# PROBABILITY 2.1 INTRO TO DISCRETE PROBABILITY

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#### AGENDA

- DEFINITIONS
- AXIOMS
- EQUALLY LIKELY OUTCOMES

#### DEFINITIONS

**Sample Space:** The set  $\Omega$  of all possible outcomes of an experiment.

- Single coin flip:  $\Omega = \{H, T\}$
- Two coin flips:  $\Omega = \{HH, HT, TH, TT\}$
- Roll of a die:  $\Omega = \{1,2,3,4,5,6\}$

#### **Event:** Any subset $E \subseteq \Omega$ .

- Getting at least one head in two coin flips:  $E = \{HH, HT, TH\}$
- Rolling an even number:  $E = \{2,4,6\}$

<u>Mutually Exclusive</u>: Events E and F are mutually exclusive if  $E \cap F = \emptyset$  (i.e., they can't simultaneously happen).

•  $E = \{2,4,6\}$  and  $F = \{1,3\}$ , then  $E \cap F = \emptyset$ .





# EXAMPLE: WEIRD DICE (SAMPLE SPACE)



SUPPOSE I ROLL TWO 4-SIDED DICE. WHAT'S THE SAMPLE SPACE (SET OF POSSIBLE OUTCOMES)?

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SUPPOSE I ROLL TWO 4-SIDED DICE. WHAT'S THE SAMPLE SPACE (SET OF

POSSIBLE OUTCOMES)?



DIE 2 (RED)

		1	L	J	Ť
	1	(1, 1)	(1, 2)	(1, 3)	(1, 4)
4 50	2	(2, 1)	(2, 2)	(2, 3)	(2, 4)
DIE 1 (BLUE)	3	(3, 1)	(3, 2)	(3, 3)	(3, 4)
	4	(4, 1)	(4, 2)	(4, 3)	(4, 4)



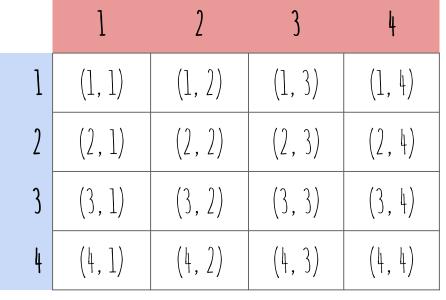
LET D1 BE THE VALUE OF THE BLUE DIE, AND D2 THE VALUE OF THE RED DIE.

WHAT OUTCOMES MATCH THESE EVENTS?



DIE 2 (RED)

A.	DI	=	







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📣 DIE 2 (RED)

A. | | | | | = |

]



	1	2	3	4
l	$(1, 1)^{A}$	$(1, 2)^{A}$	$(1,3)^{A}$	(1, 4) A
2	(2, 1)	(2, 2)	(2, 3)	(2, 4)
3	(3, 1)	(3, 2)	(3, 3)	(3, 4)
4	(4, 1)	(4, 2)	(4, 3)	(4, 4)



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WHAT OUTCOMES MATCH THESE EVENTS?



A DIE 2 (RED)

A. D] = ]

 $\beta. D1 + D2 = 6$ 



		6 <sup>1</sup> q
DIE	1	(BLUE

	1	2	3	4
1	$(1, 1)^{A}$	$(1, 2)^{A}$	$(1,3)^{A}$	(1, 4) A
2	(2, 1)	(2, 2)	(2, 3)	(2, 4)
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📣 DIE 2 (RED)

 $A. \quad D] = 1$ 

 $\beta. \quad D1 + D2 = 6$ 



	1	2	3	4
1	$(1, 1)^{A}$	$(1, 2)^{A}$	$(1,3)^{A}$	(1, 4) 4
2	(2, 1)	(2, 2)	(2, 3)	(2, 4) <sup>B</sup>
3	(3, 1)	(3, 2)	(3, 3) B	(3, 4)
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WHAT OUTCOMES MATCH THESE EVENTS?



A DIE 2 (RED)

 $\begin{bmatrix} & & & \\ & & & \\ & & & \end{bmatrix} = 2 \times 2$ 



	1	2	3	4
l	$(1, 1)^{A}$	$(1, 2)^{A}$	$(1,3)^{A}$	(1, 4) <sup>A</sup>
2	(2, 1)	(2, 2)	(2, 3)	(2, 4) B
3	(3, 1)	(3, 2)	(3, 3) B	(3, 4)
4	(4, 1)	(4, 2) B	(4, 3)	(4, 4)



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WHAT OUTCOMES MATCH THESE EVENTS?



