

(a) Two sides test:

$$\text{reject } H_0 \text{ if } z = \left| \frac{\bar{x} - \mu_0}{\sigma/\sqrt{n}} \right| > z_{0.025} = 1.96$$

$z = 1.8$ , so we accept  $H_0: \mu = 0$  at the 5% significance level.

(b) One side test:

$$\text{reject } H_0 \text{ if } z = \frac{\bar{x} - \mu_0}{\sigma/\sqrt{n}} > z_{0.05} = 1.645$$

Since  $z = 1.8$  we reject  $H_0: \mu = 0$  at the 5% significance level.

(c) from a: we reject  $H_0$  if

$$|\bar{x}| > \mu_0 + z_{0.025} \frac{\sigma}{\sqrt{n}} = 0 + 1.96 \times \frac{6}{\sqrt{8}} = 0.49$$

$$\text{Power} = P(\text{Reject } H_0 \mid \mu = \mu_1)$$

$$\mu_1 = 1: P(|\bar{x}| > 0.49) \sim N(\mu_1, \frac{\sigma^2}{n})$$

$$= 1 - P(|\bar{x}| \leq 0.49) \sim N(-1, \frac{1}{16})$$

$$= 1 - P\left(\frac{-0.49 - (-1)}{1/4} < Z < \frac{0.49 - (-1)}{1/4}\right)$$

$$= 1 - P(2.04 < Z < 5.96) = 0.9793 \quad \text{Z} \sim N(0, 1)$$

Similarly:

$$\mu = -1/2 \quad \text{Power} = 1 - P(0.04 < Z < 2.96) = 0.5160$$

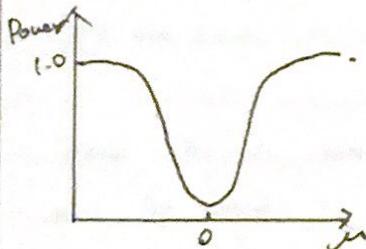
$$\mu = 1/2 \quad \text{Power} = 1 - P(-3.96 < Z < -0.04) = 0.5100$$

$$\mu = 1 \quad \text{Power} = 1 - P(-5.96 < Z < -2.04) = 0.9793$$

$$\text{at } \mu = 0 \quad \text{Power} = \alpha = 0.05$$

$$\text{as } \mu \rightarrow -\infty \quad \text{Power} \rightarrow 1$$

$$\text{as } \mu \rightarrow \infty \quad \text{Power} \rightarrow 1$$



Summary  
Power

$P(\text{reject } H_0 \mid H_0 \text{ false}) = 1 - \beta$  而当待检测的参数存在差异时，使用的方法才能能够检测到这种差异的概率。

$$P(\text{Type I error}) = P(\text{reject } H_0 \mid H_0 \text{ true}) = \alpha$$

$$P(\text{Type II error}) = P(\text{accept } H_0 \mid H_0 \text{ false}) = \beta$$

1. true.

5

Sta

(17) In a medical study of patients given a drug and a placebo, sixteen patients were paired up by allocating people of the same sex and similar age to each pair. One of each pair received a drug and one other the placebo. The response score  $\delta$  for each patient was found.

Pair number 1 2 3 4 5 6 7 8

Given number 0.16 0.97 1.57 0.55 0.62 1.12 0.68 1.69

Given placebo -0.11 0.13 0.77 1.19 0.46 0.41 0.40 1.28

Difference 0.05 0.84 0.08 -0.64 0.16 0.71 0.08 0.41

Are the responses for the drug and the placebo significantly different?

Match Pairs: Model:  $\delta_i \sim N(\mu_{\delta_i}, \sigma_{\delta_i}^2)$

$$\bar{\delta} = 0.326 \quad S_{\delta}^2 = 0.2481 \quad S_{\delta} = 0.499 \quad n=8.$$

$$H_0: \mu_{\delta} = 0 \quad v/s \quad H_1: \mu_{\delta} \neq 0$$

Reject  $H_0$  if  $|t| > t_{\alpha/2}$

$$|t| = \left| \frac{\bar{\delta}}{S_{\delta}/\sqrt{n}} \right| = \left| \frac{0.326}{0.499/\sqrt{8}} \right| = 1.88 \quad t_{0.05} = 1.89$$

so do not reject  $H_0$  at 10% level.

