

LECTURE 3

- Readings: Sections 1.5

Lecture outline

- Review
- Independence of two events
- Independence of a collection of events

Review

$$\mathbf{P}(A \mid B) = \frac{\mathbf{P}(A \cap B)}{\mathbf{P}(B)}, \text{ assuming } \mathbf{P}(B) > 0.$$

- Multiplication rule:

$$\mathbf{P}(A \cap B) = \mathbf{P}(B) \cdot \mathbf{P}(A \mid B) = \mathbf{P}(A) \cdot \mathbf{P}(B \mid A)$$

- Total probability theorem:

$$\mathbf{P}(B) = \mathbf{P}(A)\mathbf{P}(B \mid A) + \mathbf{P}(A^c)\mathbf{P}(B \mid A^c)$$

- Bayes rule:

$$\mathbf{P}(A_i \mid B) = \frac{\mathbf{P}(A_i)\mathbf{P}(B \mid A_i)}{\mathbf{P}(B)}$$



Extended Radar Example



① ? ②

- Threat alert affects the outcome

$P(\dots | \text{Threat})$

		Radar		
		Low(0)	Medium(?)	High(1)
Airplane	Absent	0.1125	0.05	0.0125
	Present	0.055	0.22	0.55

$P(\dots | \text{No Threat})$

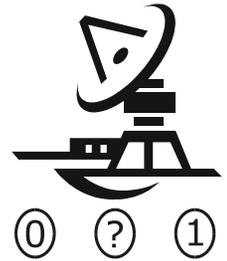
		Radar		
		Low(0)	Medium(?)	High(1)
Airplane	Absent	0.45	0.20	0.05
	Present	0.02	0.08	0.20

- $P(\text{Threat}) = \text{Prior probability of threat} = p$



Extended Radar Example

(continued)



- A =Airplane, R =Radar Reading

$$P(A, R) = P(\text{Threat})P(A, R|\text{Threat}) + P(\text{No Threat})P(A, R|\text{No Threat})$$

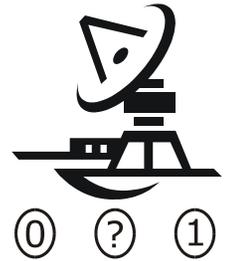
- If we let $p = P(\text{Threat})$, then we get:

		Radar		
		Low(0)	Medium(?)	High(1)
$P(A, R)$	Airplane			
	Absent	$0.45 - 0.3375p$	$0.20 - 0.15p$	$0.05 - 0.0375p$
	Present	$0.02 + 0.0145p$	$0.08 + 0.14p$	$0.20 + 0.35p$



Extended Radar Example

(continued)



$P(A, R)$

		Radar		
		Low(0)	Medium(?)	High(1)
Airplane	Absent	$0.45 - 0.3375p$	$0.20 - 0.15p$	$0.05 - 0.0375p$
	Present	$0.02 + 0.0145p$	$0.08 + 0.14p$	$0.20 + 35p$

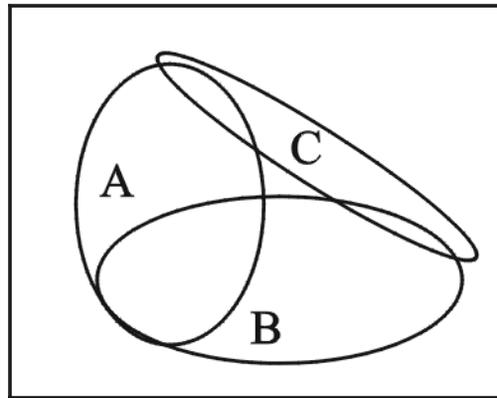
- Given the Radar registered High, and a plane was absent, What is the probability that there was a threat?
- How does the decision region behave, as a function of p ?

Independence of Two Events

- Definition: $P(A \cap B) = P(A) \cdot P(B|A)$
- Recall:
 - Independence of B from A :
$$P(B|A) = P(B)$$
 - By symmetry, $P(A|B) = P(A)$
- Examples:
 - A and B are disjoint.
 - Independence of A^c and B .
 - $P(A|B) = P(A|B^c)$

Conditioning may affect independence

- Assume A and B are independent:

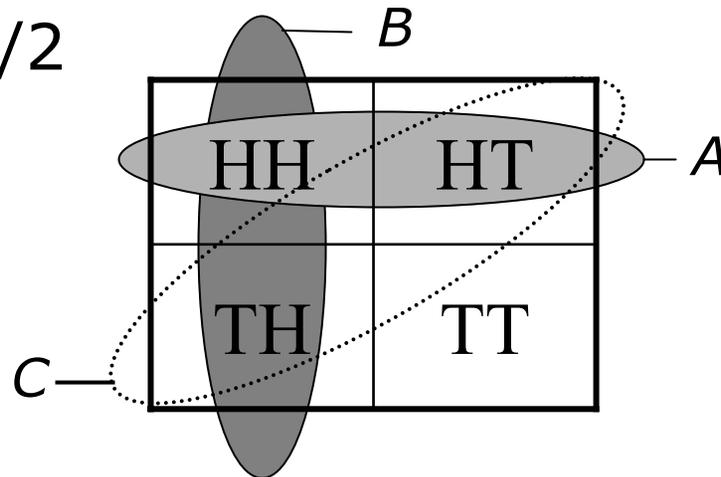


- If we are told that C occurred, are A and B independent?

