# CSE 150A Intro to AI: Probabilistic Reasoning

Discussion Session 1
Probability Review, Bayes, marginal / conditional independence and python coding

## Agenda

- Review of concepts
- Marginal & Conditional Independence, Joint Probability
- Bayes Theorem
- Introduction to Bayesian Networks
- D-separation

# Introduction Python / Libraries

Intro to python

## **Concepts Review**

- Non-negativity: Probabilities are never negative P(A)≥0
- **2. Entire Sample Space:** the total probability over the sample space is 1  $P(A) + P(\neg A) = 1$
- **3. AND:** the probability that both event A and event B happen  $P(A, B) = P(A \land B)$   $P(A \land B) = P(A) + P(B) P(A \lor B)$
- **4. OR:** the probability that either A or B happens  $P(A \lor B) = P(A) + P(B) P(A \land B)$  (General Formula)

If A and B are **disjoint** (meaning they cannot both happen, so  $A \land B = \emptyset$ ), then:  $P(A \lor B) = P(A) + P(B)$ 

#### Concepts Review: Exercise 1

In a survey among students in a university, two events were studied:

- Some students attended an AI Club meeting.
- Some students attended a Math Workshop.

A portion of students attended both the Al Club meeting and the Math Workshop. In total, **40**% of students attended the **Al Club meeting**, **50**% attended the **Math Workshop**, and **20**% attended **both** events.

- 1. What is the probability that a student attended either the Al Club meeting or the Math Workshop (or both)?
- 2. What is the probability that a student attended neither of the two events?
- 3. Could these two events be disjoint? Why or why not?

## Concepts Review: Exercise 1 - solution

#### Let:

A = Student attended the AI Club meeting B = Student attended the Math Workshop From the problem:

- P(A) = 0.40
- P(B) = 0.50
- $P(A \land B) = 0.20$
- 1. Probability of either event:  $P(A \lor B) = P(A) + P(B) P(A \land B) = 0.40 + 0.50 0.20 = 0.70$
- 2. Probability of neither event:  $P(\text{neither}) = 1 - P(A \lor B) = 1 - 0.70 = 0.30$
- 3. Are A and B Disjoint?:  $P(A \land B) = 0.20 \neq 0$

## **Conditional Probability**

$$P(X = x_i | Y = y_i)$$

"What is my belief that  $X=x_i$  if I already know  $Y=y_j$ "

Sometimes, knowing Y gives you information about X, i.e., changes your belief in X. In this case X and Y are said to be dependent.

$$P(X = x_i | Y = y_i) \neq P(X = x_i)$$

# **Axioms of Conditional Probability**

Which of the following axioms hold for conditional probabilities?

- A.  $P(X = x_i \mid Y = y_j) \ge 0$
- B.  $\sum_{i} P(X = x_i | Y = y_{\square}) = 1$
- C.  $\sum \Box P(X = x_i \mid Y = y \Box) = 1$
- D. A and B only
- E. A, B and C

# Joint Probability

$$P(X = x_i, Y = y_j)$$
"What is my belief that  $X = x_i$  and that  $Y = y_j$ "

#### Which of the following is always true?

A. 
$$P(X = x_i \text{ or } Y = y_j) \leq P(X = x_i, Y = y_j)$$

B. 
$$P(X = x_i \text{ or } Y = y_j) \ge P(X = x_i, Y = y_j)$$

C. 
$$P(X = x_i \text{ or } Y = y_j) = P(X = x_i, Y = y_j)$$

D. None of the above.

# Marginal Independence

$$P(X = x_i | Y = y_i) = P(X = x_i)$$

Sometimes knowing Y does not change your belief in X. In this case, X and Y are said to be independent.

$$P(W = w_i | Y = y_j) = P(W = w_i)$$

Where W = weather today

For which variable Y is the above statement most likely true?

- A. Y = The weather yesterday
- B. Y =The day of the week
- C. Y = The temperature

## More independence

Consider two students Roberto and Sabrina, who both took the same test. Define the following random variables:

R = Roberto aced the test

S = Sabrina aced the test

What is the most logical relationship between P(R = 1) and  $P(R = 1 \mid S = 1)$ ?

A. 
$$P(R = 1) = P(R = 1|S = 1)$$

B. 
$$P(R = 1)$$
 >  $P(R = 1|S = 1)$ 

C. 
$$P(R = 1)$$
 <  $P(R = 1|S = 1)$ 

# Conditional Independence

What if you also know the test was easy (variable T)?