Summary

5

Sta

Test for existence of regression

- (18) For each one of seven cars the weight  $x$  (lbs) and fuel consumption  $y$  (gallon/100 miles) are given below:

Weight  $x$ : 3600 3800 4100 2200 2600 2900 2000

Fuel Consumption,  $y$ : 4.9 6.5 3.3 3.6 4.6 2.9

Note that  $\sum x_i = 21000 \quad \sum x_i^2 = 66820000$

$$\sum x_i y_i = 103530 \quad \sum y_i = 32.3$$

$$\bar{y} = 160.73 \quad n = 7$$

Obtain the regression line for fuel consumption given weight and perform a test for the existence of regression.

$$1. S_x^2 = \frac{1}{n-1} (\sum x_i^2 - \frac{1}{n} (\sum x_i)^2) = \frac{1}{6} (66820000 - \frac{(21000)^2}{7}) = 636666.67$$

$$S_y^2 = \frac{1}{n-1} (\sum y_i^2 - \frac{1}{n} (\sum y_i)^2) = \frac{1}{6} (160.73^2 - \frac{32.3^2}{7}) = 1.9431$$

$$S_{xy} = \frac{1}{n-1} (\sum x_i y_i - \frac{1}{n} (\sum x_i)(\sum y_i)) = \frac{1}{6} (103530 - \frac{21000 \cdot 32.3}{7}) = 1105.$$

$$\hat{\beta}_1 = \frac{S_{xy}}{S_x^2} = 0.0017356.$$

$$\hat{\beta}_0 = \bar{y} - \hat{\beta}_1 \bar{x} = 160.73 - 0.0017356 \times 3000 = -0.5925$$

so the equation of the fitted line is

$$Y = -0.5925 + 0.0017356 X$$

Test for existence of regression:  $100\alpha\%$  level

$H_0: \beta_1 = 0$  vs  $H_1: \beta_1 \neq 0$  reject  $H_0$  if

$$|t| = \left| \frac{\hat{\beta}_1}{\hat{\sigma}_{\beta_1}} \right| > t_{n-2, \alpha/2}$$

$$t_{5, 0.025} = 2.571$$

Summary

5

$$\hat{\sigma}^2 = \frac{(n-1)}{(n-2)} [S_y^2 - \hat{\beta}_1^2 S_x^2] = 0.036317$$

$$|t| = \sqrt{\frac{0.0017356}{0.036317/6 \times 63666.6}} = 17.80 \gg 2.6 \text{ or } 2.57$$

Very strong evidence that the true slope is non-zero.

Sta

- ① The following data give the frequency distribution of the size of casual groups of people on a spring afternoon in a park

Size of group	1	2	3	4	5	6
---------------	---	---	---	---	---	---

Frequency	1486	694	195	37	10	1
-----------	------	-----	-----	----	----	---

Expected Frequency  $1521.6 \quad 668.7 \quad 198.4 \quad 44.1 \quad 7.9 \quad 1.3 \Rightarrow \sum = 2423$

A suggest model for the probability  $\Pr_r$  of a group size  $r$  is  $r$  is infinite.

$$\Pr_r = \frac{\mu^r e^{-\mu}}{r! (1-e^{-\mu})}, r = 1, 2, 3, \dots$$

\*: The use of the  $\chi^2$  dist. is only valid for large samples.

$\chi^2$  is valid when all  $e_i \geq 5$ ; if some  $e_i < 5$ , then collect some groups together so that all expected frequency are  $> 5$ .

Summary

where  $\mu$  is estimated to be 0.89 for this data set. Does this give a good fit to the data?

Total number of group:  $1486 + 694 + 195 + 37 + 10 + 1 = 2423$

Combine 5-6 together

Degrees of freedom:

$$5 - 1 - 1 = 3$$

Groups Constraint Estimate  $\mu$

$$\chi_{\text{obs}}^2 = \frac{(1486 - 1521.6)^2}{1521.6} + \dots + \frac{(11 - 5.2)^2}{5.2} = 2.712$$

reject  $H_0$  if  $\chi_{\text{obs}}^2 > \chi_{3, 0.05}^2 = 7.815$  at 5%.

so we could not reject  $H_0$  at 5% level.

It's a good fit to the data.

Sta.

