



CS5824: Advanced Machine Learning

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Please keep your face covering on!



COMPUTER SCIENCE
VIRGINIA TECH.

Basics on Probability

Coin Flips

- You flip a coin with one side labeled head (H) and the other labeled tail (T)
 - **Experiment**: the action of tossing coins
 - **Outcome space**: {H, T}
 - **Event**: any subset of the outcome space, e.g., {H}
 - **P(Event)**: the probability of an event occurring
- You flip 100 coins
 - Head with probability 0.5
 - How many heads would you expect?

Coin Flips cont.

- You flip a coin
 - Head with probability p
 - Binary random variable
 - Bernoulli trial with success probability p
- You flip k coins
 - How many heads would you expect
 - Number of heads X : discrete random variable
 - Binomial distribution with parameters k and p

Discrete Random Variables

- Random variables (RVs) which may take on only a **countable** number of **distinct** values
 - E.g., the total number of heads X you get if you flip 100 coins
- X is a RV with arity k if it can take on exactly one value out of $\{x_1, \dots, x_k\}$,
 - E.g., the possible values that X can take are 0, 1, 2,..., 100

Probability of Discrete RV

- Probability mass function (pmf): $P(X = x_i)$
- Easy facts about pmf
 - $P(X = x_i) \in [0,1]$
 - $\sum_i P(X = x_i) = 1$
 - $P(X = x_i \cap X = x_j) = 0$ if $i \neq j$
 - $P(X = x_i \cup X = x_j) = P(X = x_i) + P(X = x_j)$ if $i \neq j$

Common Distributions

- Uniform $X \sim U[1, \dots, N]$
 - X takes values $1, 2, \dots, N$
 - $P(X = i) = 1/N$
 - E.g., picking balls of different colors from a box
- Binomial $X \sim \text{Bin}(n, p)$
 - X takes values $0, 1, \dots, n$
 - $P(X = i) = \binom{n}{i} p^i (1-p)^{n-i}$
 - E.g., coin flips

Coin Flips of Two Persons

- Your friend and you both flip coins
 - Head with probability 0.5
 - You flip 50 times; your friend flip 100 times
 - How many heads will both of you get

Joint Distribution

- Given two discrete RVs X and Y , their **joint distribution** is the distribution of X and Y together

– E.g., $P(\text{You get 21 heads AND you friend get 70 heads})$

- $$\sum_x \sum_y P(X = x \cap Y = y) = 1$$

– E.g.,

$$\sum_{i=0}^{50} \sum_{j=0}^{100} P(\text{You get } i \text{ heads AND your friend get } j \text{ heads}) = 1$$

Conditional Probability

- $P(X = x | Y = y)$ is the probability of $X = x$, given the occurrence of $Y = y$
 - E.g., you get 0 heads, given that your friend gets 61 heads
- $$P(X = x | Y = y) = \frac{P(X = x \cap Y = y)}{P(Y = y)}$$

Law of Total Probability

- Given two discrete RVs X and Y , which take values in $\{x_1, \dots, x_m\}$ and $\{y_1, \dots, y_n\}$, we have

$$\begin{aligned} P(X = x_i) &= \sum_j P(X = x_i \cap Y = y_j) \\ &= \sum_j P(X = x_i | Y = y_j) P(Y = y_j) \end{aligned}$$

Marginalization

Marginal Probability

Joint Probability

$$P(X = x_i) = \sum_j P(X = x_i \cap Y = y_j)$$

$$= \sum_j P(X = x_i | Y = y_j) P(Y = y_j)$$

Conditional Probability

Marginal Probability

Bayes Rule

- X and Y are discrete RVs...

$$P(X = x | Y = y) = \frac{P(X = x \cap Y = y)}{P(Y = y)}$$



$$P(X = x_i | Y = y_j) = \frac{P(Y = y_j | X = x_i) P(X = x_i)}{\sum_k P(Y = y_j | X = x_k) P(X = x_k)}$$

Independent RVs

- Intuition: X and Y are independent means that $X = x$ **neither** makes it **more or less** probable that $Y = y$
- Definition: X and Y are independent iff
$$P(X = x \cap Y = y) = P(X = x)P(Y = y)$$

More on Independence

- $P(X = x \cap Y = y) = P(X = x)P(Y = y)$

A diagram consisting of a horizontal line with two arrows pointing downwards from its center. The left arrow points to the first equation, and the right arrow points to the second equation.

$$P(X = x|Y = y) = P(X = x) \quad P(Y = y|X = x) = P(Y = y)$$

- **E.g.**, no matter how many heads you get, your friend will not be affected, and vice versa

