

Lectures of May 12 2006 - AM

Scribe: Robert Muntean, Mohamed Ghadie

## 1 Independence of Events and Experiments

Recall from last lecture:

From last example, we can note some definitions that are used for cancer tests such as:

Sensitivity: used to express how good a test can detect cancer given that a person has cancer X.

$$\text{Sensitivity} = P[+ / X]$$

Specificity: used to express how good a test can detect that a person does not have cancer given that the person does not really have cancer X.

$$\text{Specificity} = P[- / X^c]$$

False negative rate: used to express the chances that the test would not detect cancer given that the person tested has cancer X.

$$\text{False negative rate} = P[- / X]$$

False positive rate: used to express the chances that the test would detect cancer in a person given that the tested person does not have cancer X.

$$\text{False positive rate} = P[+ / X^c]$$

Using the previously proven identity  $P[A/B] + P[A^c/B] = 1$

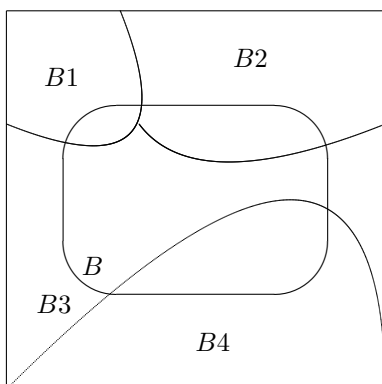
We can conclude the following:

$$\text{Sensitivity} + \text{false negative rate} = 1.$$

$$\text{Specificity} + \text{false positive rate} = 1.$$

We also recall that if a sample space S and a set of events  $B_1, B_2, \dots, B_n$  are defined such that  $S = B_1 \cup B_2 \cup \dots \cup B_n$ ,

$$\sum P[B_i/B] = 1 \text{ as the figure below illustrates.}$$



### 1.1 Independent Events

Events A and B are said to be independent if  $P[A \cap B] = P[A] \cdot P[B]$ .

#### Example

”Card-drawing” experiment:

Pick a card from a regular deck.

Let A be the event that a drawn card is red.

Let B be the event that the card is an Ace.

Are A and B independent?

Solution:  $P[A \cap B] = 2 \cdot \frac{1}{52} = \frac{1}{26}$

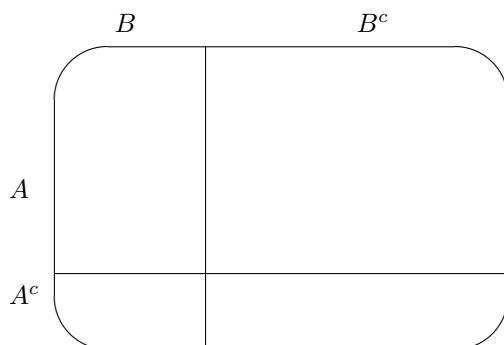
$$P[A] = 26 \cdot \frac{1}{52} = \frac{1}{2}$$

$$P[B] = 4 \cdot \frac{1}{52} = \frac{1}{13}$$

$$P[A \cap B] = P[A] \cdot P[B]$$

Therefore A and B are independent.

Consider sample space S with events A and its complement and B and its complement



If A and B are independent, it's easy to show that:

$$P[A/B] = \frac{P[A \cap B]}{P[B]} = \frac{P[A] \cdot P[B]}{P[B]} = P[A] \quad (\text{if } P[B] \neq 0)$$

$$P[B/A] = P[B] \quad (\text{if } P[A] \neq 0)$$

$$P[A/B^c] = P[A]$$

Proof:

$$P[A/B^c] = \frac{P[A \cap B^c]}{P[B^c]} = \frac{P[A] - P[A \cap B]}{P[B^c]} = \frac{P[A] - P[A] \cdot P[B]}{P[B^c]} = \frac{P[A](1 - P[B])}{P[B^c]} = P[A]$$

From this, we see that

$$P[A \cap B^c] = P[B^c] \cdot P[A]$$

$\Rightarrow$  A and  $B^c$  are independent

Example:

