

## Testing for Bivariate Normal

Suppose two variables  $x$  and  $y$  form B.N.D

$$(\mu_1, \mu_2; \sigma_1^2, \sigma_2^2, \rho)$$

$$f_{X,Y}(x,y) = \frac{1}{2\pi\sigma_1\sigma_2} \frac{e^{-\frac{1}{2(1-\rho^2)} \left\{ \frac{(x-\mu_1)^2}{\sigma_1^2} - 2\rho \frac{(x-\mu_1)(y-\mu_2)}{\sigma_1\sigma_2} + \frac{(y-\mu_2)^2}{\sigma_2^2} \right\}}}{}$$

$$\mu_1 - \mu_2 \in \mathbb{R}, 0 < \sigma_1, \sigma_2 < \infty, -1 < \rho < 1$$

• Here consider R.S from B.N.  $(\mu_1, \mu_2, \sigma_1^2, \sigma_2^2, \rho)$

• of size  $n$   $(x_1, y_1), (x_2, y_2), \dots, (x_n, y_n)$

Here we want to test <sup>difference test</sup>  $\mu_x$  and  $\mu_y$  if  $x$  and  $y$  variables are measured in same units

consider a test  $H_0: \mu_x - \mu_y = \epsilon_0$  or  $\epsilon_0 \neq 0$

$$H_1: \mu_x - \mu_y \neq \epsilon_0$$

Here let  $Z = X - Y, \sim N(\mu_x - \mu_y, \sigma_x^2 + \sigma_y^2 - 2\rho\sigma_x\sigma_y)$

$$\text{Here } z_i = x_i - y_i, \bar{z} = \bar{x} - \bar{y}$$

$$S_z^2 = \frac{1}{n-1} \sum (z_i - \bar{z})^2$$

Here our problem is reduced

to test  $H_0: \mu_z = \epsilon_0$  or  $\mu_z \neq \epsilon_0$

where  $\sigma^2$  unknown

So test statistic is

$$T = \frac{\sqrt{n}(\bar{Z} - \mu_0)}{s} \sim t_{n-1}$$

the test procedure is for  $H_0: \mu = \mu_0$

vs  $H_1: \mu \neq \mu_0$  is same as

either  $(H_0: \mu = \mu_0$  vs  $H_1: \mu > \mu_0$ )

ie if  $|T| > t_{\frac{\alpha}{2}, n-1}$  then reject

$H_0$  at level  $\alpha$  vs  $H_1: \mu \neq \mu_0$

$$W = \{ |T| < t_{\frac{\alpha}{2}, n-1} \text{ or } |T| > t_{\frac{\alpha}{2}, n-1} \}$$

for  $H_0: \mu = \mu_0$  vs  $H_1: \mu > \mu_0$

$$W = \{ |T| > t_{\alpha, n-1} \}$$

$$= \left\{ z: \frac{\sqrt{n}(\bar{Z} - \mu_0)}{s} > t_{\alpha, n-1} \right\}$$

for  $H_0: \mu = \mu_0$  vs  $H_1: \mu < \mu_0$

$$W = \left\{ z: \frac{\sqrt{n}(\bar{Z} - \mu_0)}{s} < -t_{\alpha, n-1} \right\}$$

of the known as paired t test.

### Real life example

Suppose weights of  $n$  patients before they are subjected to change a diet and after a lapse of ~~the~~ six months are given. Then to test any significant change in weight gain we can use paired t-test.

• Testing for  $\rho = 0$

Here  $r = \frac{\frac{1}{n} \sum x_i y_i - \bar{x} \bar{y}}{\sqrt{v(x)} \sqrt{v(y)}}$  Sample correlation coefficient

Here  $H_0: \rho = 0$  agt.  $H_1: \rho \neq 0$

Under  $H_0: \rho = 0$  ;  $t = \frac{r\sqrt{n-2}}{\sqrt{1-r^2}}$   $\stackrel{H_0}{\sim} t_{n-2}$

So  $t = \frac{r\sqrt{n-2}}{\sqrt{1-r^2}}$   $\stackrel{H_0}{\sim} t_{n-2}$  is test

Statistic  $t$  of under  $H_0$   $E(t) = 0$

ie under  $H_0$  we expect  $t \approx 0$

So, here if  $|t| > t_{\frac{\alpha}{2}; n-1}$

(at confidence level)  $\rightarrow$  reject  $H_0$

$H_0: \rho = 0$  vs  $H_1: \rho \neq 0$  (two-tailed)

So critical region is

$$W = \left\{ \text{sig} \left| \frac{r \sqrt{n-2}}{\sqrt{1-r^2}} \right| > t_{\frac{\alpha}{2}; n-1} \right\}$$

• Suppose  $(X_1, X_2)$  has a Bivariate Normal distribution with

unknown parameters  $(\mu_1, \mu_2, \sigma_1, \sigma_2, \rho)$

test  $H_0: \beta = 0$  vs  $H_1: \beta \neq 0$

where  $\beta$  is the regression coefficient of  $X_2$  on  $X_1$

Soln

$$E(X_2 | X_1 = x) = \mu_2 + \rho \frac{\sigma_2}{\sigma_1} (x - \mu_1)$$

is the regression equation of  $X_2$  on  $X_1$

$$\beta = \rho \frac{\sigma_2}{\sigma_1}$$

$$H_0: \beta = 0 \Rightarrow \rho \frac{\sigma_2}{\sigma_1} = 0 \text{ i.e. } \rho = 0$$

(as  $\sigma_2 > 0$ )

So testing is reduced to  $H_0: \rho = 0$  vs.  $\rho \neq 0$

• ~~PO~~

Testing for  $\frac{\partial x}{\partial y}$

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$\{(x_1, y_1), (x_2, y_2), \dots, (x_n, y_n)\}$  are the  
R.S of size  $n$  from  $BN(\mu_1, \sigma_1, \mu_2, \sigma_2, \rho)$

Here  $H_0: \frac{\partial x}{\partial y} = \epsilon_0$  vs.  $H_1: \frac{\partial x}{\partial y} \neq 0$

$$\text{Let } U = X + \epsilon_0 Y$$
$$V = X - \epsilon_0 Y$$

wee  $\epsilon_0 = \frac{\partial x}{\partial y}$

Here  $U$  and  $V$  are jointly distributed  
as normal variables

$$\text{with } \text{cov}(U, V) = \sigma_x^2 - \epsilon_0^2 \sigma_y^2 = 0$$

So under  $H_0: \frac{\partial x}{\partial y} = \epsilon_0$

$$U = X + \epsilon_0 Y$$

$$V = X - \epsilon_0 Y$$

Here  $\text{cov}(U, V) = 0$  i.e.  $\frac{\partial x^2}{\partial y^2} = \epsilon_0^2$

Sol:  $H_0: \frac{\sigma_x}{\sigma_y} = 1$  ag.  $H_1: \frac{\sigma_x}{\sigma_y} \neq 1$

reduces to  $H_0: \rho_{uv} = 0$

Here test statistic is

$$t = \frac{r_{uv} \sqrt{n-2}}{\sqrt{1-r_{uv}^2}}$$

So the test is as earlier [See earlier test p. 20]

Ex) Suppose the weights of  $n$  pigs before and after application of a special food supplement in a farm are recorded, under the suitable assumption provide a test procedure for testing whether the variability in the pig weight before and after applying food supplement is same or not.

Hint:- Use test procedure for  $H_0: \sigma_x = \sigma_y$  i.e.  $H_0: \frac{\sigma_x}{\sigma_y} = 1$  ag  $H_1: \frac{\sigma_x}{\sigma_y} \neq 1$