

$$\text{Var}(x_1 + x_2) = \text{Var}(x_1) + \text{Var}(x_2) + 2 \text{cov}(x_1, x_2)$$

$$z \sim N(0, 1)$$



$$x = z$$

$$y = z^2$$



$$\text{Cov}(x, y)$$

$$E((x - E(x))(y - E(y)))$$

$$= E((x - E(x))(y - E(y)))$$

$$= E[xy] - E(x)E(y) - xE(y) + E(x)E(y)$$

$$= E[z^2] - E[z]E[z]$$

$$= E[z^2] - E[z]^2$$

$$\int_{-\infty}^{\infty} z^2 \exp\left(-\frac{z^2}{2}\right) dz = 0$$

$$= 0$$

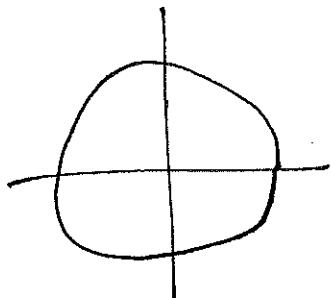
x, y are uncorrelated

however, they are not independent

- if you know the value of x ,
you also know the value of y

covariance / correlation measures
linear association.

If the relationship is nonlinear, we can still have $\text{cov}(X, Y) = 0$



$$f_{x,y}(x, y) = \begin{cases} \frac{1}{2\pi} & x^2 + y^2 \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

X, Y are dependent

$$\text{cov}(X, Y) = E[XY] - E[X]E[Y]$$

$$E[XY] = \iint xy f_{x,y}(x, y) dx dy$$

$$= \frac{1}{2\pi} \iint xy dx dy.$$

interior
of O

$$= \frac{1}{2\pi} \iint_O xy dx dy + \frac{1}{2\pi} \iint_{O^c} xy dx dy$$

$= 0$ the two integrals have the same value but opposite signs.

Again - dependent RVs with covariance = 0

Correlation.

Covariance - the size of the covariance depends on the scale of the variation in X, Y .

Depends on the units of measurement.

Correlation - standardized version of covariance.

$$\begin{aligned}\text{Corr}(X, Y) &= \frac{\text{Cov}(X, Y)}{\text{SD}(X) \text{SD}(Y)} \\ &= \text{Cov} \left(\frac{X - E(X)}{\text{SD}(X)}, \frac{Y - E(Y)}{\text{SD}(Y)} \right)\end{aligned}$$

Correlation is the covariance between standardized versions of the RVs.

Theorem: $-1 \leq \text{Corr}(X, Y) \leq 1$. - we know what "large" values are.

Proof: Assume X, Y are standardized.

$$\text{Var}(X+Y) = \text{Var}(X) + \text{Var}(Y) + 2\text{Cov}(X, Y)$$

$$\text{Var}(X-Y) = \text{Var}(X) + \text{Var}(Y) - 2\text{Cov}(X, Y)$$

let $\rho = \text{Cov}(X, Y)$
 Greek letter rho ρ

$$\text{Var}(\cdot) \geq 0$$

$$\text{Var}(X) + \text{Var}(Y) + 2 \text{Cov}(X, Y) \geq 0$$

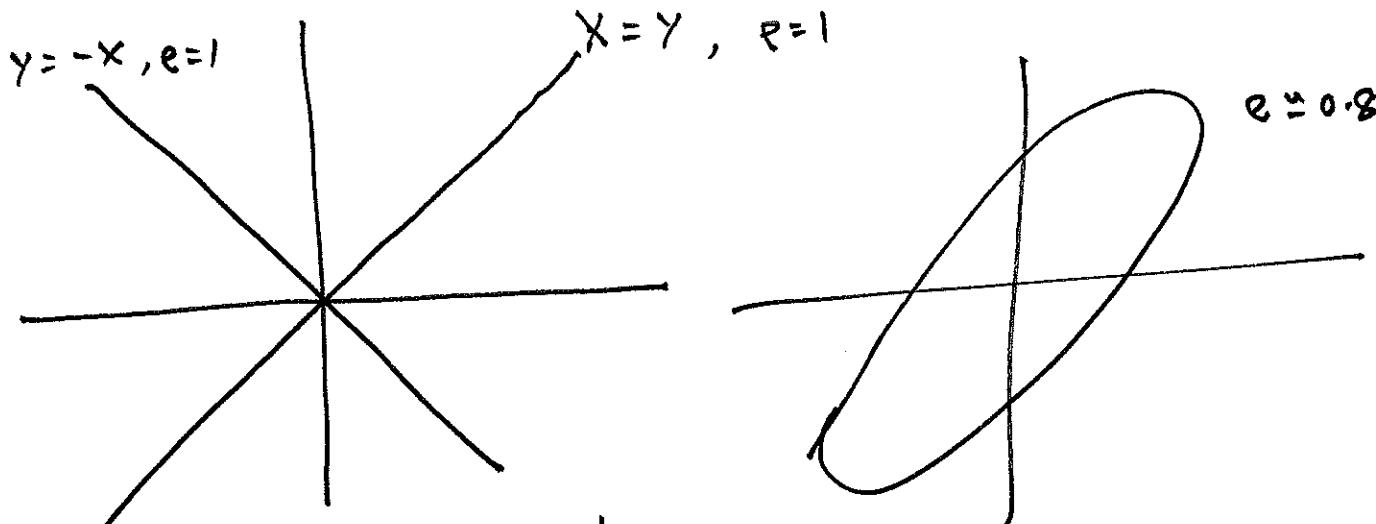
$$\text{Var}(X) + \text{Var}(Y) - 2 \text{Cov}(X, Y) \geq 0$$

$$2 + 2\rho \geq 0$$

$$2 - 2\rho \geq 0$$

hence $-1 \leq \rho \leq 1.$

Correlation is
between -1 and +1.



- nonlinear dependence
may have $\rho = 0$
 even though
 X, Y are dependent.

Variance of Binomial.

$$X \sim \text{Bin}(n, p)$$

$$X = X_1 + X_2 + \dots + X_n \quad X_i \sim \text{iid Bernoulli}(p)$$

$$\begin{aligned}\text{Var}(X_i) &= E(X_i^2) - (E(X_i))^2 \\ &= p - p^2 \\ &= p(1-p) \\ &= pq\end{aligned}$$

$$\text{Var}(X) = npq.$$

X_i can be considered an indicator variable for success on the i^{th} trial.

X_i^2 is an indicator variable for the same event.