

Review:

Tuesday, February 9, 2021 4:00 PM

$$P(A|B) = \frac{P(A \cap B)}{P(B)} \quad \text{given } P[B] > 0$$

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

# Independent events

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Def. 2 events,  $A, B$  are independent iff

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

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Immediate consequence:

if  $P[B] > 0$ ,  $A, B$  are indep, we can write

$$P[A|B] = P[A \cap B] / P[B]$$

$$(indep) = P[A]P[B] / \cancel{P[B]}$$

$$\rightarrow P[A|B] = P[A]$$

Knowing  $B$  occurred gives us no info about  $A$

## Theorems

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THM. If  $P[A] = 0$ , then  $A$  is independent of any event  $B$

Proof  $P[A]P[B] = 0$

and  $P[A \cap B] = 0$  so  $A, B$  are indep.

THM If  $A, B$  are indep. then so too are

a)  $A$  and  $B^c$

b)  $A^c$  and  $B$

c)  $A^c$  and  $B^c$

# Example

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I roll one red, one blue die

A: I roll a 5 on red

B: the sum of the dice is even

C: The sum of the dice is 10

$$P[A] = 1/6$$

$$P[B] = 1/2$$

$$P[C] = 3/36$$

$$P[A \cap B] = P[A] P[B|A] = \frac{1}{6} \cdot \frac{1}{2} = \frac{1}{12} = P[A] P[B]$$

✓ A, B: indep.

$$P[A \cap C] = P[A] P[C|A] = \frac{1}{6} \cdot \frac{1}{6} = \frac{1}{36} \neq P[A] P[C]$$

so A, C are dependent

One more thing...

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Def.  $A, B, C$  are mutually indep. if

a)  $A, B$  indep.  $B, C$  indep.,  $A, C$  indep.

$$b) P[A \cap B \cap C] = P[A]P[B]P[C]$$

go to  $n$  events, with

- events mutually independent in arbitrary collections

ex: Yahtzee roll 5 indep. dice

note: independence is not mutual exclusivity

$$A, B \text{ m.e. then } P[A \cap B] = 0$$

even if  $P[A] \neq 0 \neq P[B]$ !

most cases ( $P[A] \neq 0, P[B] \neq 0$ )

m.e.  $\Rightarrow$  dependence

# Total Probability THM

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Theorem Suppose  $B_1, \dots, B_k$  are mutually exclusive and exhaustive (i.e.  $\bigcup_{i=1}^k B_i = S$ ) we call  $\{B_i\}_{i=1}^k$  a partition of  $S$ . For any such partition and any event

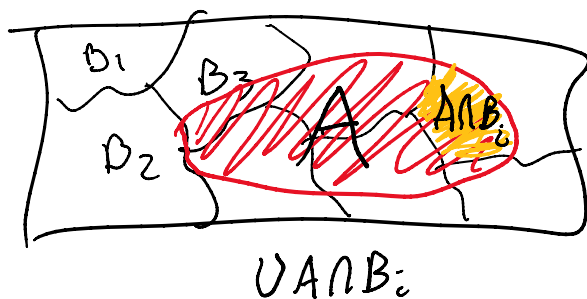
$$A \subset S, \quad P[A] = \sum_{i=1}^k P[B_i] P[A|B_i]$$

Proof.  $P[B_i] P[A|B_i] = P[A \cap B_i]$

$$\rightarrow \sum_i P[B_i] P[A|B_i] = \sum_i P[A \cap B_i]$$

$$\text{(m.e.)} = P\left[\bigcup_{i=1}^k A \cap B_i\right]$$

$$\text{(exhaustive)} = P[A] \quad \underline{\underline{\quad}}$$



# Bayes' Theorem

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$$P[B|A] = \frac{P[A|B] P[B]}{P[A]}$$

Can relate one conditional direction  
to the other

i.e. if I know  $P[A|B]$  and the individual probs,  
I can find  $P[B|A]$

more often used formulation:

$\{B_i\}_{i=1}^k$  is a partition of  $S$  then

$$P[B_i|A] = \frac{P[B_i] P[A|B_i]}{\sum_{j=1}^k P[B_j] P[A|B_j]}$$

↑ TPT to rewrite  $P[A]$

Proof.  $P[B|A] = \frac{P[A \cap B]}{P[A]}$ ,  $P[A|B] = \frac{P[A \cap B]}{P[B]}$

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

↓  
Distributing  $P[B] P[A|B]$

$$\rightarrow P[B|A] = \frac{P[B] P[A|B]}{P[A]} //$$



## Example

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In NY  $\sim 1\%$  of people have COVID-19

One common test has a false positive rate of 5%

and a false negative rate of 2%

Q: given I test positive, what is the probability that I have COVID-19?

