

Measures of Bivariate Association

Things to learn: covariance
correlation
location invariance
scale invariance

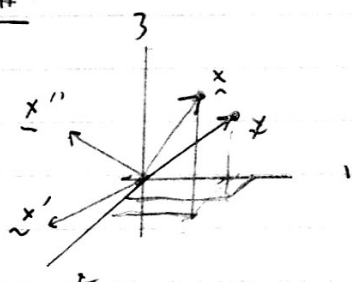
Data: $\{(x_i, y_i) \mid i=1 \dots n\}$ unit level notation
 $\{\underline{x}, \underline{y}\}$ vector notation

$$\underline{x} = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \quad \underline{y} = \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix}$$

Questions: How might we measure the "association" between $\underline{x}, \underline{y}$?

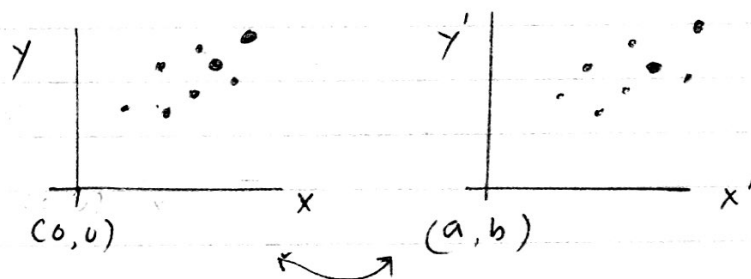
big/pos x_i 's occur w/ $\begin{cases} \text{big/pos } y_i\text{'s} \rightarrow \text{"positive association"} \\ \text{small/neg } y_i\text{'s} \rightarrow \text{"negative association"} \end{cases}$

Idea #1
 $n=3$



$$\begin{aligned} \text{assoc}(\underline{x}, \underline{y}) &= \underline{x} \cdot \underline{y} / n \quad (\text{vector}) \\ &= \sum x_i y_i / n \quad (\text{summation}) \\ &= \text{mean}(x * y) \quad (R) \end{aligned}$$

Problem #



$$\text{assoc}(\underline{x}, \underline{y}) \neq \text{assoc}(\underline{x}', \underline{y}')$$

Remedy #1

$$\begin{aligned} \text{assoc}(\underline{x}, \underline{y}) &= (\underline{x} - \bar{x} \underline{1}) \cdot (\underline{y} - \bar{y} \underline{1}) / n \\ &= \sum (x_i - \bar{x})(y_i - \bar{y}) / n \end{aligned}$$

$$\text{Recall, } \text{Cov}(\underline{x}, \underline{y}) = \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{n-1}$$

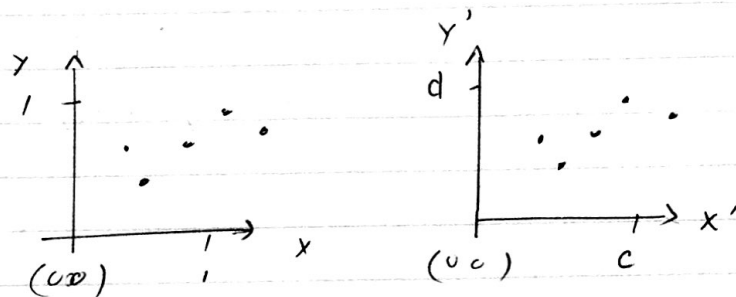
= "sample covariance"

$$= \text{mean}((x - \text{mean}(x)) * (y - \text{mean}(y)))$$

$$= \frac{n}{n-1} \text{Cov}(\underline{x}, \underline{y})$$

Exercise: Show $\text{cov}(\underline{x}, \underline{y}) = \text{cov}(\underline{x} + a\underline{1}, \underline{y} + b\underline{1}) \quad \forall a, b$
 $\Rightarrow$ covariance is a location invariant statistic

Problem



$$\text{cov}(\underline{x}, \underline{y}) \neq \text{cov}(\underline{x}', \underline{y}')$$

Example: let $\underline{x}' = \underline{x}$
 $\underline{y}' = 3\underline{y}$ $\text{cov}(\underline{x}, \underline{y}) \stackrel{?}{\neq} \text{cov}(\underline{x}', \underline{y}')$

are $(\underline{x}, \underline{y})$ "more associated" than $(\underline{x}', \underline{y}')$?

Observation: Covariance depends on the scale of the measurements

Remedy #2: let $\underline{r}_y = (\underline{y} - \bar{y}\underline{1})$ ($r_y[i] = y[i] - \bar{y}$)
 $\underline{r}_x = (\underline{x} - \bar{x}\underline{1})$

$$\text{Result} \quad \text{cov}(\underline{x}, \underline{y}) = \underline{r}_y \cdot \underline{r}_x / (n-1) = \text{cov}(\underline{r}_x, \underline{r}_y) \quad (\text{why?})$$

$$\text{now let} \quad \underline{z}_y = (\underline{y} - \bar{y}\underline{1}) / \|\underline{y} - \bar{y}\underline{1}\| \quad \left\{ \begin{array}{l} \|\underline{y} - \bar{y}\underline{1}\| = (\sum (y_i - \bar{y})^2)^{1/2} \\ z_y[i] = \frac{y[i] - \bar{y}}{(\sum (y_i - \bar{y})^2)^{1/2}} \end{array} \right.$$

$$= (\underline{y} - \bar{y}\underline{1}) / (s_{yy})^{1/2}$$

$$\text{let} \quad \underline{z}_x = (\underline{x} - \bar{x}\underline{1}) / \|\underline{x} - \bar{x}\underline{1}\|$$

$$= (\underline{x} - \bar{x}\underline{1}) / s_{xx}^{1/2}$$

$\tilde{z}_x$ and $\tilde{z}_y$ are centered + scaled versions of $\underline{x}, \underline{y}$.