A. 
$$P(R = 1|T = 1) = P(R = 1|T = 1, S = 1)$$

B. 
$$P(R = 1|T = 1) > P(R = 1|T = 1, S = 1)$$

C. 
$$P(R = 1|T = 1) < P(R = 1|T = 1, S = 1)$$

R and S are conditionally independent given T. I.e., if you already know T, knowing S does not give you additional information about R.

# More independence

Consider two events:

B = A burglar breaks into your apartment

E = An earthquake occurs

Are these events independent or dependent? (i.e., does knowing that one happened change your belief in the other?)

- A. They are independent because knowing that one happened does not change your belief that the other happened.
- B. They are dependent, because knowing that one happened changes your belief that the other happened.

# **Conditional Dependence**

$$P(B = 1) = P(B = 1|E = 1) = P(B = 1|E = 0)$$

Now consider a third event:

A = Your alarm goes off

Which of the following relationships best models beliefs about the world?

A. 
$$P(B = 1|A = 1) = P(B = 1|A = 1, E = 1)$$

B. 
$$P(B = 1|A = 1) > P(B = 1|A = 1, E = 1)$$

C. 
$$P(B = 1|A = 1) < P(B = 1|A = 1, E = 1)$$

#### Exercise 2

R = 1 if remote worker else 0

M = 1 if attend meeting else 0

W = 1 if submit report else 0

#### In a company:

- 70% of employees are remote and 30% work in the office
- Among remote workers:
  - 40% attend daily meetings, 50% submit weekly reports and 20% both attend meetings and submit reports
- Among office workers:
  - 80% attend daily meetings, 50% submit weekly reports and 40% both attend meetings and submit reports

Are the events M (attend meetings) and W (submit reports) conditionally independent given location (R = 1 or R = 0)?

#### Exercise 2 - Solution

**Note:** When two events are conditionally independent, their joint probability equals the product of their individual conditional probabilities.

- P(M=1 | R=1)=0.40
- P(W=1 | R=1)=0.50
- P(M=1,W=1|R=1)=0.20
- 0.40\*0.50 = 0.20 equality holds

#### For Office Workers:

- P(M=1 | R=0)=0.80
- P(W=1 | R=0)=0.50
- P(M=1,W=1 | R=0)=0.40
- 0.80 0.50=0.40 equality holds

**Answer:** Since the product of the individual probabilities equals the joint probability in both groups, M and W are conditionally independent given location (R).

# **Bayes Theorem**

## Bayes Theorem: Mathematical Definition

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

- P(A | B): Posterior probability (probability of A given B)
- P(B | A): Likelihood (probability of B given A)
- P(A): Prior probability of A
- P(B): Marginal probability of B (normalizing constant)

## The Big Idea: Reverse Reasoning

Bayes' Theorem helps us reason backward — from observed evidence back to the likely cause.

Consider the following scenario:

- Only 1% of applicants get a job offer from Google.
- Suppose 90% of Google hires had strong GitHub portfolios.
- But 10% of all applicants also have strong GitHub portfolios.

Now imagine a friend tells you someone has a strong GitHub portfolio.

What's the probability they actually got the job at Google?

Answer: 9%

## Bayes' Rule exercise 2

A company has three factories: A, B, and C.

- Factory A produces 30% of all products, with 2% defect rate
- Factory B produces 50% of all products, with 1% defect rate
- Factory C produces 20% of all products, with 3% defect rate

A product is selected at random and found to be defective.

What is the probability it came from Factory C?

#### Exercise 2 - Solution

#### P(C|D) = ?

• We can Use bayes theorem:

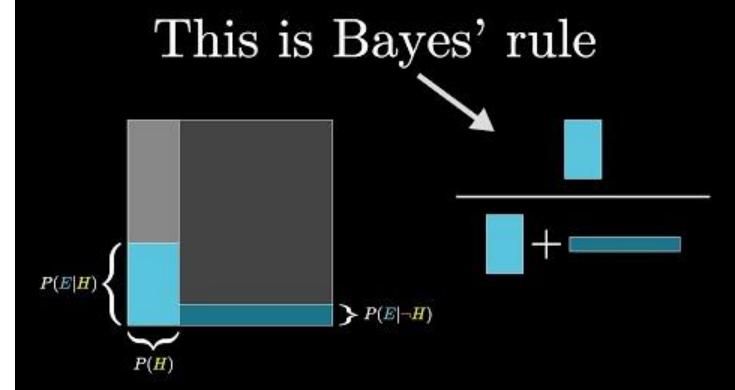
$$P(C \mid D) = P(D \mid C) * P(C)$$

$$P(D)$$

- P(C) = 20%
- $P(D \mid C) = 3\%$
- P(D) = total probability that a product is defective (we will compute this) = ?
   P(D)= P(D|A) P(A) + P(D|B) P(B) + P(D|C) P(C)
   =0.02\*0.30+0.01\*0.50+0.03\*0.20=0.006+0.005+0.006=0.017
- $P(C|D) = (0.03 * 0.2) / (0.017) \approx 0.353$

