

Expected value of functions of random variables:

Recap: if X is a continuous random variable

$$E(g(X)) = \int_{-\infty}^{+\infty} g(x) f_X(x) dx$$

Def: If X, Y are cont. rand. var., then

$$E(g(X, Y)) = \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} g(x, y) f_{X,Y}(x, y) dx dy$$

$$\left(\text{cf. if } X, Y \text{ are discrete rand. var., then} \right)$$

$$E(g(X, Y)) = \sum_x \sum_y g(x, y) p(x, y)$$

Prop: $E(\cdot)$ is linear (Note: We proved this already for discrete random variables)

Pf: X, Y cont. random variables, $a, b \in \mathbb{R}$

Want to show: $E(aX + bY) = aE(X) + bE(Y)$.

$$E(ax + by) = \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} (ax + by) f_{X,Y}(x,y) dx dy$$

$$g(x,y) = ax + by$$

$$= \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} [ax f_{X,Y}(x,y) + by f_{X,Y}(x,y)] dx dy$$

$$= a \iint_{-\infty}^{+\infty} x f_{X,Y}(x,y) dx dy + b \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} y f_{X,Y}(x,y) dx dy$$

$$\left. \begin{aligned} f_X(x) &= \int_{-\infty}^{+\infty} f_{X,Y}(x,y) dy \\ f_Y(y) &= \int_{-\infty}^{+\infty} f_{X,Y}(x,y) dx \end{aligned} \right\} = a \underbrace{\int_{-\infty}^{+\infty} x f_X(x) dx}_{=E(X)} + b \underbrace{\int_{-\infty}^{+\infty} y f_Y(y) dy}_{=E(Y)}$$

$$= a E(X) + b E(Y).$$

□

$$\begin{aligned} \text{Recall: } \text{Cov}(X,Y) &= E((X-E(X))(Y-E(Y))) \\ &= E(XY) - E(X)E(Y). \end{aligned}$$

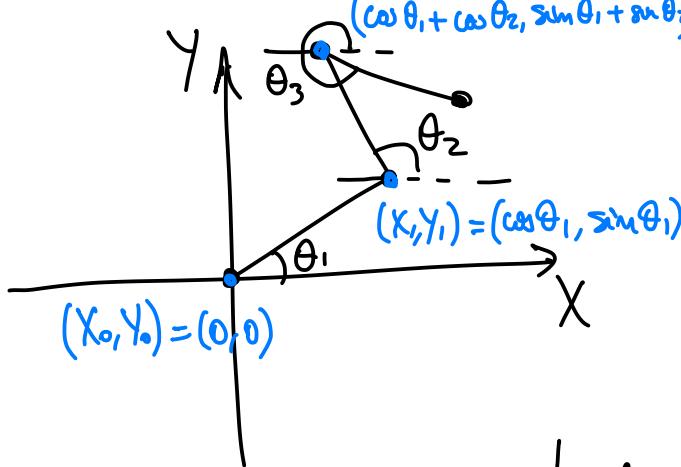
Cor: If X, Y are independent, then

$$E(XY) = E(X) \cdot E(Y).$$

Also, if g, h are continuous functions, then

$$E(g(X)h(Y)) = E(g(X))E(h(Y)).$$

Ex: Random walk in the plane, n steps:



Take steps of length 1 in a random and independent directions:
 $\theta_i \sim \text{Unif}([0, 2\pi])$
at each step $1 \leq i \leq n$.

Q: Find the expected value of the square of the distance to the origin after n steps.

Position after n steps: $\left(\sum_{i=1}^n \cos \theta_i, \sum_{i=1}^n \sin \theta_i \right)$

$$E \left(\left(\sum_{i=1}^n \cos \theta_i \right)^2 + \left(\sum_{i=1}^n \sin \theta_i \right)^2 \right) \stackrel{\text{linearity}}{=} \quad$$

$$= E \left(\left(\sum_{i=1}^n \cos \theta_i \right)^2 \right) + E \left(\left(\sum_{i=1}^n \sin \theta_i \right)^2 \right)$$

$$= E \left(\sum_{i=1}^n \cos^2 \theta_i + \sum_{i \neq j} \cos \theta_i \cos \theta_j \right)$$

$$+ E \left(\sum_{i=1}^n \sin^2 \theta_i + \sum_{i \neq j} \sin \theta_i \sin \theta_j \right)$$

linearly

$$\stackrel{!}{=} \sum_{i=1}^n E(\cos^2 \theta_i) + \sum_{i \neq j} E(\cos \theta_i \cos \theta_j)$$

$$+ \sum_{i=1}^n E(\sin^2 \theta_i) + \sum_{i \neq j} E(\sin \theta_i \sin \theta_j)$$

$$= \sum_{i=1}^n \left(E(\cos^2 \theta_i) + E(\sin^2 \theta_i) \right) + \sum_{i \neq j} E(\cos \theta_i \cos \theta_j) + E(\sin \theta_i \sin \theta_j)$$

$$= \sum_{i=1}^n E(\underbrace{\cos^2 \theta_i + \sin^2 \theta_i}_1) + \sum_{i \neq j} E(\cos \theta_i) E(\cos \theta_j) + E(\sin \theta_i) E(\sin \theta_j)$$

$$\begin{cases} E(\cos \theta_i \cos \theta_j) = E(\cos \theta_i) E(\cos \theta_j) \text{ by } \underline{\text{indep.}} \\ E(\sin \theta_i \sin \theta_j) = E(\sin \theta_i) E(\sin \theta_j) \end{cases}$$

$$= n + 0$$

$$= n.$$

$$\begin{cases} E(\cos \theta_i) = \frac{1}{2\pi} \int_0^{2\pi} \cos \theta_i d\theta_i = 0 \\ E(\sin \theta_i) = \frac{1}{2\pi} \int_0^{2\pi} \sin \theta_i d\theta_i = 0 \end{cases}$$

$$A = E(\text{square dist. to origin}) = \# \text{ steps} = n.$$

Conditional expectation:

Discrete random variables:

Probability mass function of X given that $Y=y$:

$$P_{X|Y}(x,y) = P(X=x | Y=y) = \frac{P(X=x, Y=y)}{P(Y=y)} = \frac{P_{XY}(x,y)}{P_Y(y)}$$

Continuous random variables:

Probability density function of X given that $Y=y$:

$$f_{X|Y}(x,y) = \frac{f_{X,Y}(x,y)}{f_Y(y)} \quad \text{if } f_Y(y) > 0$$

Conditional expectation:

Discrete r.v.:

$$E(X | Y=y) = \sum_x x \underbrace{P(X=x | Y=y)}_{P_{X|Y}(x,y)} = \sum_x x P_{X|Y}(x,y)$$

Cont. r.v.:

$$E(X | Y=y) = \int_{-\infty}^{+\infty} x f_{X|Y}(x,y) dx$$

Conditional Expectation as a Random Variable:

$g(y) = E(X|Y=y)$ is a function of y

This defines a random variable $g(Y)$; denoted

$E(X|Y)$. ← is itself a random variable, that takes value $E(X|Y=y)$ when Y takes value y

Q: $E(E(X|Y)) = ?$

A: Conditional Expectation Formula (holds for both discrete or cont. r.v.)

$$E(E(X|Y)) = E(X).$$

Pf: Assuming X, Y discrete;

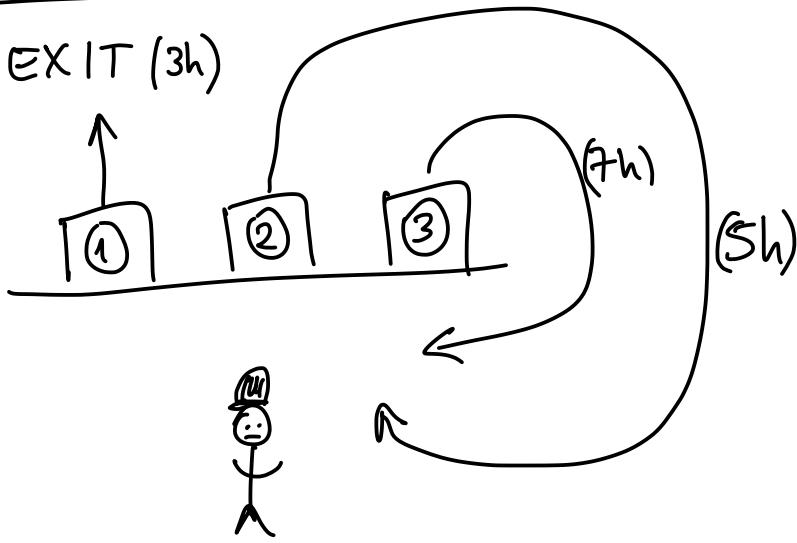
$$\begin{aligned} E(E(X|Y)) &= \sum_y E(X|Y=y) p_Y(y) \\ &= \sum_y \sum_x x p_{X|Y}(x,y) p_Y(y) \end{aligned}$$

$$\begin{aligned}
 &= \sum_y \sum_x x \cdot \frac{p_{X,Y}(x,y)}{p_X(y)} \cdot \cancel{p_Y(y)} \\
 &= \sum_x x \left(\sum_y p_{X,Y}(x,y) \right) = \sum_x x p_X(x) \\
 &\qquad\qquad\qquad \underbrace{\qquad\qquad\qquad}_{p_X(x)} \qquad\qquad\qquad = E(X).
 \end{aligned}$$

(Compare with:
 $P(E) = P(E|F)p(F) + P(E|F^c)p(F^c)$)

□

Exercise: A miner is trapped in a mine with 3 doors.



Q: What is the expected number of hours that miner will take to reach exit?

$X = \# \text{ hours until exit}$
 $Y = \text{door chosen.}$

Want: $E(X) = E(X|Y=1)p(Y=1)$
 $+ E(X|Y=2)p(Y=2)$
 $+ E(X|Y=3)p(Y=3)$

$$= \frac{1}{3} \left(\underline{\mathbb{E}(X|Y=1)} + \underline{\mathbb{E}(X|Y=2)} + \underline{\mathbb{E}(X|Y=3)} \right)$$

$$\left. \begin{array}{l} \mathbb{E}(X|Y=1) = 3 \\ \mathbb{E}(X|Y=2) = 5 + \mathbb{E}(X) \\ \mathbb{E}(X|Y=3) = 7 + \mathbb{E}(X) \end{array} \right\} \begin{aligned} &= \frac{1}{3} \left(\underline{3} + \underline{5 + \mathbb{E}(X)} + \underline{7 + \mathbb{E}(X)} \right) \\ &= \frac{1}{3} (15 + 2\mathbb{E}(X)) \end{aligned}$$

Thus: $\mathbb{E}(X) = 5 + \frac{2}{3} \mathbb{E}(X).$

$$\frac{1}{3} \mathbb{E}(X) = 5 \Rightarrow \boxed{\mathbb{E}(X) = 15}$$