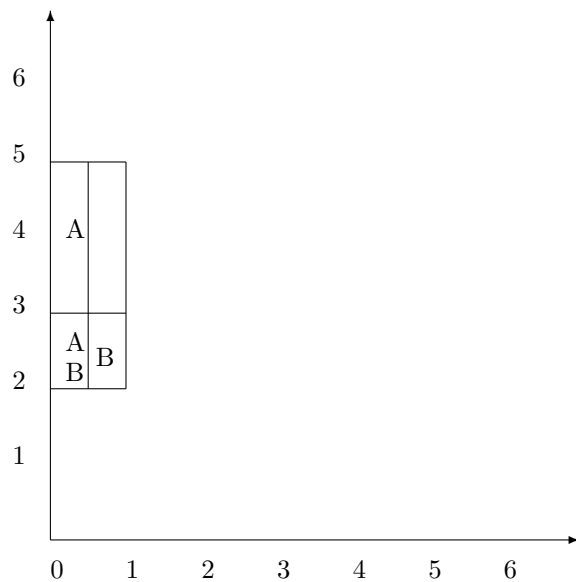
First pick a number X uniformly randomly from interval (0, 1), and pick another number Y uniformly randomly from interval (2, 5)

$$A = [X \leq \frac{1}{2}]$$

$$B = [Y \leq 3]$$

$$C = [X + Y \geq 3]$$

Which pair(s) of events are independent?



$$P[A \cap B] = \frac{1}{2} \cdot \frac{1}{3} = \frac{1}{6}$$

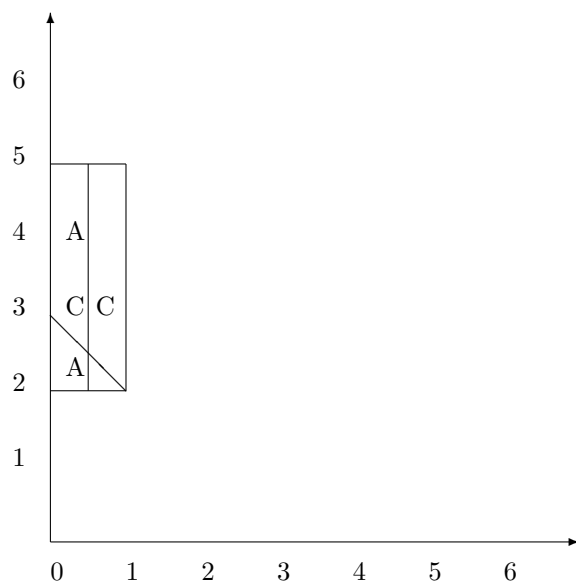
$$P[A] = \frac{1}{2}$$

$$P[B] = \frac{1}{3}$$

$$P[A \cap B] = P[A] \cdot P[B]$$

$\Rightarrow$  A and B are independent

This result agrees with our expectations because event A operates on some interval of X, whose limits do not depend on the value of Y, and event B operates on some interval of Y, whose limits do not depend on the value of X.



$$P[C] = \frac{5}{6}$$

$$P[A \cap C] = \frac{1.125}{3} = \frac{5}{12}$$

$$P[A] \cdot P[C] = \frac{5}{12} \neq P[A \cap C]$$

Therefore A and C are not independent.

This result was also expected because the value of X will definitely affect the result of X + Y. So the probability that event C occurs depends on the probability that event A occurs.

## 1.2 Independence of Multiple Events

Events  $A_1, A_2, \dots, A_n$  are said to be independent if for any integer  $k \geq 2$  and any choice of  $i_1, i_2, \dots, i_k$  such that  $1 \leq i_1 < i_2 < i_3 < \dots < i_k \leq n$ ,

$$P[A_{i_1} \cap A_{i_2} \cap \dots \cap A_{i_k}] = P[A_{i_1}] \cdot P[A_{i_2}] \cdot \dots \cdot P[A_{i_k}]$$

For example, events A, B, and C are said to be independent if

$$P[A \cap B] = P[A] \cdot P[B]$$

$$P[A \cap C] = P[A] \cdot P[C]$$

$$P[B \cap C] = P[B] \cdot P[C]$$

$$P[A \cap B \cap C] = P[A] \cdot P[B] \cdot P[C]$$

### Example

Pick a number X from set S  $\{0, 1\}$  at random where each number is picked with equal likelihood, then pick a number Y from set S  $\{0, 1\}$  where each number is picked with equal likelihood.