Conditionally Independent RVs

- Intuition: X and Y are conditionally independent given Z means that once Z is **known**, the value of X does not add any **additional** information about Y
- Definition: X and Y are conditionally independent given Z iff

$$P(X = x \cap Y = y | Z = z) = P(X = x | Z = z)P(Y = y | Z = z)$$

More on Conditional Independence

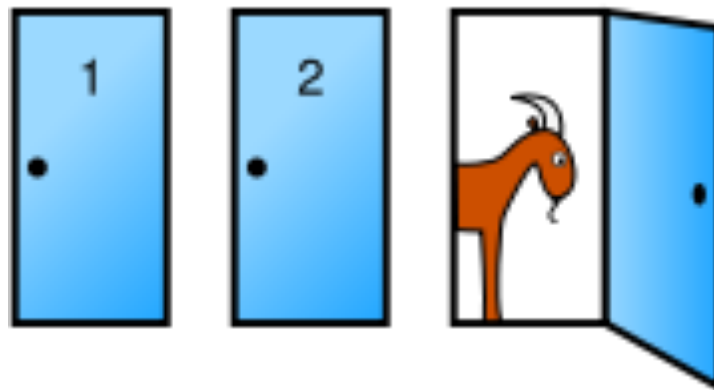
$$P(X = x \cap Y = y | Z = z) = P(X = x | Z = z) P(Y = y | Z = z)$$

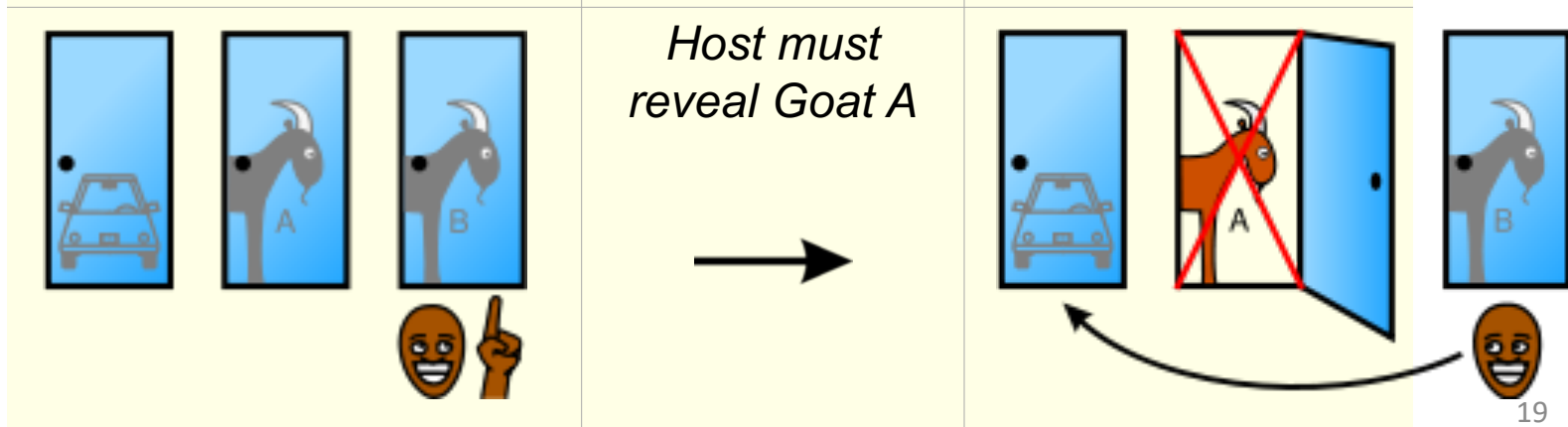
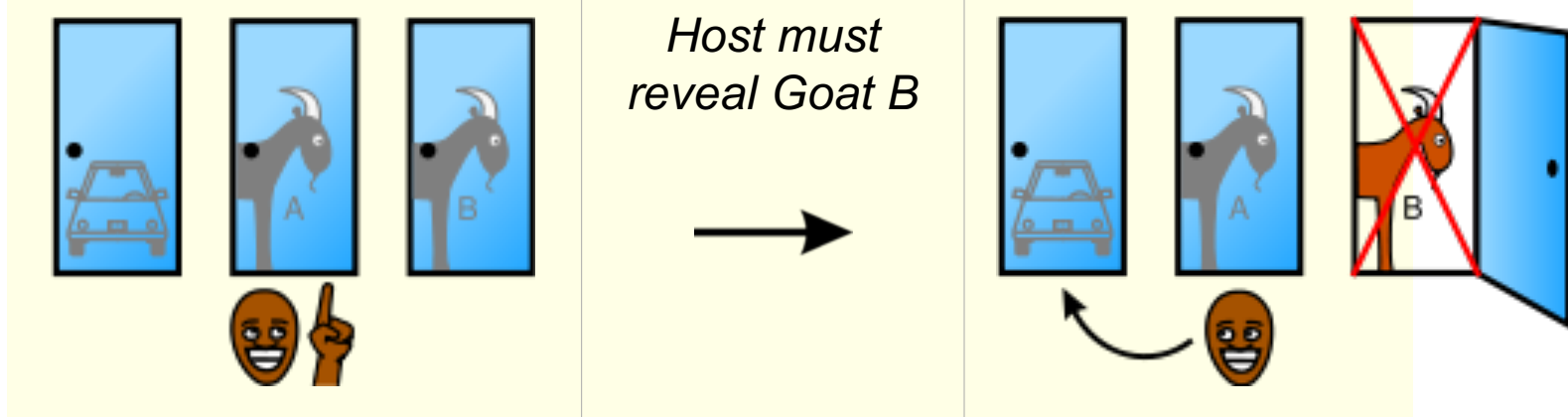
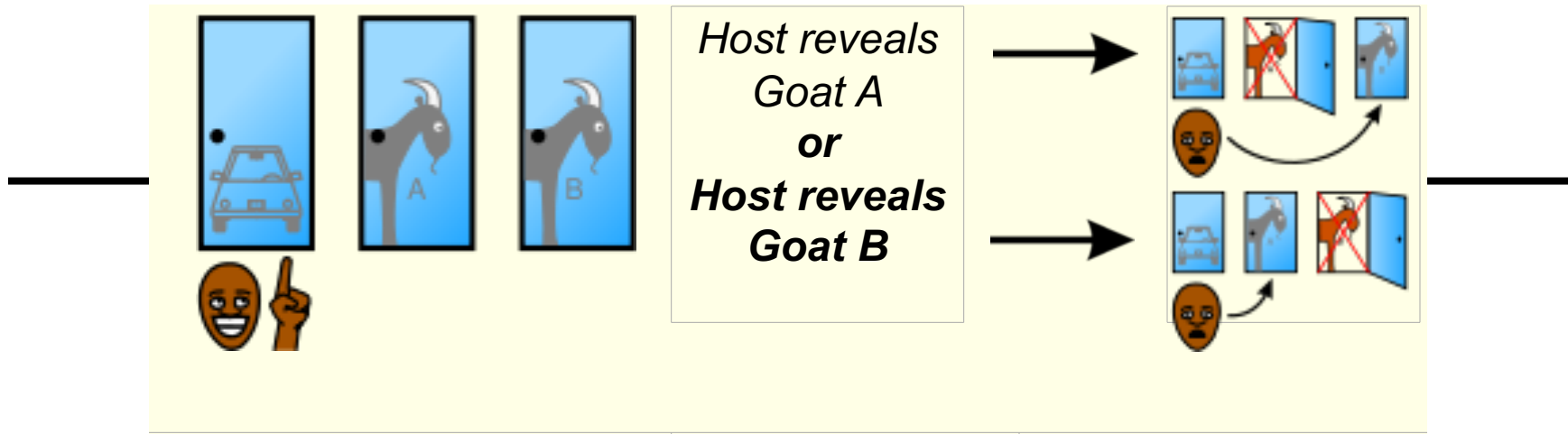
$$P(X = x | Y = y, Z = z) = P(X = x | Z = z)$$