Event  $V$ : I have the virus

$T^+$ : I test positive

$T^-$ : I test negative

Want:  $P[V|T^+]$

false negative:  ~~$P[V|T^-]$~~   $P[T^-|V] = 0.02$

false positive:  $P[T^+|V^c]$

$$P[V|T^+] = \frac{P[T^+|V]P[V]}{P[T^+]}$$

$$P[V] = 0.01 \quad (\text{given})$$

$$P[T^+|V] = 1 - P[T^-|V] \quad (\text{as } T^+ \text{ and } T^- \text{ are exh.})$$
$$= 0.98$$

$$P[T^+] = ? \quad \underline{TPT}$$

$$P[T^+] = P[V] P[T^+|V] + P[V^c] P[T^+|V^c]$$

0.99                      false +  
↓                              0.05

↑                              ↑  
0.01                              0.98

$$= 0.0098 + 0.0495$$

$$P[V|T^+] = \frac{P[T^+|V] P[V]}{P[V] P[T^+|V] + P[V^c] P[T^+|V^c]}$$
$$= \frac{(0.98)(0.01)}{(0.98)(0.01) + (0.99)(0.05)}$$

$$\approx 16.5\%$$

Why!?! only 1% of people have it!

# Terminology

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$$P[B_k|A] = \frac{P[A|B_k]P[B_k]}{\sum_i P[A|B_i]P[B_i]}$$

Call  $P[B_i]$  the "priors"

Call  $P[B_k|A]$  the "posterior"

## Discrete Distributions

$\mathcal{S}$  - sample space. A set of items that may not be numbers ex.  $\{\text{heads, tails}\}$ ,  $\mathbb{C}$

Random Variables:  $X: \mathcal{S} \rightarrow \mathbb{R}$

$$\text{range}(X) = \{x \mid x = X(s) \text{ for some } s \in \mathcal{S}\}$$

call this "the spec of R.V.  $X$ "

usually denote as  $S \subset \mathbb{R}$  easier to manipulate than  $\mathcal{S}$

For any experiment you can choose  $\infty$  random variables

$$\begin{array}{l} X(\text{heads}) = 1 \\ X(\text{tails}) = 0 \end{array} \quad , \quad \begin{array}{l} Y(\text{heads}) = -53.1 \\ Y(\text{tails}) = \frac{e}{12\pi^2} \end{array} \quad \leftarrow \begin{array}{l} \text{less natural, but} \\ \text{valid!} \end{array}$$

If  $S \subset \mathbb{R}$ , a natural R.V. is the identity, so  $S = \mathcal{S}$  and  $X(s) = s$

Ex. Die roll  $S = \{1, 2, 3, 4, 5, 6\}$

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$X(s) = s, S = \mathcal{S}$

# Probability

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In chapter 1, we covered probabilities on  $\mathcal{S}$  we are interested in working in  $S$ . Map probability to this new space.

$$A \subset S, \text{ we define } P[A] \equiv P[X^{-1}(A)]$$

$$X^{-1}(A) = \{s \in \mathcal{S} \mid X(s) \in A\} \subset \mathcal{S}$$

Ex. I draw a ball from a pot labeled a, b or c

$$X(a) = 0, X(b) = 1, X(c) = 2$$

$$\mathcal{S} = \{a, b, c\}, S = \{0, 1, 2\}$$

prob. that I draw an a or b can be written in

$$S \text{ as } P[\{0, 1\}] = P[\{a, b\}]$$

$$X^{-1}(\{0, 1\}) = \{a, b\} \subset \mathcal{S}$$

Ex.

$$\text{Roll a die, } X(1) = X(3) = X(5) = 0$$

$$X(2) = X(4) = X(6) = 1$$

$$\begin{aligned}
 P[X=1] &= P[\underbrace{\{1\}}_S] = P[X^{-1}(1)] \\
 &= P[\underbrace{\{2, 4, 6\}}_S] \\
 &= 1/2
 \end{aligned}$$

Ex I roll 2 indep. dice, sample space

$$S = \{(1,1), (1,2), (1,3), \dots, (6,6)\} \subset \mathbb{R}^2$$

$$P[s \in S] = 1/36 \text{ for any singleton } s$$

$X((a,b)) = a+b \in \mathbb{R}$   $X$  maps tuple to the sum of dice

$$S = \{2, 3, \dots, 12\}$$

$$\begin{aligned}
 P[X=5] &= P[X^{-1}(5)] = P[\underbrace{\{(1,4), (4,1), (2,3), (3,2)\}}_S] \\
 &= 4/36
 \end{aligned}$$

$$X^{-1}(5) = \{s \in S \mid X(s) = 5\} = \{(1,4), (4,1), (2,3), (3,2)\}$$

# PMF

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Def. The probability mass function (PMF) of a discrete r.v.  $X$  is a function  $f: S \rightarrow \mathbb{R}$  s.t.