Check: $\|\tilde{z}_x\|^2 = \sum \tilde{z}_x[i]^2 = 1$

Consider $\text{assoc}(\underline{x}, \underline{y}) = \tilde{z}_x \cdot \tilde{z}_y = \sum \tilde{z}_x[i] \tilde{z}_y[i]$

$$= \sum \left(\frac{x_i - \bar{x}}{s_{xx}^{1/2}} \right) \left(\frac{y_i - \bar{y}}{s_{yy}^{1/2}} \right)$$

$$= \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{\left(\sum (x_i - \bar{x})^2 \sum (y_i - \bar{y})^2 \right)^{1/2}}$$

$$= \frac{s_{xy}}{(s_{xx} \cdot s_{yy})^{1/2}} = \text{Cor}(\underline{x}, \underline{y})$$

= "sample correlation"

Other forms:

$$\text{Var}(\underline{x}) = \sum (x_i - \bar{x})^2 / n - 1 = s_{xx} / n - 1 = s_x^2$$

$$\text{Var}(\underline{y}) = \sum (y_i - \bar{y})^2 / n - 1 = s_{yy} / n - 1 = s_y^2$$

$$\text{Cov}(\underline{x}, \underline{y}) = \sum (x_i - \bar{x})(y_i - \bar{y}) / n - 1 = s_{xy} / n - 1$$

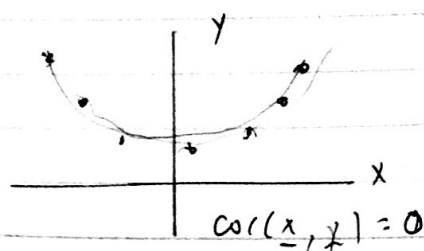
$$\rightarrow \text{Cor}(\underline{x}, \underline{y}) = \frac{\text{Cov}(\underline{x}, \underline{y})}{(\text{Var}(\underline{x}) \text{Var}(\underline{y}))^{1/2}}$$

Exercise: Show $\text{Cor}(\underline{x}, \underline{y}) = \text{Cor}(a\underline{x} + b\underline{1}, c\underline{y} + d\underline{1})$ for $a, c > 0$

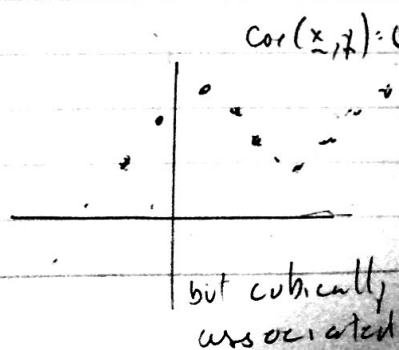
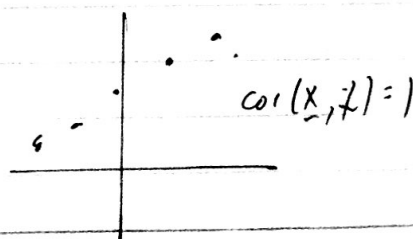
(what happens if $a > 0, c = 0$?
 $c < 0$?)

$\Rightarrow \text{Cor}(\underline{x}, \underline{y})$ is a location and scale invariant measure of association.

Comments: (1) Cor is a measure of linear association



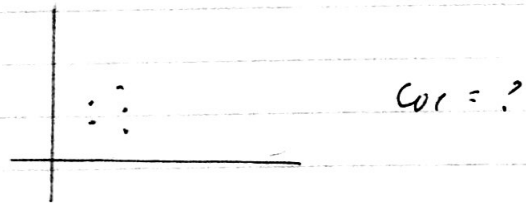
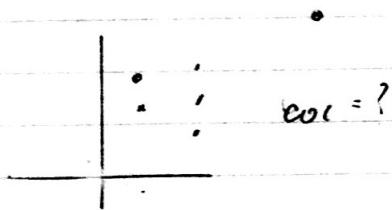
but quadratically associated



(2) $\text{Cor}(x, x)$ is not invariant to monotone transformations of the data:

$$\cos(\underline{x}, \underline{z}) \neq \cos(\cos \underline{x}, \cos \underline{z}) \text{ in general}$$

③ $\text{cor}(\underline{x}, \underline{y})$ is not robust to outliers



(4) Extremes of correlation: $-1 \leq \text{Cor}(\tilde{x}, \tilde{z}) \leq 1$ (see Cauchy-Schwarz)

$\text{cov}(\underline{x}, \underline{z}) = 0 \Rightarrow \underline{x}, \underline{z} \text{ are uncorrelated}$ (rarely happens in a sample)

$$\text{Cor}(\hat{x}, \hat{y}) = 1 \quad (\Leftrightarrow) \quad y_i = \beta_0 + \beta_1 x_i, i=1 \dots n, \text{ some } \beta_0, \beta_1 \neq 0$$
$$= -1 \quad (=) \quad " \quad \text{Surse } \beta_0, \beta_1 < 0$$

This leads to the idea of a linear regression model

$$\text{Cor}(\underline{x}, \underline{x}) = r \quad \Leftrightarrow \quad y_i = \beta_0 + \beta_1 x_i + \epsilon_i$$

magnitude of deviations
depend on r