(179) In order to test the lifetime of  
Shall batteries used to power clocks. 40  
batteries were chosen at random  
and tested. Their times (in months) to  
failure were:

18 11 25 36 40 72 33 51 1 12

46 28 87 75 24 11 23 13 45 2

40 29 14 59 1 7 39 54 16 3

8 2 52 20 9 6 7 26 31 38

The manufacturer claims that the lifetimes  
have an exponential dist. with mean 30 months.

If we assume this, calculate a, b, c and d  
s.t.

$$\begin{aligned} P(a < X < b) &= P(a < X < b) = P(b < X < c) = P(c < X < d) \\ &= P(d < X < \infty) = \frac{1}{5} \end{aligned}$$

Construct a table of expected and observed frequency  
for the above five intervals and hence test  
the manufacturer's claim by using a goodness  
of fit test at 5% level.

$X \sim \text{Exponential}$

def:  $f(x) = 1 - e^{-\lambda x} = p$ .  $\therefore \lambda = \frac{1}{30}$   
 $\lambda = 1/30$  we take  $p = \frac{1}{5}, \frac{2}{5}, \frac{3}{5}, \frac{4}{5}$   
 $\Rightarrow x = -\lambda \ln(1-p) = -30 \ln(1-p)$

$$\therefore a = 6.7 \quad b = 15.3 \quad c = 22.5 \quad d = 48.3$$

Summary

count the number.	obs.	0-6.7	6.7-15.3	15.3-22.5	22.5-48.3	48.3+
	$\Rightarrow$	6	9	7	10	8

$$\text{Exp. } \frac{1}{5} \times 40 = 8 \quad 8 \quad 8 \quad 8 \quad 8$$

$$X^2 = \sum \frac{(O_i - E_i)^2}{E_i} = 1.25$$

5

degree of freedom:  $5 - 1 = 4 \quad \chi^2_{10, 0.05} = 9.488$

so don't reject  $H_0$ . no reason to support that the manufacturer's claim is incorrect.

Change of basis.  
LM.

Procedure of finding  
 $M_{B_1}^{B_2}$

1. Form  $[I \ B_1' \ | \ B_2]$

2. row operation

3.  $[I \ 2 \ | \ M_{B_1}^{B_2}]$

truth: 这个做法严格意义上存在问题. 反着用

$e_1, e_2$  表示  $B_2$  的基  
才是  $B_1 \rightarrow B_2$  的:

$B_1 \xrightarrow{M_{B_1}^{B_2}} B_2$

$\begin{matrix} \leftarrow \\ M_{B_2}^{-1} \\ B_1 \end{matrix}$  逆矩阵可  
以变换回去.

Summary

(18)  $B_1 = (e_1, e_2) \quad e_1 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}, e_2 = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$

$$B_2 = (c_1, c_2) \quad c_1 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}, c_2 = \begin{pmatrix} 1 \\ -1 \end{pmatrix}$$

$v = (2, 3)$  write the coordinates of  $v$

with respect to basis  $B_2$ .

① Find  $M_{B_1}^{B_2}$

② Find  $Mv$ .

Write as linear combination:

$$e_1 = \begin{pmatrix} 1 \\ 0 \end{pmatrix} = k_1 c_1 + k_2 c_2 = \frac{1}{2} c_1 + \frac{1}{2} c_2$$

$$e_2 = \begin{pmatrix} 0 \\ 1 \end{pmatrix} = k'_1 c_1 + k'_2 c_2 = \frac{1}{2} c_1 + (-\frac{1}{2}) c_2 \quad \text{顺序反了.}$$

$$M_{B_1}^{B_2} = \begin{pmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & -\frac{1}{2} \end{pmatrix}$$

→ 该线性师上课讲错了

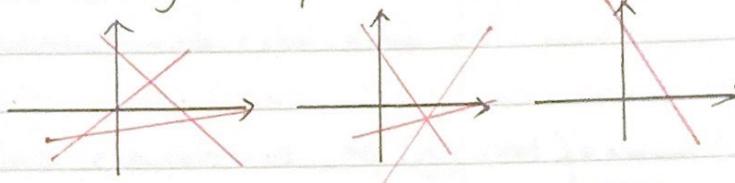
$$Mv = \begin{pmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & -\frac{1}{2} \end{pmatrix} \begin{pmatrix} 2 \\ 3 \end{pmatrix} = \begin{pmatrix} \frac{5}{2} \\ -\frac{1}{2} \end{pmatrix}$$

$$v = \frac{5}{2} c_1 - \frac{1}{2} c_2 = 2e_1 + 3e_2$$

6 . . . . .

