

ECE 528 – Introduction to Random Processes in ECE

Lecture 10: More on Multiple Random Variables

No Slides - This Lecture Was Given Using My iPad – Summary Class Notes

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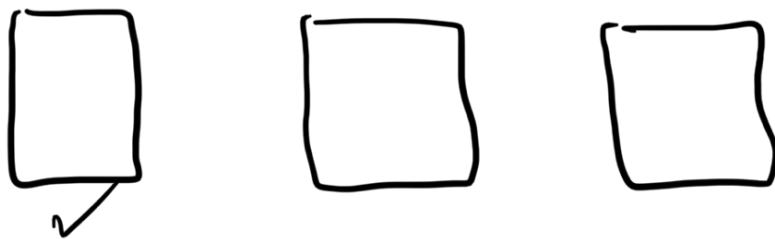
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Monty Hall Problem



2 Goats (G) Do Not Switch Switch
 1 Ferrari (F) $P(W) = 1/3$ $P(W) = 2/3$
 $P(L) = 2/3$ $P(L) = 1/3$

<u>You Pick</u>	<u>Prize Door</u>	<u>No Sw</u>	<u>Sw</u>
1	1	W	L
1	2	L	W
1	3	L	W
2	1	L	W
2	2	W	L
2	3	L	W
3	1	L	W
3	2	L	W

3

3

$$\frac{L}{W}$$

$$\frac{3 \text{ wins}}{1/3}$$

$$\frac{L}{6 \text{ wins}} = \frac{2}{3}$$

Pick door #1

Monty shows a behind #2

event A: Ferrari behind #1

event B: ^{Car} open door #2 showing G

$$P(A|B) = \frac{P(B|A) \cdot P(A)}{P(B)}$$

$$= \frac{1/2 \times 1/3}{1/3 \times 1/2 + 1/3 \times 0 + 1/3 \times 1}$$

$$= 1/3$$

more on Conditional Exp.

$E[X | Y=y]$ exists for
every y in Y .

a R.V.
 \downarrow
 $E[E[X | Y]] = E[X]$

Law of Iterated Expectations

Discrete RVs

$$E[E[X | Y]] = \sum_{y \in R_Y} \overbrace{E[X | Y=y]}^{\text{Cond. Exp.}} P_Y(y)$$

$$= \sum_{y \in R_Y} \sum_{x \in R_X} x P_{X|Y=y}(x) P_Y(y)$$

$$\vdots$$
$$= E[X]$$

For Cont. RVs

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

$$\begin{aligned}
 &= \int_{-\infty}^{\infty} \underbrace{\int_{-\infty}^{\infty} u f_{x|Y}(u) du}_{\text{Cond. Exp.}} f_Y(y) dy \\
 &= \vdots \\
 &= E[X]
 \end{aligned}$$

Example

roll a dice, whatever number
comes up toss a coin that
many times.

Q: What is the expected number
of Heads?

Please do it at home!

Conditional Variance

If X is a cont. RV

$$E[X^2 | A] = \int_A x^2 f(x|A) dx$$

$$\text{Var}[X|A] = E[X^2|A] - [E[X|A]]^2$$

If X is a discr. RV

$$E[X^2 | A] = \sum_A x^2 p(x|A)$$

$$\text{Var}[X|A] = E[X^2|A] - [E[X|A]]^2$$



• Once X is observed,

$E[Y|X]$ is a numb

• Before X is observed

$E[Y|X]$ is a RV

• If X & Y are indep RVs,
then

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

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

and $\text{Var}[Y|X] = \text{Var}[Y]$

• If X & Y RVs, a & b cons
then

$$E[a + bY|X] = a + bE[Y|X]$$

Proof

$$E[\underbrace{a + bY}_{g(Y)} | X = x] = \int (a + by) \cdot f_{Y|X}(y|x) dy$$

$$= a \underbrace{\int f_{Y|X}(y|x) dy}_{=1} + b \underbrace{\int y f_{Y|X}(y|x) dy}_{E[Y|X=x]}$$

$$= a + bE[Y|X] \quad \checkmark$$

Conditional Expectation-

For X & Y disc. RVs,

$$E[Y | X = a] = \sum_Y y \cdot P_{Y|X}(y|x)$$

↑
cond. mean

conditional variance

$$\begin{aligned} V(Y | X = a) &= \sum (y - E[Y | X = a])^2 \\ &\quad P_{Y|X}(y|a) \\ &= E[Y^2 | X = a] - (E[Y | X = a])^2 \end{aligned}$$

In general

$$E[h(Y) | X = a] = \sum_Y h(y) P_{Y|X}(y|x)$$

For X & Y continuous RVs,

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

$$\begin{aligned}\text{Var}[Y | X=a] &= \int_{-\infty}^{\infty} [y - E(Y | X=a)]^2 f_{Y|X}(y|x) dy \\ &= E[Y^2 | X=a] - (E[Y | X=a])^2\end{aligned}$$

In general

$$E[h(Y) | X=a] = \int_{-\infty}^{\infty} h(y) f_{Y|X}(y|x) dy$$

Example: X & Y cont. RVs
with joint pdf

$$f_{X,Y}(x, y) = \begin{cases} e^{-y} & y > 0, 0 \leq x \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

Q: Find $E[Y]$, $\text{Var}[Y]$, $E[Y^2]$, $E[Y | X=a]$

$$V(X|Y=2) = ?$$

Conditional Expectation

For RVs X & Y , we have

$$V(X) = V[E(X|Y)] + E[V(X|Y)]$$

Proof

$$\textcircled{1} \quad \text{Var}[X] = E[X^2] - E[X]^2$$

$$\textcircled{2} \quad \text{Var}[X|Y=y] = E[X^2|Y=y] - E[X|Y=y]^2$$

$$\Rightarrow \text{Var}[X]$$

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

use
Iterative
Expect.

