Lecture 2.2: Independence

2013/09/11

Previously... Conditional Probability

The conditional probability of an event A, given the event B, is defined by

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

provided $\mathbb{P}(B) > 0$.

This is read as "the probability of A given B."

Previously... Multiplication Rule (of Probability)

Another way to consider conditional probability is via the multiplication rule:

For any events A and B

$$\mathbb{P}(A \cap B) = \mathbb{P}(A|B) \times \mathbb{P}(B) = \mathbb{P}(B|A) \times \mathbb{P}(A)$$

For any n events E_1, \ldots, E_n

$$\mathbb{P}(E_1 \cap E_2 \cap \ldots \cap E_n)$$

$$= \mathbb{P}(E_2 \cap \ldots \cap E_n | E_1) \times \mathbb{P}(E_1)$$

$$= \mathbb{P}(E_3 \cap \ldots \cap E_n | E_1, E_2) \times \mathbb{P}(E_2 | E_1) \times \mathbb{P}(E_1)$$

$$= \ldots$$

$$= \mathbb{P}(E_n | E_1, \ldots, E_{n-1}) \times \ldots \times \mathbb{P}(E_2 | E_1) \times \mathbb{P}(E_1)$$

Previously... Monty Hall Problem

Assume w/o loss of generality that the car is behind Door 1:

$$\mathbb{P}(\mathsf{Open}\ 2\ \cap\ \mathsf{Picked}\ 1) \ = \ \mathbb{P}(\mathsf{Open}\ 2\ |\ \mathsf{Picked}\ 1) \times \mathbb{P}(\mathsf{Picked}\ 1) = \frac{1}{2} \times \frac{1}{3} = \frac{1}{6}$$

$$\mathbb{P}(\mathsf{Open}\ 3\ \cap\ \mathsf{Picked}\ 1) \ = \ \mathbb{P}(\mathsf{Open}\ 3\ |\ \mathsf{Picked}\ 1) \times \mathbb{P}(\mathsf{Picked}\ 1) = \frac{1}{2} \times \frac{1}{3} = \frac{1}{6}$$

$$\mathbb{P}(\mathsf{Open}\ 1\ \cap\ \mathsf{Picked}\ 2) \ = \ \mathbb{P}(\mathsf{Open}\ 1\ |\ \mathsf{Picked}\ 2) \times \mathbb{P}(\mathsf{Picked}\ 2) = 0$$

$$\mathbb{P}(\mathsf{Open}\ 3\ \cap\ \mathsf{Picked}\ 2) \ = \ \mathbb{P}(\mathsf{Open}\ 3\ |\ \mathsf{Picked}\ 2) \times \mathbb{P}(\mathsf{Picked}\ 2) = \frac{1}{3}$$

$$\mathbb{P}(\mathsf{Open}\ 1\ \cap\ \mathsf{Picked}\ 3) \ = \ \mathbb{P}(\mathsf{Open}\ 1\ |\ \mathsf{Picked}\ 3) \times \mathbb{P}(\mathsf{Picked}\ 3) = 0$$

$$\mathbb{P}(\mathsf{Open}\ 2\ \cap\ \mathsf{Picked}\ 3) \ = \ \mathbb{P}(\mathsf{Open}\ 2\ |\ \mathsf{Picked}\ 3) \times \mathbb{P}(\mathsf{Picked}\ 3) = \frac{1}{3}$$

We'll look at another (perhaps more intuitive) solution once we've seen the law of total probability.

Goals for Today

- ► Example of multiplication rule
- ► Define independence
- Examples

Multiplication Rule Example

An ordinary deck of 52 playing cards is randomly divided into 4 piles of 13 cards each. Compute the probability that each pile has exactly 1 ace.

Define the following events E_i , i = 1, 2, 3, 4 as follows:

- $E_1 = \{ \text{The ace of spades (WLOG) is in any of the four piles} \}$
- $E_2 = \{ \text{The ace of spades and hearts are in different piles} \}$
- **...**

Independence

Independence and Conditional Probability

Suppose we toss two fair dice. Let E denote the event that the sum of the dice is six and F denote the event that the first die equals four. Show these events are not independent:

Mutual Independence

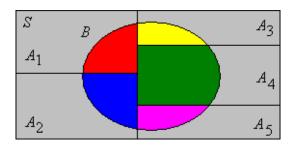
Pairwise independent events that are not mutually independent:

Three brands of IPA's (X, Y, Z) are to be ranked according to taste by a judge. Let

- 1. Event A: Brand X is preferred over Y
- 2. Event B: Brand X is ranked best
- 3. Event C: Brand X is ranked second best.
- 4. Event D: Brand X is ranked third best

If the judge is actually a fraud and ranks the beers randomly, is event A independent of B, C, and D?

Next Time: The Law of Total Probability



$$\mathbb{P}(B) = \mathbb{P}(B|A_1) \cdot \mathbb{P}(A_1) + \mathbb{P}(B|A_2) \cdot \mathbb{P}(A_2) + \mathbb{P}(B|A_3) \cdot \mathbb{P}(A_3) + \mathbb{P}(B|A_4) \cdot \mathbb{P}(A_4) + \mathbb{P}(B|A_5) \cdot \mathbb{P}(A_5)$$