Consistency of linear system.

Geometrically interpretation of consistency.



no solution      unique solution      infinite solutions

**Theorem:** A linear system is consistent if

and only if the row echelon form of its

augmented matrix contains no row of

the form  $[0 \dots 0 | b]$  where  $b \neq 0$ .

$$\begin{array}{c} \text{L} \\ \backslash \end{array} \xrightarrow{Ax=b}$$

consistent

inconsistent

/ \

unique      infinite  
solution      solution

**Proposition:** Any homogeneous system  $Ax=0$

$$\#\{\text{variables}\} = \#\{\text{pivot position of } A\} + \#\{\text{free variables}\}$$

**Theorem:** If  $A$  is a matrix with  $n$  columns

then  $\text{rank}(A) + \text{nullity}(A) = n$ .

$$[\text{number of leading variables}] + [\text{number of free variables}] = n.$$

判断出 consistent ←  
之后，观察有无  
free variables.

Summary

5

IM.

Rank-nullity theorem

$$(18) A = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix}$$

(i) find  $\text{Ker}(A)$ ,(ii) find all solutions to  $Ax = \begin{pmatrix} 0 \\ 0 \\ 6 \end{pmatrix}$ , given that

$$x_0 = \begin{pmatrix} 1 \\ -1 \\ 1 \end{pmatrix} \text{ is one such solution.}$$

$$\dim(\text{Ker}(A)) = \text{nullity } A$$

$$\dim(A^T) = \text{rank}$$

(iii) Verify ran-nullity for  $A$ :

$$\dim(\text{Ker}(A)) + \dim(\text{Ran}(A)) = n.$$

(i) Gaussian elimination:  $Ax = 0$ 

$$\text{rank}(A) + \text{nullity}(A) = n$$

$$\begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & -1 \\ 0 & 1 & 2 \\ 0 & 0 & 0 \end{pmatrix}$$

$$B' - 2B^2 + B^3 = 0 \quad \begin{pmatrix} 1 & 0 & -1 \\ 0 & 1 & 2 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} \alpha \\ \beta \\ \gamma \end{pmatrix} = \begin{pmatrix} \alpha=2 \\ \beta=-2\gamma \\ \gamma=\gamma \end{pmatrix}$$

$$x = \begin{pmatrix} 1 \\ -1 \\ 1 \end{pmatrix}$$

$$(ii) \text{ we can get } x = \begin{pmatrix} 1 \\ -1 \\ 1 \end{pmatrix} + t \begin{pmatrix} 1 \\ -1 \\ 1 \end{pmatrix} \in \mathbb{R}$$

Theorem: solving general linear equations via  
translation. If  $Ax = b$  if  $Ax_0 = b$ , then the general  
solution is  $x = x_0 + v$ ,  $v \in \text{Ker}(A)$ .

proof:  $Ax = b \Leftrightarrow Ax - Ax_0 = b - b = 0$

$$\Leftrightarrow A(x - x_0) = 0$$

$$\Leftrightarrow x - x_0 \in \text{Ker}(A)$$

$$\Leftrightarrow x = x_0 + v, v \in \text{Ker}(A).$$

(iii) from (i)  $\dim(\text{Ker}(A)) = 1$  the number of pivot  
columns in reduced echelon form  $\begin{pmatrix} 1 & 2 & -1 \\ 0 & 1 & 2 \\ 0 & 0 & 0 \end{pmatrix} \rightarrow \text{rank } 2$   
reduced echelon form.

Summary

5

LM.

$$\textcircled{18a} \quad W = \left\{ \begin{pmatrix} \alpha & 0 & 0 \\ 0 & \beta & 0 \\ 0 & 0 & \gamma \end{pmatrix} \mid \begin{array}{l} \alpha, \beta, \gamma \in \mathbb{R} \\ \alpha + \beta + \gamma = 0 \end{array} \right\} \subseteq M_{3,3}(\mathbb{R})$$