📣 DIE 2 (RED)

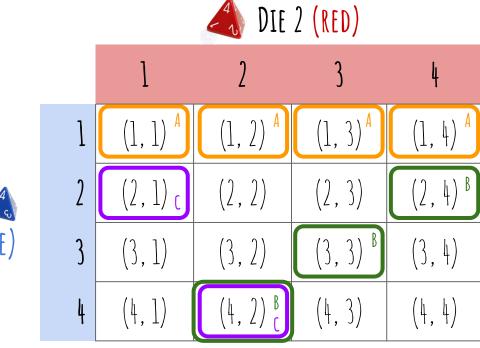


DIE 1 (BLUE)

	1	2	3	4
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3	(3, 1)	(3, 2)	(3, 3) B	(3, 4)
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ARE A AND B MUTUALLY EXCLUSIVE?

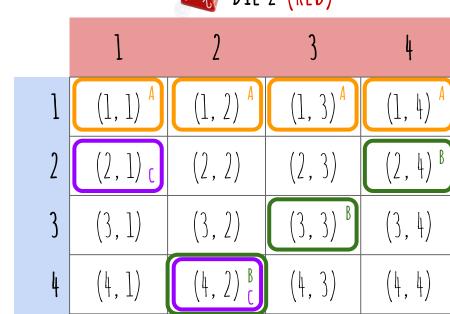






ARE A AND B MUTUALLY EXCLUSIVE?

YES.  $A \cap B = \emptyset$  (NO OVERLAP)



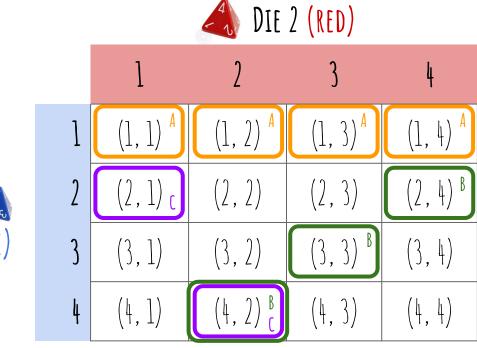








ARE B AND C MUTUALLY EXCLUSIVE?



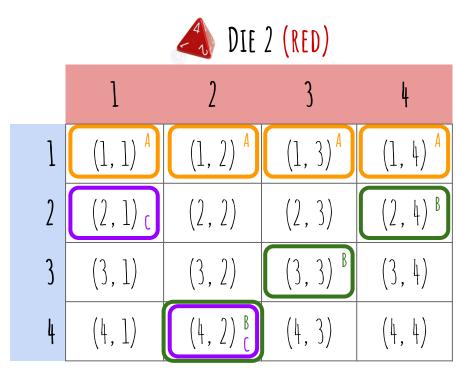




ARE B AND C MUTUALLY EXCLUSIVE?

NO. B AND C COULD HAPPEN AT THE SAME TIME (4, 2)





# RANDOM PICTURE



Let  $\Omega$  denote the sample space and  $E, F \subseteq \Omega$  be events.

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Axiom 1 (Nonnegativity):  $P(E) \ge 0$ .  $\leftarrow NO$  EVENT HAS NEGATIVE PROBABILITY.

Let  $\Omega$  denote the sample space and  $E, F \subseteq \Omega$  be events.

Axiom 1 (Nonnegativity):  $P(E) \ge 0$ .

Axiom 2 (Normalization):  $P(\Omega) = 1$ .  $\leftarrow 50$ METHING HAS TO HAPPEN (100%).

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Axiom 2 (Normalization):  $P(\Omega) = 1$ .

Axiom 3 (Countable Additivity) If E and F are mutually exclusive, then

 $P(E \cup F) = P(E) + P(F).$ 

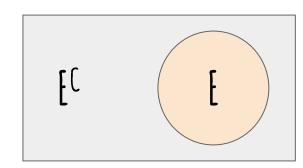
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Corollary 1 (Complementation):  $P(E^C) = 1 - P(E)$ .



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Corollary 1 (Complementation):  $P(E^{C}) = 1 - P(E)$ .

Corollary 2 (Monotonicity): If  $E \subseteq F$ ,  $P(E) \le P(F)$ .

