

## Normal correlation analysis

Part 5 of regression

$$f(x,y) = \frac{1}{2\pi\sigma_x\sigma_y\sqrt{1-\rho^2}} e^{-\frac{1}{2(1-\rho^2)} \left[ \left(\frac{x-\mu_x}{\sigma_x}\right)^2 + \left(\frac{y-\mu_y}{\sigma_y}\right)^2 - 2\rho\left(\frac{x-\mu_x}{\sigma_x}\right)\left(\frac{y-\mu_y}{\sigma_y}\right) \right]}$$

$$f(x,y; \sigma_x^2, \sigma_y^2, \rho, \mu_x, \mu_y)$$

$(x_1, y_1) \quad (x_2, y_2) \quad \dots \quad (x_n, y_n)$

use MLE to estimate values of parameters  $\mu_x, \mu_y, \sigma_x^2, \dots$

$$L(\sigma_x^2, \sigma_y^2, \rho, \mu_x, \mu_y; \vec{x}, \vec{y}) = \prod_{k=1}^n f(x_k, y_k; \sigma_x^2, \sigma_y^2, \rho, \mu_x, \mu_y)$$

$$\text{maximize via } \frac{\partial \ln L}{\partial \rho} = 0$$

$$\frac{\partial \ln L}{\partial \mu_x} = 0 \dots$$

$$\mu_x = \bar{x} \quad \sigma_x^2 = \frac{1}{n} \sum (x_i - \bar{x})^2 = s_{xx}$$

$$\mu_y = \bar{y} \quad \sigma_y^2 = \frac{1}{n} \sum (y_i - \bar{y})^2 = s_{yy}$$

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

$$\rho = \frac{S_{xy}}{\sqrt{S_{xx}S_{yy}}}$$

get estimates

$$\hat{\mu}_x = \bar{x} \quad \hat{\mu}_y = \bar{y} \quad \hat{\sigma}_x^2 = S_{xx} \quad \hat{\sigma}_y^2 = S_{yy}$$

$$\hat{\rho} = \frac{S_{xy}}{\sqrt{S_{xx}S_{yy}}}$$

ex: want interval estimate for  $\sigma_x^2$   $C \frac{\hat{\sigma}_x^2}{\sigma_x^2} \chi_{n-1}^2$

using  $\chi^2$ , t variables, can get interval estimates  
can do hyp tests for values of

$$\sigma_x^2, \sigma_y^2, \mu_x, \mu_y$$

$$\hat{\rho} = \frac{S_{xy}}{\sqrt{S_{xx}S_{yy}}}$$

sample value for this  
is referred to by  
"r"

$$\sigma_{y|x}^2 = \sigma_y^2 (1 - \rho^2)$$

$$-1 \leq \rho \leq 1$$

$$\rho = 0$$

(x & y indep)

$$\rho = \pm 1$$

$$\rho^2 = \frac{\sigma_y^2 - \sigma_{y|x}^2}{\sigma_y^2} = 1 - \underbrace{\left( \frac{\sigma_{y|x}^2}{\sigma_y^2} \right)}$$

fraction of variance "due to x"

the fraction of variance remains after x is known

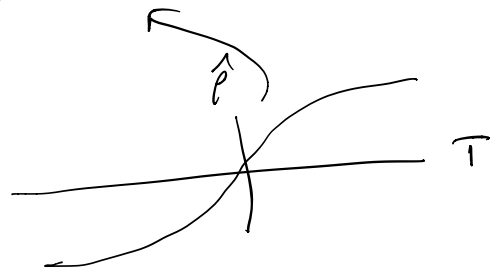
Special case: if  $\rho = 0$

$$\hat{\rho} = T = \hat{\rho} \sqrt{\frac{n-2}{1-\hat{\rho}^2}}$$

tuns out this is a t-distributed variable w/  $n-2$  degrees of freedom

$$\hat{\rho} = \frac{T}{\sqrt{n-2+T^2}}$$

non-increasing odd 1-1 function



$$P(\hat{\rho} > \lambda \mid \rho = 0)$$

$$\Leftrightarrow P\left(\hat{\rho} \sqrt{\frac{n-2}{1-\hat{\rho}^2}} > \lambda \sqrt{\frac{n-2}{1-\lambda^2}}\right)$$

$$\frac{1}{T} P(T > \lambda \sqrt{\frac{n-2}{1-\lambda^2}})$$


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What if  $\rho \neq 0$ ?  $\hat{\rho}$

$$P(\hat{\rho} > 0.7 \mid \rho = 0.5)?$$

Can write an exact expression for density func. for  $\hat{\rho}$

$$f(r) = \frac{(n-2) \Gamma(n-1) (1-\rho^2)^{\frac{n-1}{2}} (1-r^2)^{\frac{n-4}{2}}}{\sqrt{2\pi} \Gamma(n-\frac{1}{2}) (1-\rho r)^{n-3/2}} \cdot {}_2F_1\left(\frac{1}{2}, \frac{1}{2}, \frac{2n-1}{2}, \frac{\rho r+1}{2}\right)$$

$${}_2F_1(a, b; c; x) = \sum_{k=0}^{\infty} \frac{a \overline{a}_k}{c \overline{c}_k} \frac{x^k}{k!}$$

$$\overline{m}_k = m(m+1)(m+2) \dots (m+k-1) \quad \text{if } k \geq 1$$

$$\overline{m}_0 = 1$$