&amp; \text{Trace}(A) = 0

$$V_1 = \begin{pmatrix} \alpha_1 & 0 & 0 \\ 0 & \beta_1 & 0 \\ 0 & 0 & \gamma_1 \end{pmatrix}, \quad V_2 = \begin{pmatrix} \alpha_2 & 0 & 0 \\ 0 & \beta_2 & 0 \\ 0 & 0 & \gamma_2 \end{pmatrix}$$

$$V_1 + kV_2 = \begin{pmatrix} \alpha_1 + k\alpha_2 & 0 & 0 \\ 0 & \beta_1 + k\beta_2 & 0 \\ 0 & 0 & \gamma_1 + k\gamma_2 \end{pmatrix} = \begin{pmatrix} \alpha' & 0 & 0 \\ 0 & \beta' & 0 \\ 0 & 0 & \gamma' \end{pmatrix}$$

$$\begin{aligned} \alpha' + \beta' + \gamma' &= \alpha_1 + k\alpha_2 + \beta_1 + k\beta_2 + \gamma_1 + k\gamma_2 \\ &= \alpha_1 + \beta_1 + \gamma_1 + k(\alpha_2 + \beta_2 + \gamma_2) \\ &= 0 + k \cdot 0 = 0 \end{aligned}$$

$$\therefore V_1 + kV_2 = \begin{pmatrix} \alpha' & 0 & 0 \\ 0 & \beta' & 0 \\ 0 & 0 & \gamma' \end{pmatrix} \quad \& \quad \alpha' + \beta' + \gamma' = 0$$

 $\therefore W \subseteq M_{3,3}(\mathbb{R})$ 

$$V = \begin{pmatrix} \alpha & 0 & 0 \\ 0 & \beta & 0 \\ 0 & 0 & \gamma \end{pmatrix} = \begin{pmatrix} \alpha & 0 & 0 \\ 0 & \beta & 0 \\ 0 & 0 & -\alpha-\beta \end{pmatrix} = \begin{pmatrix} \alpha & 0 & 0 \\ 0 & \beta & 0 \\ 0 & 0 & -\alpha \end{pmatrix} + \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & \beta \\ 0 & 0 & -\beta \end{pmatrix}$$

$$= \alpha \underbrace{\begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & -1 \end{pmatrix}}_{GW} + \beta \underbrace{\begin{pmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & -1 \end{pmatrix}}_{\text{Trace} = 0, \text{ Both}}$$

GW Trace = 0. Both

$$\therefore W = \left\langle \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & -1 \end{pmatrix}, \begin{pmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & -1 \end{pmatrix} \right\rangle$$

$$k_1 \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & -1 \end{pmatrix} + k_2 \begin{pmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & -1 \end{pmatrix} = \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

$$\begin{pmatrix} k_1 & 0 & 0 \\ 0 & k_2 & 0 \\ 0 & 0 & k_3 \end{pmatrix} = \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} \quad k_1 = k_2 = 0$$

$B = \left\{ \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & -1 \end{pmatrix}, \begin{pmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & -1 \end{pmatrix} \right\}$  is linearly independent spanning set. Hence a basis.

$$\dim(W) = 2.$$

Summary

## Summary

### STATISTICS FORMULA SUMMARY

#### SAMPLE STATISTICS

$$S_x^2 = \frac{1}{n-1} \sum (x_i - \bar{x})^2 = \frac{1}{n-1} \sum x_i^2 - \frac{1}{n} (\sum x_i)^2$$

$$S_y^2 = \frac{1}{n-1} \sum (y_i - \bar{y})^2 = \frac{1}{n-1} \sum y_i^2 - \frac{1}{n} (\sum y_i)^2$$

$$S_{xy} = \frac{1}{n-1} \sum (x_i - \bar{x})(y_i - \bar{y}) = \frac{1}{n-1} \sum x_i y_i - \frac{1}{n} (\sum x_i)(\sum y_i)$$

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

$$\text{Lower Quartile: } \frac{n+1}{4}$$

$$\text{Upper Quartile: } \frac{3(n+1)}{4}$$

$$Q_1 = \text{Lower Quartile Range: } 2Q_1 - Q_3$$

#### Regression

$$\hat{\beta}_1 = \frac{S_{xy}}{S_x^2} \quad \hat{\beta}_0 = \bar{y} - \hat{\beta}_1 \bar{x}$$

$$\hat{\beta}_1^* = \frac{S_{xy}}{S_y^2} \quad \hat{\beta}_0^* = \bar{y} - \hat{\beta}_1^* \bar{x}$$

Predict Y from X.

