

Supplemental Lecture

Ex.

Toss a coin n times, where $P(H) = p$.

Let

$$X = \# \text{ heads}$$

and

$$Y = \# \text{ tails}$$

compute $p(X, Y)$.

Solution

observe $X + Y = n$, so

$$E[X] + E[Y] = E[X + Y] = E[n] = n$$

also

$$\begin{aligned}\text{Var}(Y) &= \text{Var}(-X + n) \\ &= (-1)^2 \text{Var}(X) = \text{Var}(X)\end{aligned}$$

also

$$\begin{aligned}(X - E[X]) + (Y - E[Y]) \\ = (X + Y) - (E[X + Y]) = n - n = 0\end{aligned}$$

so

$$(Y - E[Y]) = -(X - E[X])$$

Thus

$$\begin{aligned}\text{Cov}(X, Y) &= E[(X - E[X])(Y - E[Y])] \\ &= -E[(X - E[X])^2] \\ &= -\text{Var}(X)\end{aligned}$$

and

$$\rho(X, Y) = \frac{\text{cov}(X, Y)}{\sqrt{\text{Var}(X) \cdot \text{Var}(Y)}}$$

$$= \frac{-\text{Var}(X)}{\sqrt{\text{Var}(X)^2}}$$

$$= \frac{-\text{Var}(X)}{\text{Var}(X)} = \boxed{-1}$$

↑
Perfectly negatively
correlated.

Variance of a sum

Let X, Y be r.v.s (not necessarily independent.) Then

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

more generally :

$$\begin{aligned}\text{Var}\left(\sum_{i=1}^n X_i\right) \\ = \sum_{i=1}^n \text{Var}(X_i) + \sum_{i \neq j} \text{Cov}(X_i, X_j)\end{aligned}$$

Proof of case $n=2$

Let $U = X - E[X] \therefore E[U] = 0$

and $V = Y - E[Y] \therefore E[V] = 0$.

Then

$$\text{Var}(X+Y) = \text{Var}(U+V)$$

$$= E[(U+V) - 0]^2$$

$$= E[U^2 + V^2 + 2UV]$$

$$= E[U^2] + E[V^2] + 2E[UV]$$

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



4.3 Conditional Expectation & Variance

16

Recall

$$g(y) = E[X | Y=y]$$

is a function of y . If we substitute Y for y we get a new r.v.

$$g(Y) = E[X | Y] \quad \left\{ \begin{array}{l} \text{Conditional} \\ \text{expectation} \end{array} \right.$$

We showed

$$E[E[X | Y]] = E[g(Y)]$$

$$= \int_{-\infty}^{\infty} g(y) f_Y(y) dy$$

$$= \int_{-\infty}^{\infty} E[X | Y=y] \cdot f_Y(y) dy = E[X]$$

total expectation thm

□

law of iterated expectations:

$$E[X] = E[E[X|Y]]$$

We can do something similar with variance. define

$$h(y) = \text{Var}(X | Y=y)$$

$$= E[(X - E[X|Y=y])^2 | Y=y]$$

By substituting Y for y we get a new r.v.

$$\text{Var}(X|Y) = h(Y)$$

$$= E[(X - E[X|Y])^2 | Y]$$

called the conditional variance.

law of total variance

$$\boxed{\text{Var}(X) = E[\text{Var}(X|Y)] + \text{Var}(E[X|Y])}$$

see Proof on P. 225-227 of text.

Proof

we have

$$\text{Var}(X|Y) = E[X^2|Y] - E[X|Y]^2$$

$$\therefore E[X^2|Y] = \text{Var}(X|Y) + E[X|Y]^2$$

By iterated expectation, we have

$$\text{Var}(X) = E[X^2] - E[X]^2$$

$$= E[E[X^2|Y]] - E[E[X|Y]]^2$$

$$= E[\text{Var}(X|Y) + E[X|Y]^2] - E[E[X|Y]]^2$$

$$= E[\text{Var}(X|Y)] + E[E[X|Y]^2] - E[E[X|Y]]^2$$

$$= E[\text{Var}(X|Y)] + \text{Var}(E[X|Y])$$



Σ KCP 4.4 : Transform

4.5 Sum of a random # of r.v.s

Let $X_1, X_2, \dots$ be an (infinite) sequence of independent and identically distributed (i.i.d) r.v.s, i.e. all have same PDF or PMF denoted

$$f_X(x) \quad \text{or} \quad p_X(x)$$

Let N be a r.v. with $\text{range}(N) = \{0, 1, 2, \dots\}$.

Assume $\{N, X_1, X_2, \dots\}$ is independent.

Define

$$Y = X_1 + X_2 + \dots + X_N$$

denote by $E[X]$, and $\text{Var}(X)$

□

the common mean & variance of
all X_i ($i=1, 2, 3, \dots$)

Let $n \geq 0$ be fixed. Then

$$E[Y | N=n]$$

$$= E[X_1 + \dots + X_N | N=n]$$

$$= E[X_1 + \dots + X_n | N=n]$$

$$= E[X_1 + \dots + X_n] \quad \left\{ \text{by indep.} \right.$$

$$= n E[X]$$

Therefore

$$E[Y | N] = N E[X]$$

by law of iterated expectations

$$E[Y] = E[E[Y|N]]$$

$$= E[N \cdot E[X]]$$

$$= E[N] \cdot E[X]$$

Also

$$\text{Var}(Y | N=n)$$

$$= \text{Var}(X_1 + \dots + X_N | N=n)$$

$$= \text{Var}(X_1 + \dots + X_n | N=n)$$

$$= \text{Var}(X_1 + \dots + X_n) \quad \& \text{ by indep.}$$

$$= \text{Var}(X_1) + \dots + \text{Var}(X_n)$$

$$= n \cdot \text{Var}(X)$$

replace n by N to get

$$\text{Var}(Y|N) = N \cdot \text{Var}(X)$$

By law of total variance

$$\begin{aligned} \text{Var}(Y) &= E[\text{Var}(Y|N)] + \text{Var}(E[Y|N]) \\ &= E[N \text{Var}(X)] + \text{Var}(N \cdot E[X]) \\ &= E[N] \cdot \text{Var}(X) + E[X]^2 \cdot \text{Var}(N) \end{aligned}$$

Ex

Suppose we have 2 coins

$$\text{coin 1: } P(H) = p$$

$$\text{coin 2: } P(H) = q$$

we flip coin 1 until 1st head, let N be # flips. now flip coin 2 N times,

and let Y be # heads in
coin 2 sequence. observe

$$Y = X_1 + \dots + X_N$$

each X_i is Bernoulli with
parameter q . All coins are indep.
let $E[X]$, $\text{Var}(X)$ denote common
mean & variance of X_i . So

$$E[N] = \frac{1}{p}, \quad \text{Var}(N) = \frac{1-p}{p^2}$$

↑
geometric

and

$$E[X] = q, \quad \text{Var}(X) = q(1-q)$$

Therefore

$$E[Y] = E[N] \cdot E[X] = \boxed{\frac{q}{p}}$$

and

$$\text{Var}(Y) = E[N] \text{Var}(X) + E[X]^2 \text{Var}(N)$$

$$= \boxed{\frac{q(1-q)}{p} + \frac{q^2(1-p)}{p^2}}$$