Lecture 19: More Than Two Random Variables

Definition

If $X_1, X_2, ..., X_n$ are discrete random variables defined on the same sample space then their joint pmf is the function

$$P_{X_1,X_2,...,X_n}(x_1,x_2,...,x_n) = P(X_1 = x_1,...,X_n = x_n)$$

If $X_1, X_2, ..., X_n$ are continuous then their joint pdf is the function $f_{X_1, X_2, ..., X_n}(x_1, x_2, ..., x_n)$ such that

Definition (Cont.)

for any region A in \mathbb{R}^n

$$P((X_1, X_2, \dots, X_n) \in A) = \underbrace{\int \dots \int_A f_{X_1, X_2, \dots, X_n}(x_1, \dots, x_n) dx_1, \dots, dx_n}_{n\text{-fold multiple integral}}$$

Definition

The discrete random variables $X_1, X_2, ..., X_n$ are independent if

$$P_{X_1,...,X_n}(x_1,...,x_n) = P_{X_1}(x_1)P_{X_2}(x_2)...P_{X_n}(x_n).$$

Equivalently

$$P(X_1 = x_1, ..., X_n = x_n) = P(X_1 = x_1) ... P(X_n = x_n)$$

The continuous random variables X_1, X_2, \dots, X_n are independent if

$$f_{X_1,X_2,...,X_n}(x_1,x_2,...,x_n) = f_{X_1}(x_1)f_{X_2}(x_2)...f_{X_n}(x_n)$$

Definition

 X_1, X_2, \dots, X_n are pairwise independent if each pair $X_i, X_j (i \neq j)$ is independent. We will now see

Pairwise independence

→ Independence

of random variables

← of random variables

First we will prove ←

Theorem.

 X_1, X_2, \dots, X_n independent $\Rightarrow X_1, X_2, \dots, X_n$ are pairwise independent.

From now on we will restrict to the case n=3 so we have THREE random variables X, Y, Z.

How do we get

$$P_{X,Y}(x,y)$$
 from $P_{X,Y,Z}(x,y,z)$

Answer (left to you to prove)

$$P_{X,Y}(x,y) = \sum_{\text{all } z} P_{X,Y,Z}(x,y,z) \tag{\#}$$

Now we can prove X, Y, Z independent.

 \implies X, Y independent

Since X, Y, Z are independent we have

$$P_{X,Y,Z}(x,y,z) = P_X(x)P_Y(y)P_Z(z)$$
 (##)

Now play the RHS of (##) into the RHS of (#)

$$P_{X,Y}(x,y) = \sum_{\text{all } z} P_X(x) P_Y(y) P_Z(z)$$

$$= P_X(x) P_Y(y) \left(\sum_{\text{all } z} P_Z(z) \right)$$

$$= P_X(x) P_Y(y)$$

This proves X and Y are independent Identical proofs prove the pairs X, Z and Y, Z are independent.

Now we construct X, Y, Z (actually X_A , X_B , X_C) so that each <u>pair</u> is independent but the triple X, Y, Z is <u>not</u> independent.

A Variation on the Cool Counter example

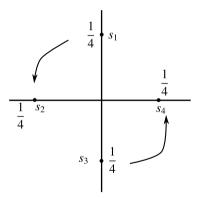
Lets go back to the "cool counter example", Lecture 16, page 18 of three events A, B, C which are pairwise independent but no independent so

$$P(A \cap B \cap C) \neq P(A)P(B)P(C)$$

The idea is to convert the three events to random variables X_A , X_B , X_C so that $X_A = 1$ on A and O on A' etc.

In fact we won't need the corner points (-1,-1), (-1,1), (1,-1) and (1,1) we put $S_1=(0,1)$, $S_2=(-1,0)$, $S_3=(0,1)$, $S_4=(1,0)$ and retain their probabilities so

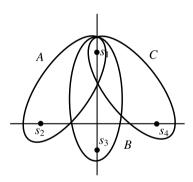
$$P(\{S_j\}) = \frac{1}{4}, \quad 1 \le j \le 4$$



We define

$$A = \{s_1, s_2\}$$

 $B = \{s_1, s_3\}$
 $C = \{s_1, s_4\}$



We define X_A , X_B , X_C on S by

$$X_A(s_j) = \begin{cases} 1, & \text{if } s_j \in A \\ 0, & \text{if } s_j \notin A \end{cases}$$

$$X_B(s_j) = egin{cases} 1, & ext{if } s_j \in B \ 0, & ext{if } s_j \notin B \end{cases}$$
 $X_C(s_j) = egin{cases} 1, & ext{if } s_j \in C \ 0, & ext{if } s_j \notin C \end{cases}$

So
$$P(X_A = 1) = P(\{S_1, S_2\}) = \frac{1}{2}$$

 $P(X_A = 0) = P(\{S_3, S_4\}) = \frac{1}{2}$

and similarly for X_B and X_C .

So
$$X_A, X_B$$
 and X_C

are Bernoulli random variables

Let's compute the joint pmf of X_A and X_B . We know the margin

X_A	0	1	
0			1/2
1			1/2
	1/2	1/2	

The subset where $X_A = 1$ is the subset $\{s_1, s_2\}$ so we write an equality of events

$$(X_A = 1) = \{s_1, s_2\}$$

Similarly

$$(X_A = 0) = \{s_3, s_4\}$$

 $(X_B = 1) = \{s_1, x_3\}, (X_B = 0) = \{s_2, s_4\}$
 $(X_C = 1) = \{s_1, s_4\}, (X_C = 0) = \{s_2, s_3\}$

Hence

$$(X_A = 0) \cap (X_B = 0) = \{S_4\}$$
so $P(X_A = 0, X_B = 0) = \frac{1}{4}$

$$(X_A = 0) \cap (X_B = 1) = \{S_3\}$$
so $P(X_A = 0, X_B = 1) = \frac{1}{4}$

$$(X_A = 1) \cap (X_B = 0) = \{S_2\}$$

$$P(X_A = 1, X_B = 0) = \frac{1}{4}$$

$$(X_A = 1) \cap (X_B = 1) = \{S_1\}$$

$$P(X_A = 1, X_B = 1) = P(\{S_1\}) = \frac{1}{4}$$

etc.

So the joint proof of X_A and X_B is

X_A	0	1	
0	1/4	1/4	1/2
1	1/4	1/4	1/2
	1/2	1/2	

so X_A and X_B are independent. The same is true for X_A and X_C and X_B and X_C . Now we show the triple X_A , X_B and X_C is NOT independent.

We will show

$$P(X_A = 1, X_B = 1, X_C = 1)$$

 $\neq P(X_A = 1)P(X_B = 1)P(X_C = 1)$

The RHS =
$$\left(\frac{1}{2}\right)\left(\frac{1}{2}\right)\left(\frac{1}{2}\right) = \frac{1}{8}$$

The left-hand side is the probability of the event

$$(X_A = 1) \cap (X_B = 1) \cap (X_C = 1)$$

= $\{S_1, S_2\} \cap \{S_1, S_3\} \cap \{S_1, S_4\}$
= $\{S_1\}.$

So
$$P(X_A=1,X_B=1,X_C=1)=P(\{S_1\})=\frac{1}{4}$$
 so
$$LHS=\frac{1}{4}$$

$$RHS=\frac{1}{8}$$

Remark

This counter example is more or less the some as the "cool counter example". We just replaced (more or less) A, B, C by their "characteristic functions".