Predict X from Y

#### Symmetry and Skewness

$$g = \frac{1}{n} \sum \left( \frac{x_i - \bar{x}}{S} \right)^3$$

#### OQ Plot:

1. Rank:  $y_1 < y_2 < \dots < y_m$
2. Find  $x_i$  s.t:  $f(x_i) = \frac{i-\frac{1}{2}}{n}$   $\bar{f}(x_i)$  is cdf.
3. Plot  $y_i$  against  $x_i$ .

	$\bar{x} \pm 2\sqrt{\frac{S}{n}}$
one tail	$t = \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}} \sim t_{n-1}$
two tail	$\bar{x}_1 - \bar{x}_2 \pm 2\sqrt{\frac{S^2}{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}}$
$S^2$	$t = \frac{\bar{x}_1 - \bar{x}_2 - (\mu_1 - \mu_2)}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}} \sim N(0, 1)$
linear combination	$\bar{x}_1 - \bar{x}_2 \sim t_{n_1+n_2-2} \cdot S_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}$
maximum difference	$t = r \sqrt{\frac{n-2}{1-\rho^2}} \sim t_{n-2}$
progression	$F = \frac{S_1^2 / S^2}{S_2^2 / S^2} \sim F_{n_1-1, n_2-1}$
$Y$	$\hat{\beta}_1 \pm t_{n-2} \cdot S_p \sqrt{\frac{6}{n-1} S^2}$
individual $Y$	$\hat{\beta}_0 + \hat{\beta}_1 x_0 \pm t_{n-2} \cdot S_p \sqrt{\frac{1}{n} + \frac{(x_0 - \bar{x})^2}{(n-1) S^2}}$
one proportion	$t = \frac{\hat{p} - p_0}{\sqrt{\frac{p_0(1-p_0)}{n}}} \sim N(0, 1)$
proportion	$\chi^2 = \frac{(p_i - \hat{p}_i)^2}{\hat{p}_i(1-\hat{p}_i)} \sim \chi^2_{n-1}$
	$\hat{x} = \frac{x_1 + x_2}{n_1 + n_2}$

Summary

$$P = \frac{\tau}{n}$$

$$\hat{\sigma}^2 = \frac{(n-1)S^2 + (k-1)S_k^2}{(n-2)} [S_k^2 - (\hat{\beta}_1 S_p)^2]$$

5

18.4

LMs.

- Eigenvector and eigenvalue.
- Repeated value.

$$Av = \lambda v.$$

$$(A - \lambda I)v = 0$$

A

$$(18.3) A = \begin{pmatrix} 1 & -4 & 2 \\ -4 & 1 & -2 \\ 2 & -2 & -2 \end{pmatrix}$$

Characteristic:  $c(\lambda) = \det(A - \lambda I)$ 

$$c(\lambda) = \begin{vmatrix} 1-\lambda & -4 & 2 \\ -4 & 1-\lambda & -2 \\ 2 & -2 & -2-\lambda \end{vmatrix} = -C3 + \lambda^2(1-\lambda - 6)$$

$$\lambda_1 = 6, \quad \lambda_2 = \lambda_3 = -3$$

①  $\lambda_1 = 6$ .  
 $v_1 = \begin{pmatrix} \alpha \\ \beta \\ \gamma \end{pmatrix} \quad \begin{pmatrix} -5 & -4 & 2 \\ -4 & -5 & -2 \\ 2 & -2 & -8 \end{pmatrix} \begin{pmatrix} \alpha \\ \beta \\ \gamma \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$

$$\begin{pmatrix} -5 & -4 & 2 & | & 0 \\ -4 & -5 & -2 & | & 0 \\ 2 & -2 & -8 & | & 0 \end{pmatrix}$$

$$B' - B^2 = -\frac{1}{2}B^3 \quad B \begin{pmatrix} 2 \\ 1 \\ 1 \end{pmatrix} = 0_3$$

$$v_1 = t \begin{pmatrix} 2 \\ -2 \\ 1 \end{pmatrix} \quad E(A; 1) = \langle (2, -2, 1) \rangle.$$

②  $\lambda_2 = \lambda_3 = -3$      $v = \begin{pmatrix} \alpha \\ \beta \\ \gamma \end{pmatrix}$

$$\begin{pmatrix} 4 & -4 & 2 \\ -4 & 4 & -2 \\ 2 & -2 & 1 \end{pmatrix} \begin{pmatrix} \alpha \\ \beta \\ \gamma \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

