheet

2  $\rightarrow z = \frac{1}{2} \ln\left(\frac{1+\rho}{1-\rho}\right)$  Get  $\rho$  from statistical test

3  $\rightarrow \lambda = (z - z_0) \sqrt{n-3}$  Formula sheet

note that the value of  $\lambda$  is the value that you must compare with critical value on a z-distribution table.

## ⑤ Confidence Interval for population correlation coefficient and fisher's z-transformation:

① Find  $(z_1, z_2)$   
↳ fisher's

② Find  $(\rho_1, \rho_2)$   
↳ pop. corr. coeff.

①  $z \pm E$

↳  $z = \frac{1}{2} \ln\left(\frac{1+r}{1-r}\right)$       ↳  $E = z_{\frac{\alpha}{2}} * \frac{1}{\sqrt{n-3}}$

$$\therefore z_{(1,2)} = z \pm z_{\frac{\alpha}{2}} * \frac{1}{\sqrt{n-3}}$$

②  $\rho = \frac{e^{2z} - 1}{e^{2z} + 1}$

\*  $\rho_1 = \frac{e^{2z_1} - 1}{e^{2z_1} + 1}$

\*  $\rho_2 = \frac{e^{2z_2} - 1}{e^{2z_2} + 1}$

C.I. =  $(\rho_1, \rho_2)$



# Chapter 12:

Questions: When they ask for testing of 3 or more samples.

① Hypothesis testing of more than 2 samples using one way Anova Model:

1 →  $H_0: \mu_1 = \mu_2 = \dots = \mu_k$  vs  $H_1: \mu_1 \neq \mu_2 \neq \dots = \mu_k$   
↳ at least one pair not equal.

2 → Test Statistic =  $F = \frac{MS_B}{MS_W}$

Formula sheet  
↓

$$* MS_B = \frac{SS_B}{k-1}$$

→

$$SS_B = \sum n\bar{y}^2 - \frac{(\sum ny)^2}{N}$$

Degree of freedom →  $k-1$

↳  $N$  is total sample size

// ↓ //

$$* MS_W = \frac{SS_W}{N-k}$$

→

$$SS_W = \sum (n-1)S^2$$

Degree of freedom →  $N-k$

## ② ANOVA notes:

\* D.F numerator =  $k-1$

$$\frac{MS_B}{MS_W}$$

\* D.F denominator =  $N-k$

\*  $SS_T = SS_B + SS_W$  → didn't see any questions in this.

## ③ Least Significant Difference test (LSD):

↳ LSD is used to see which means are not equal to each other

↳ We compare two means individually, so we use the equation from chapter 8 for two independent samples.

Test Stat:

$$t = \frac{\bar{y}_1 - \bar{y}_2}{S.p. \cdot \sqrt{\frac{1}{n} + \frac{1}{m}}}$$

↳ the  $MS_W$  is used for all groups throughout so  $S.p.$  is the same for any two means.

$$S.p. = \sqrt{MS_W}$$

$$d.f = N - k$$

$$MS_W = \frac{SS_W}{N-k} = \frac{\sum (n_i - 1) \cdot s_i^2}{N-k}$$

Example:

$$H_0: \mu_1 = \mu_2$$

Vs

$$H_1: \mu_1 \neq \mu_2$$

$$\textcircled{2} \text{ Test stat } \Rightarrow t = \frac{\bar{y}_1 - \bar{y}_2}{S.p \sqrt{\frac{1}{n} + \frac{1}{m}}} \quad S.p = \sqrt{MS_w}$$

$\textcircled{3}$  Use d.f and  $\alpha$  to find critical values.

Thus either reject or accept  $H_1$ .

" لَا يُكَلِّفُ اللَّهُ نَفْسًا إِلَّا وُسْعَهَا "