$$P(Y = y | X = x, Z = z) = P(Y = y | Z = z)$$

Monty Hall Problem

- You're given the choice of three doors: Behind one door is a car; behind the others, goats.
- You pick a door, say No. 1
- The host, who knows what's behind the doors, opens another door, say No. 3, which has a goat.
- Do you want to pick door No. 2 instead?





Monty Hall Problem: Bayes Rule

- C_k : the car is behind door k , $k = 1, 2, 3$
- $P(C_k) = 1/3$
- H_{ij} : the host opens door j after you pick door i
- $$P(H_{ij} | C_k) = \begin{cases} 0 & i = j \\ 0 & j = k \\ 1/2 & i = k \\ 1 & i \neq k, j \neq k \end{cases}$$

Monty Hall Problem: Bayes Rule cont.

- WLOG, $i=1$ (your choice), $j=3$ (the host's choice)

- $$P(C_1 | H_{13}) = \frac{P(H_{13} | C_1) P(C_1)}{P(H_{13})}$$

- $$P(H_{13} | C_1) P(C_1) = \frac{1}{2} \cdot \frac{1}{3} = \frac{1}{6}$$

Monty Hall Problem: Bayes Rule cont.

- $$\begin{aligned} P(H_{13}) &= P(H_{13}, C_1) + P(H_{13}, C_2) + P(H_{13}, C_3) \\ &= P(H_{13} | C_1) P(C_1) + P(H_{13} | C_2) P(C_2) \\ &= \frac{1}{6} + 1 \cdot \frac{1}{3} \\ &= \frac{1}{2} \end{aligned}$$
- $$P(C_1 | H_{13}) = \frac{1/6}{1/2} = \frac{1}{3}$$

Monty Hall Problem: Bayes Rule cont.

- $P(C_1 | H_{13}) = \frac{1/6}{1/2} = \frac{1}{3}$
- $P(C_2 | H_{13}) = 1 - \frac{1}{3} = \frac{2}{3} > P(C_1 | H_{13})$
- *You should switch!*