

LECTURE 2

- **Readings:** Sections 1.3 and 1.4

Lecture outline

- Review of probabilistic models
- Conditional probability
- Three important tools:
 - Multiplication rule
 - Total probability theorem
 - Bayes' rule

Review of probabilistic models

- **Sample space** Ω of **outcomes**
- **Events** are subsets of Ω
- **Probabilities** assigned to events by $P(\cdot)$

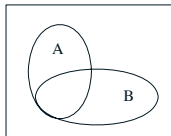
Axioms:

1. **Nonnegativity:** $P(A) \geq 0$
2. **Additivity:** If $A \cap B = \emptyset$, then $P(A \cup B) = P(A) + P(B)$
More generally, if A_1, A_2, \dots are disjoint then

$$P(A_1 \cup A_2 \cup \dots) = P(A_1) + P(A_2) + \dots$$

3. **Normalization:** $P(\Omega) = 1$

Conditional probability



- $P(A | B)$ = probability of A , given that B occurred
– B is our new universe
- **Definition:** Assuming $P(B) \neq 0$,

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

($P(A | B)$ is not defined when $P(B) = 0$)

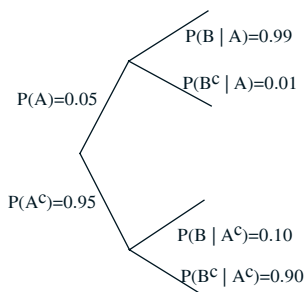
Die roll example

		first roll X			
		1	2	3	4
second roll Y	1	•	•	•	•
	2	•	•	•	•
	3	•	•	•	•
	4	•	•	•	•

- Let B be the event: $\min(X, Y) = 2$
- Let $M = \max(X, Y)$
- $P(\{M = 1\} | B) =$
- $P(\{M = 2\} | B) =$

Models based on conditional probabilities

- Event A : Airplane is flying above
Event B : Something registers on radar screen

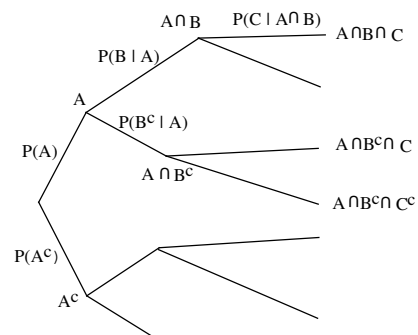


$$P(A \cap B) = ?$$

$$P(B) = ?$$

$$P(A | B) = ?$$

Multiplication rule

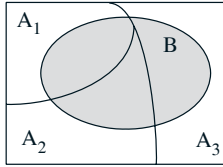


$$P(A \cap B) =$$

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

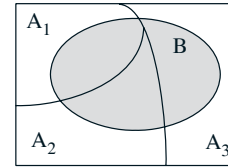
Total probability theorem

- Partition of Ω into A_1, A_2, A_3 with $P(B | A_i)$ easy for each i



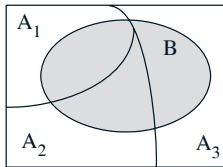
- One way of computing $P(B)$:

$$\begin{aligned} P(B) &= P(A_1)P(B | A_1) \\ &+ P(A_2)P(B | A_2) \\ &+ P(A_3)P(B | A_3) \end{aligned}$$

Bayes' rule

- Have “prior” probabilities $P(A_i)$ and $P(B | A_i)$ for each i
- Want to “update beliefs” based on B :

compute $P(A_i | B)$ for each i

Bayes' rule

$$\begin{aligned} P(A_i | B) &= \frac{P(A_i \cap B)}{P(B)} \\ &= \frac{P(A_i)P(B | A_i)}{P(B)} \\ &= \frac{P(A_i)P(B | A_i)}{\sum_j P(A_j)P(B | A_j)} \end{aligned}$$