Augmented Matrix:  $\begin{pmatrix} 4 & -4 & 2 & | & 0 \\ -4 & 4 & -2 & | & 0 \\ 2 & -2 & 1 & | & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 4 & -4 & 2 & | & 0 \\ 0 & 0 & 0 & | & 0 \\ 0 & 0 & 0 & | & 0 \end{pmatrix}$

$$4\alpha - 4\beta + 2\gamma = 0 \quad \gamma = -2\alpha + 2\beta,$$

$$B^1 \ B^2 \ B^3.$$

解得: 可表示为

$$(3 \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} + Y \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix})$$

最后的空间都是一样的

Summary

$$v = \begin{pmatrix} \alpha \\ \beta \\ \gamma \end{pmatrix} = \begin{pmatrix} \alpha \\ \beta \\ -2\alpha + 2\beta \end{pmatrix} \quad \alpha \neq 0, \beta \neq 0$$

$$= \begin{pmatrix} \alpha \\ 0 \\ -2\alpha \end{pmatrix} + \begin{pmatrix} 0 \\ \beta \\ 2\beta \end{pmatrix} = \alpha \begin{pmatrix} 1 \\ 0 \\ -2 \end{pmatrix} + \beta \begin{pmatrix} 0 \\ 1 \\ 2 \end{pmatrix}$$

Eigenspace:  $\begin{pmatrix} x \\ y \\ z \end{pmatrix} = \alpha \begin{pmatrix} 1 \\ 0 \\ -2 \end{pmatrix} + \beta \begin{pmatrix} 0 \\ 1 \\ 2 \end{pmatrix} + \gamma \cdot 0 \in \mathbb{R}$

$$x = \alpha, \quad y = \beta, \quad z = -2\alpha + 2\beta$$

$$2x - 2y + z = 0.$$

求 Range 也可这样

5

Diagonalisation of A.

$$\lambda_1 = 6, \quad v_1 = \begin{pmatrix} 2 \\ -2 \\ 1 \end{pmatrix}$$

$$\lambda_2 = \lambda_3 = -3, \quad v_2 = \begin{pmatrix} 1 \\ 0 \\ -2 \end{pmatrix}, \quad v_3 = \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix}$$

Linearly independent.

$$\text{take } P = (v_1, v_2, v_3) = \begin{pmatrix} 2 & 1 & 0 \\ -2 & 0 & 1 \\ 1 & -2 & 2 \end{pmatrix}$$

$$D = \begin{pmatrix} \lambda_1 & 0 & 0 \\ 0 & \lambda_2 & 0 \\ 0 & 0 & \lambda_3 \end{pmatrix} = \begin{pmatrix} 6 & 0 & 0 \\ 0 & -3 & 0 \\ 0 & 0 & -3 \end{pmatrix}$$

$$\text{Then } P^{-1}AP = D.$$

Calculate  $P^{-1}$ 

$$\begin{pmatrix} 2 & 1 & 0 \\ -2 & 0 & 1 \\ 1 & -2 & 2 \end{pmatrix} \rightarrow \left( \begin{array}{ccc|ccc} 2 & 1 & 0 & 1 & 0 & 0 \\ -2 & 0 & 1 & 0 & 1 & 0 \\ 1 & -2 & 2 & 0 & 0 & 1 \end{array} \right)$$

$$P^{-1} = \frac{1}{9} \begin{pmatrix} 2 & -2 & 1 \\ 5 & 4 & -2 \\ 4 & 5 & 2 \end{pmatrix}$$

$$P^{-1}AP = \begin{pmatrix} 6 & 0 & 0 \\ 0 & -3 & 0 \\ 0 & 0 & -3 \end{pmatrix} = D$$