**Answer: 0.353** 

## Bayes Rule



#### Exercise 3

#### In a security system:

- 2% of all packages are dangerous (D)
- A scanner raises an alert (A) with:
  - 95% chance if the package is dangerous
  - 5% chance if the package is safe

A secondary check (S) is performed, which depends on whether the alert was triggered. The secondary check flags a package:

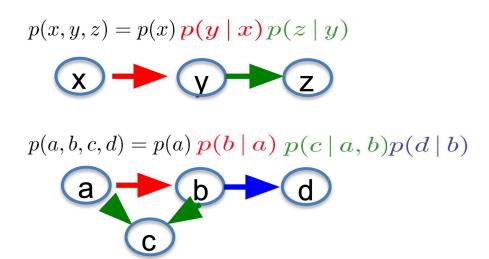
- 80% of the time if the alert was triggered (P(S | A=1) = 0.80)
- 10% of the time if no alert was triggered (P(S | A=0) = 0.10)

#### A package was flagged by the secondary check.

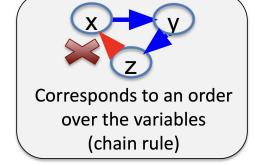
Q1: What is the probability the package is actually dangerous (P(D | S))? Answer: 0.104
Q2: Are events D and S conditionally dependent given A? Answer: Yes

## Introduction to Bayesian Networks

- Directed graphical model
- Nodes associated with variables
- "Draw" independence in conditional probability expansion
  - Parents in graph are the RHS of conditional

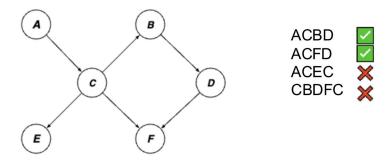


#### Graph must be acyclic



#### Paths in DAGs

- Definition
   A path is any sequence of nodes connected by edges (regardless of their directionalities); it is also assumed that no nodes repeat.
- Examples



## D-separation in DAGs

Motivation

How is conditional independence in a BN encoded by the structure of its DAG?

Theorem

$$P(X,Y \mid E) = P(X \mid E) P(Y \mid E)$$

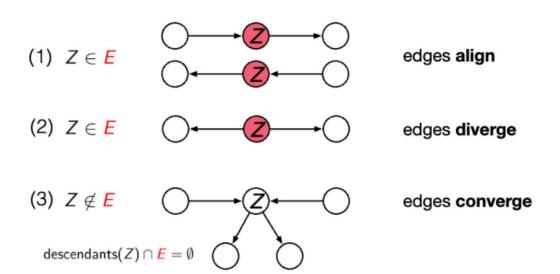
if and only if every path from a node in X to a node in Y is blocked by E.

What counts as a path, and when is it blocked?

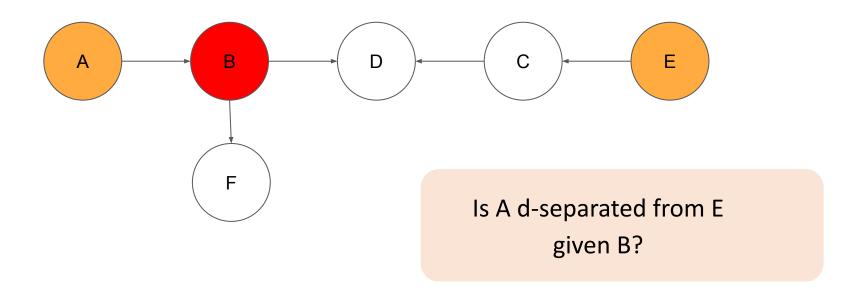
#### **Blocked Paths**

#### Definition

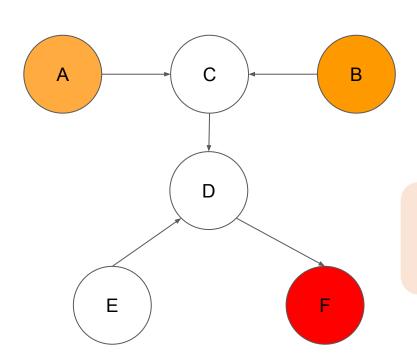
A path  $\pi$  is blocked by a set of nodes E if there exists a node  $Z \in \pi$  for which one of the three following conditions hold.



## D-separation: Exercise 1



## D-separation: Exercise 2



Is A d-separated from B given F?

#### Colab Notebooks Time!

Intro to NetworkX

Probabilities notebook

Thank you!