Conditioning may affect independence

- Example 1:

- Two independent fair ($p=1/2$) coin tosses.
- Event A : First toss is H
- Event B : Second toss is H
- $\mathbf{P}(A) = \mathbf{P}(B) = 1/2$

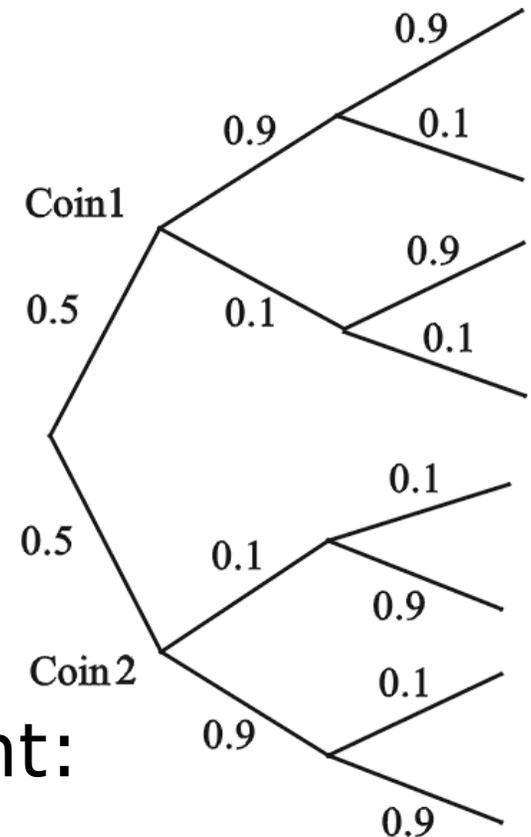


- Event C : The two outcomes are different.
- Conditioned on C , are A and B independent?

Conditioning may affect independence

- Example 2:

- Choice between two unfair coins, with equal probability.
- $P(H|\text{coin 1}) = 0.9$,
 $P(H|\text{coin 2}) = 0.1$
- Keep tossing the chosen coin.



- Are future tosses independent:

- If we know we chose coin A?
- If we do not know which coin we chose?
- Compare: $P(\text{toss } 11 = H)$

$$P(\text{toss } 11 = H \mid \text{first 10 tosses are H})$$

Independence of a Collection of Events

- Intuitive definition:

- Information about some of the events tells us nothing about probabilities related to remaining events.

- Example: $\mathbf{P}(A_1 \cap (A_2^c \cup A_3) \mid A_5 \cap A_6^c)$
 $= \mathbf{P}(A_1 \cap (A_2^c \cup A_3))$

- Mathematical definition:

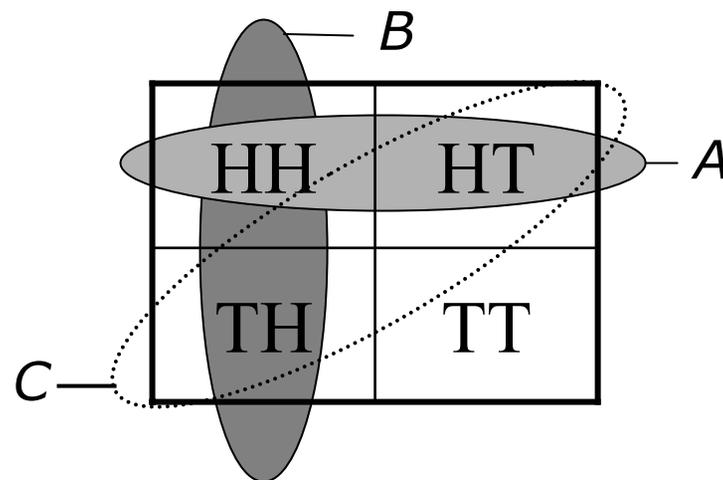
- For any distinct i, j, \dots, q :

$$\mathbf{P}(A_i \cap A_j \cap \dots \cap A_q) = \mathbf{P}(A_i)\mathbf{P}(A_j) \dots \mathbf{P}(A_q)$$

Independence vs. Pairwise Independence

- Example 1 Revisited:
 - Two independent fair ($p=1/2$) coin tosses.
 - Event A : First toss is H
 - Event B : Second toss is H
 - Event C : The two outcomes are different.

- $P(C) = P(A) = P(B) = \frac{1}{2}$
- $P(C \cap A) = \frac{1}{4}$
- $P(C | A \cap B) = 0$



- Pairwise independence **does not** imply independence.

The King's Sibling

- The king comes from a family of two children.
- What is the probability that his sibling is female?