Compute  $Z := (X + Y) \bmod 2$   
 Record  $(X, Y, Z)$ .

The sample space  $S := \{000, 011, 101, 110\}$   
 Check whether event  $A := [X = 1]$ ,  $B := [Y = 1]$  and  $C := [Z = 1]$  are independent.

$$P[A] = 2 \cdot \frac{1}{4} = \frac{1}{2}$$

$$P[B] = 2 \cdot \frac{1}{4} = \frac{1}{2}$$

$$P[C] = 2 \cdot \frac{1}{4} = \frac{1}{2}$$

$$P[A \cap B] = P[110] = \frac{1}{4}$$

$$P[A \cap C] = P[101] = \frac{1}{4}$$

$$P[B \cap C] = P[011] = \frac{1}{4}$$

$$P[A \cap B] = P[A] \cdot P[B]$$

$$P[B \cap C] = P[B] \cdot P[C]$$

$$P[A \cap C] = P[A] \cdot P[C]$$

$$P[A \cap B \cap C] = P[111] = P[\Phi] = 0$$

$$\text{But } P[A] \cdot P[B] \cdot P[C] = \frac{1}{8}$$

Therefore A, B, and C are not independent; consistent with our intuition.

But note: A and B are independent, B and C are independent, and A and C are independent.

By just looking at the elements of S, we see that the value of the first bit X alone does not tell us whether the third bit Z is a 1 or 0. Also the value of the second bit Y alone does not tell us whether the third bit Z is a 1 or 0. But we can see that  $Z = X \text{ xor } Y$ . That is why the three events together are dependent.

We conclude that a set of events may not be totally independent even though it is pairwise independent.

### 1.3 Independence of Experiments

Independence of experiments are typical assumptions for a sequence of experiments that are not related in any way.

Suppose we perform a sequence of experiments  $E_1, E_2, \dots, E_n$ , and suppose these experiments are independent.

Let  $S_1, S_2, \dots, S_n$  be the sample space of  $E_1, E_2, \dots, E_n$ . The assumption that these experiments are independent is understood as follows:

For any sequence of events  $A_1, A_2, \dots, A_n$  with  $A_i \subseteq S_i$ ,

$$P[A_1 \times A_2 \times \dots \times A_n] = P[A_1 \times S_2 \times S_3 \times \dots \times S_n] \cdot P[S_1 \times A_2 \times S_3 \times \dots \times S_n] \cdot P[S_1 \times S_2 \times \dots \times S_n]$$

$$A_3 \cdots \times S_n] \cdot \cdots \cdot P[S_1 \times S_2 \times \cdots \times S_{n-1} \times A_n]$$

Note that  $P[A_i \times S_{i+1} \times \cdots \times S_n]$  is the probability that event  $A_i$  happens in Experiment  $E_i$  regardless of the outcomes of the other experiments because this cross product gives us all patterns of the combination of  $A_i$  with the outcomes of all other experiments.

Aside Notation: For any 2 sets A and B, the cartesian product  $A \times B$  is the set  $\{(a, b) : a \in A, b \in B\}$

### Example

Pick a card from a deck and put it back. Do it again. Record the pair of cards (with order).

$$P[A, 3] = \frac{4 \times 4}{52 \times 52} = \frac{1}{169}$$

where  $P[A, 3]$  is the probability of the first card being an Ace and the second being a 3

$$P[A, \text{anything}] = \frac{4 \times 52}{52 \times 52} = \frac{1}{13}$$

$$P[\text{anything}, 3] = \frac{52 \times 4}{52 \times 52} = \frac{1}{13}$$

$$\Rightarrow P[A, 3] = P[A, \text{anything}] \cdot P[\text{anything}, 3]$$