$$a) f(x) \geq 0 \quad \forall x \in S$$

$$b) \sum_{x \in S} f(x) = 1$$

$$c) P[X \in A] = \sum_{x \in A} f(x)$$

we usually say  $f(x) = 0$  for  $x \notin S$  and call  $S$  the "Support" of  $f$ . In this case,  $f: \mathbb{R} \rightarrow \mathbb{R}$  technically



# Uniform dist.

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$$|S| = N$$

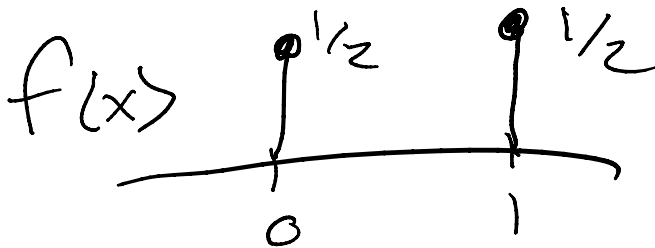
↑ # elements

ex. flipping coin  $N=2$   
rolling a die  $N=6$

$$f(x) = 1/N \quad \forall x \in S$$

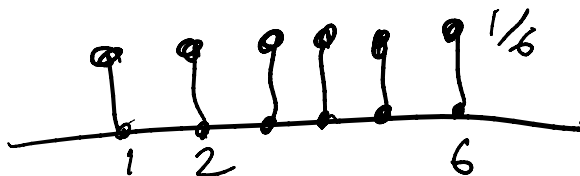
say  $X(\text{heads})=0, X(\text{tails})=1$

$$f(0) = f(1) = 1/2$$



say  $Y(1)=1, Y(2)=2, \dots, Y(6)=6$

pmf.  $g(y) = 1/6, y=1, 2, \dots, 6$



these are examples of uniform distributions

# EX: Not uniform

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$X$  is sum of 2 dice (from before)

$$S = \{2, 3, \dots, 12\}$$

$$P[X=2] = P[\{(1,1)\}] = 1/36 = P[X=12]$$

$$P[X=3] = P[\{(1,2), (2,1)\}] = 2/36 = P[X=11]$$

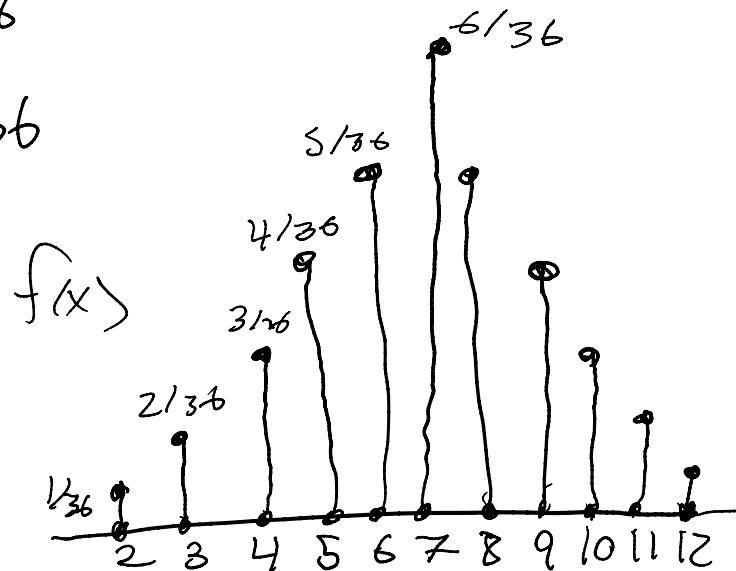
$$P[X=4] = P[X=10] = 3/36$$

$$P[X=5] = P[X=9] = 4/36$$

$$P[X=6] = P[X=8] = 5/36$$

$$P[X=7] = 6/36$$

$$f(x) = \begin{cases} 1/36, & x=2, 12 \\ 2/36, & x=3, 11 \\ 3/36, & x=4, 10 \\ 4/36, & x=5, 9 \\ 5/36, & x=6, 8 \\ 6/36, & x=7 \end{cases}$$



$$f(x) = \frac{6 - |7 - x|}{36}, \quad x=2, \dots, 12$$

# Exam 1

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in 2 weeks - 2/23

- I will post it at 4 PM
- we won't have class
- Due at 4 PM on 2/24

Covers The material from Lectures 1-3  
- all of chapter 1  
- chapter 2.1

Focus - methods of enumeration, basic proofs involving laws of probability

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