Corollary 3 (Inclusion-Exclusion):  $P(E \cup F) = P(E) + P(F) - P(E \cap F)$ .



THINK BACK TO THE 4-SIDED DICE. SUPPOSE EACH DIE IS FAIR.

INTUITIVELY, WHAT IS THE PROBABILITY THAT THE ADDIE 2 (RED)



TWO DICE SUM TO 6?	(D] +	D2 = {	)
--------------------	-------	--------	---

	1	2	3	4
1	(1, 1)	(1, 2)	(1, 3)	(1, 4)
2	(2, 1)	(2, 2)	(2, 3)	(2, 4)
3	(3, 1)	(3, 2)	(3, 3)	(3, 4)
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INTUITIVELY, WHAT IS THE PROBABILITY THAT THE ADDIE 2 (RED)

TWO DICE SUM TO	6) (D]	+ D2	= (0)
-----------------	--------	------	-------

2	3	4
(1, 2)	(1, 3)	(1, 4)
(2, 2)	(2, 3)	(2, 4) B

EACH OF THE 16 OUTCOMES	IS
EQUALLY LIKELY.	DIE 1 (BLUE)
3/16.	

#### EQUALLY LIKELY OUTCOMES

If  $\Omega$  is such that outcomes are equally likely, then for any event  $E \subseteq \Omega$ ,

$$P(E) = \frac{|E|}{|\Omega|}$$



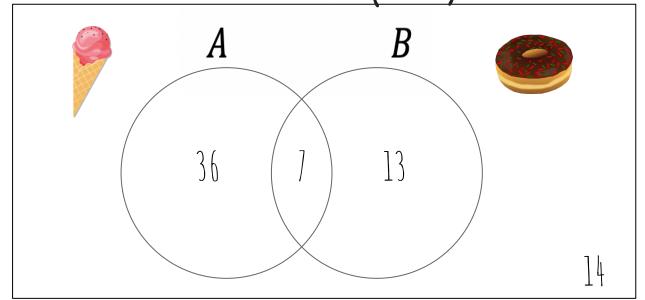
# PROBABILITY 2.2 CONDITIONAL PROBABILITY

ALEX TSUN

#### AGENDA

- CONDITIONAL PROBABILITY
- BAYES THEOREM
- LAW OF TOTAL PROBABILITY (LTP)
- BAYES THEOREM + LTP

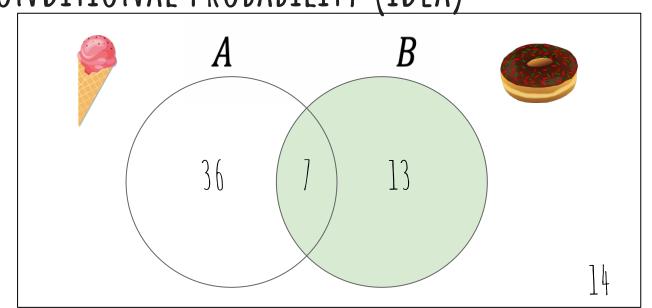
CONDITIONAL PROBABILITY (IDEA)





WHAT'S THE PROBABILITY THAT SOMEONE LIKES ICE CREAM GIVEN THEY LIKE DONUTS?

CONDITIONAL PROBABILITY (IDEA)





WHAT'S THE PROBABILITY THAT SOMEONE LIKES ICE CREAM GIVEN THEY LIKE DONUTS?

$$P(A|B) = \frac{7}{20} = \frac{|A \cap B|}{|B|} = \frac{|A \cap B|/|\Omega|}{|B|/|\Omega|} = \frac{P(A \cap B)}{P(B)}$$

#### CONDITIONAL PROBABILITY

<u>Conditional Probability</u>: The (conditional) probability of A given an event B happened is

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

An equivalent and useful formula is  $P(A \cap B) = P(A|B)P(B)$ .



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



Does P(A|B) = P(B|A)? No!!



Does 
$$P(A|B) = P(B|A)$$
? No!!

Let A be the event you are wet. Let B be the event you are swimming.



Does 
$$P(A|B) = P(B|A)$$
? No!!

Let A be the event you are wet.

Let B be the event you are swimming.

$$P(A|B)=1$$

# CONDITIONAL PROBABILITY (REVERSAL)



Does 
$$P(A|B) = P(B|A)$$
? No!!