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

↓

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

$$\begin{aligned}
&= E \left[\text{Var} [X | Y=y] + E [X | Y=y] \right] \\
&\quad - E \left[E [X | Y=y] \right]^2 \\
&= E \left[\text{Var} [X | Y=y] \right] + E \left[E [X | Y=y] \right] \\
&\quad - E \left[E [X | Y=y] \right]^2 \\
&= E \left[\text{Var} [X | Y=y] \right] + \text{Var} [E [X | Y=y]] \\
&= \text{Var} [X]
\end{aligned}$$

Covariance & Correlation

RV X & Y

μ_x & μ_y means

σ_x^2 & σ_y^2 variances

$$\text{Cov}(X, Y) = \left\{ \sum_i \sum_j (x_i - \mu_x) \cdot (y_j - \mu_y) \right\}$$

Continuous

$f(x, y)$
discrete

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} (x - \mu_x)(y - \mu_y) f(x, y) dx dy$$

If X & Y are indep.

$$\begin{aligned} \text{Cov}(X, Y) &= E[X - \mu_x] E[Y - \mu_y] \\ &= (\mu_y - \mu_x)(\mu_y - \mu_y) = 0 \end{aligned}$$

$$(X - \mu_x)(Y - \mu_y) = XY - X\mu_y - \mu_x Y + \mu_x \mu_y$$

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

Ex rolling 2 fair dice

\downarrow
 $\{1, 2, \dots, n\}$

x \ y	0	1	2
0	$21/36$	$4/36$	$25/36$
1	$8/36$	$27/36$	$10/36$
2	$1/36$	0	$1/36$
$f_Y(y)$	$30/36$	$6/36$	1

$$E[XY] = \sum_i \sum_j x_i y_j f(x_i, y_j)$$

$$= 1 \cdot 1 \cdot \frac{2}{36} = \frac{1}{18}$$

$$\mu_X = E[X] = \sum_i x_i f_X(x_i)$$

$$= 0 \cdot \frac{25}{36} + 1 \cdot \frac{10}{36} + 2 \cdot \frac{1}{36}$$

$$\mu_Y = E[Y] = \sum_j y_j f_Y(y_j)$$

$$= \frac{1}{3}$$

$$= \frac{30}{36} \cdot 0 + \frac{6}{36} \cdot 1 + 1 \cdot 2$$

$$= 0 - \frac{1}{36} + \frac{1}{36} = \frac{1}{6}$$

$$\text{cov}(X, Y) = \frac{1}{18} - \frac{1}{3} \cdot \frac{1}{6} = 0$$

X & Y are NOT independent

$$\text{cov} = 0$$

$$P(2, 1) = 0 \neq \frac{1}{36} \left(\frac{1}{36} \right) = P_X(2)P_Y(1)$$

Example

$$f_{X,Y}(x, y) = \begin{cases} x e^{-x(y+1)} & 0 < x < \infty \\ & 0 < y < \infty \\ 0 & \text{otherwise} \end{cases}$$

$$h_X = 1$$

h_Y : doesn't exist

→ ...

→ $\text{Cov}(X, Y)$ as undefined

$$E[XY] = \int_0^{\infty} \int_0^{\infty} xy e^{-x(y+1)} dx dy = 1$$

Example

$$f_{XY}(x, y) = \begin{cases} e^{-(x+y)} & 0 < x < \infty \\ & 0 < y < \infty \\ 0 & \text{otherwise} \end{cases}$$

$$h_x = h_y = 1$$

Covariance ?

$$E[XY] - h_x h_y$$

$$E[XY] = \int_0^{\infty} \int_0^{\infty} xy e^{-(x+y)} dx dy =$$

-

$$E(XY) - \mu_x \mu_y = 1 - 1 = 0$$

$$X^* = \frac{X - \mu_x}{\sigma_x}$$

$$Y^* = \frac{Y - \mu_y}{\sigma_y}$$

$$\rho = \text{Cov}(X^*, Y^*) = \frac{\text{Cov}(X, Y)}{\sigma_x \sigma_y}$$

↑
correlation
coeff.

① $-1 \leq \rho \leq +1$

② if X & Y indep then $\rho = 0$

③ X & Y satisfy linear
identity $Y = aX + b$
∴

$$\rho = \pm 1$$

$$X \text{ \& } Y \text{ RV}$$

$$S = X + Y \quad \text{Variance?}$$

$$E[S] = \mu_x + \mu_y$$

$$\text{Var}[S] = \text{Var}[X] + \text{Var}[Y] \quad \text{if } X \text{ \& } Y \text{ ind}$$

In general

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

$$= E[((X - \mu_x) + (Y - \mu_y))^2]$$

$$= E[(X - \mu_x)^2] + E[(Y - \mu_y)^2]$$

$$+ 2 E[(X - \mu_x)(Y - \mu_y)]$$

$$\boxed{\text{Var}[S] = \text{Var}[X] + \text{Var}[Y] + 2 \text{Cov}[X, Y]}$$

Move on jointly Gaussian RVs

Change of coordinates from Cartesian to polar.

obtaining Rayleigh distribution

Rayleigh \longrightarrow Exponential distribution