Orthogonal

 $\{v_1, v_2, v_3\}$  Three orthogonal eigenvectors

$$\text{Normalised: } \hat{v}_i = \frac{v_i}{\|v_i\|}$$

 $\{\hat{v}_1, \hat{v}_2, \hat{v}_3\}$  three orthonormal eigenvectors

$$\hat{v}_i^T \hat{v}_j = 0 \quad i \neq j$$

$$\hat{v}_i^T \hat{v}_i = 1 \quad i = j$$

Summary

Diagonalise matrix  $P$

$$P = [\underline{\hat{V}_1}; \underline{\hat{V}_2}; \underline{\hat{V}_3}]$$

$$\text{since } P^{-1} A P = D = \begin{bmatrix} \lambda_1 & 0 & 0 \\ 0 & \lambda_2 & 0 \\ 0 & 0 & \lambda_3 \end{bmatrix}$$

$$\text{But } P^T P = I$$

$$\Rightarrow P^{-1} = P^T$$

$$\Rightarrow P^T A P = D$$

### LINEAR TRANSFORMATION

MAPPINGS BETWEEN VECTOR SPACES  $V$  AND  $W$

•  $W = V$  is possible

• Usually described by Matrix.

Definition:

$T: V \rightarrow W$  is a linear transformation if

for all  $v_1, v_2 \in V, k \in \mathbb{R}$

$$[T(v_1 + v_2) = T(v_1) + T(v_2)] \text{ & } [T(kv_1) = kT(v_1)]$$

Equivalently:

$$T(v_1 + kv_2) = T(v_1) + kT(v_2)$$

$T: V \rightarrow W$

$$v \in V \Rightarrow T(v) \in W$$

$$T(v_1 + \underset{\substack{\uparrow \\ \text{in } V}}{kv_2}) = T(v_1) + \underset{\substack{\uparrow \\ \text{in } W}}{kT(v_2)} \quad \forall v_1, v_2 \in V$$

Notice:  $T(\underset{\substack{\uparrow \\ \text{in } V}}{0}) = \underset{\substack{\uparrow \\ \text{in } W}}{0}$

Summary

5

Test for linear Transformation

$$\text{eq: } T: \mathbb{R}^2 \rightarrow \mathbb{R}^2$$

$$T \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} ax+y+1 \\ x-y \end{bmatrix}$$

$$T(0) = T \begin{bmatrix} 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 0+0+0+1 \\ 0-0 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \neq \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$\therefore T$  cannot be a linear transformation.

$$\text{eq: } V = \mathbb{R} \quad W = M_{2 \times 2}(\mathbb{R})$$

$$T_1(x) = \begin{bmatrix} x & 0 \\ 0 & 1 \end{bmatrix} \quad T_2(x) = \begin{bmatrix} x & 0 \\ 0 & 0 \end{bmatrix}$$

$$\text{In } V, '0' = 0. \quad \text{In } W, '0' = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$$

$$T_1(0) = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \neq \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} = 0$$

$\therefore T_1$  is not a L.T. (let  $v_1 = x_1, v_2 = x_2$ )

$$T_2(v_1 + kv_2) = T_2(x_1 + kx_2)$$

$$T_2(v_1 + kv_2) = T_2(x_1 + kx_2) = \begin{bmatrix} x_1 + kx_2 & 0 \\ 0 & 0 \end{bmatrix}$$

$$= \begin{bmatrix} x_1 & 0 \\ 0 & 0 \end{bmatrix} + k \begin{bmatrix} x_2 & 0 \\ 0 & 0 \end{bmatrix}$$

$$= T_2(v_1) + kT_2(v_2)$$

$$T_2(v_1 + kv_2) = T_2(v_1) + kT_2(v_2)$$

$T_2$  is a L.T.

$$\text{eq: } V = \mathbb{R}^3 \quad W = \mathbb{R}$$

$$T_3 \begin{bmatrix} x \\ y \\ z \end{bmatrix} = xy \quad T_4 \begin{bmatrix} x \\ y \\ z \end{bmatrix} = x+y+z$$

Both cases:  $T(0) = 0$

$$T_3 \text{ take } v = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \quad v+v = \begin{bmatrix} 2 \\ 2 \\ 2 \end{bmatrix} \quad T_3(v+v) = 2 \times 2 = 8$$

$$\text{But } T_3(v) + T_3(v) = 2T_3(v) = 2 \times 1 = 2 \neq 8$$

$$\therefore T_3(v+v) \neq T_3(v) + T_3(v)$$

Summary

$T_3$  is not a L.T.

$$\begin{aligned} T_4(kv_1 + kv_2) &= (x_1 + kx_2) + (y_1 + ky_2) + (z_1 + kz_2) \\ &= (x_1 + y_1 + z_1) + k(x_2 + y_2 + z_2) \end{aligned}$$

• •

$\therefore T_4$  is a L.T.