Let A be the event you are wet.

Let  ${\it B}$  be the event you are swimming.

$$P(A|B)=1$$

$$P(B|A) \neq 1$$

#### BAYES THEOREM

Bayes Theorem: Let A, B be events with nonzero probability. Then,

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

Allows us to "reverse" the conditioning!

P(A) is called the <u>prior</u> (our belief without knowing anything), and P(A|B) is called the <u>posterior</u> (our belief after learning B).

By definition of conditional probability,

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



By definition of conditional probability,

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

Swapping A, B gives

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



By definition of conditional probability,

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

Swapping A, B gives

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

But 
$$P(A \cap B) = P(B \cap A)$$
, so

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



By definition of conditional probability,

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

Swapping A, B gives

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

But  $P(A \cap B) = P(B \cap A)$ , so

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

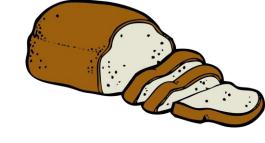
Dividing both sides by P(B) gives

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

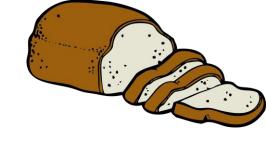


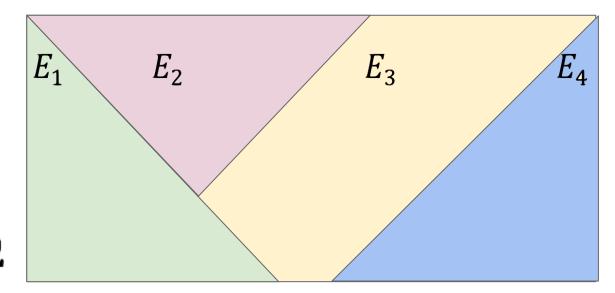
#### RANDOM PICTURE



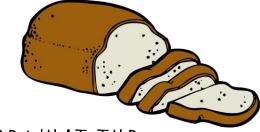




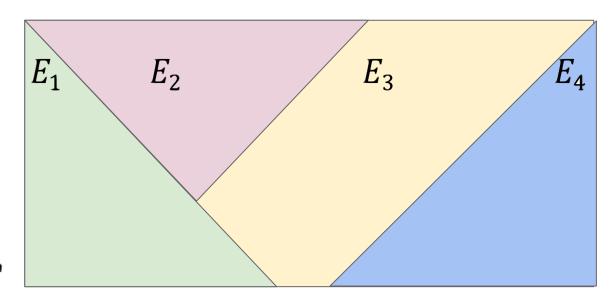




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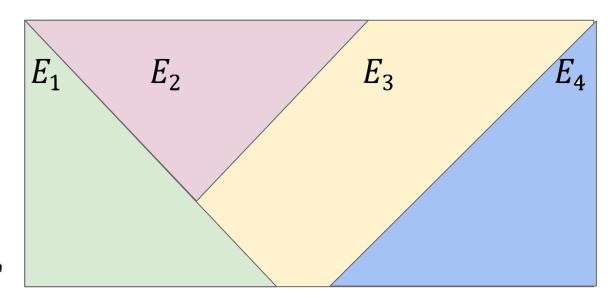


WHAT SINGLE ENGLISH WORD WOULD YOU USE TO DESCRIBE WHAT THE FOLLOWING EVENTS DO TO THE SAMPLE SPACE?



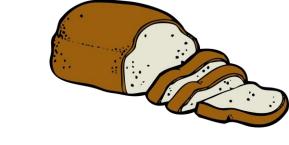


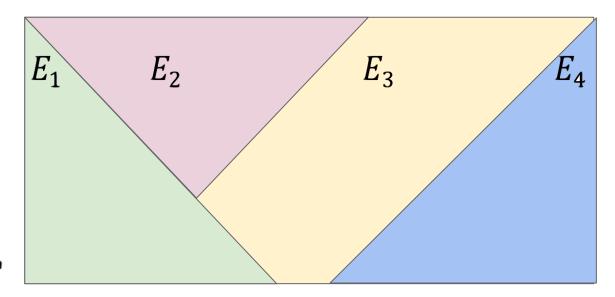
WHAT SINGLE ENGLISH WORD WOULD YOU USE TO DESCRIBE WHAT THE FOLLOWING EVENTS DO TO THE SAMPLE SPACE? **PARTITION** 





WHAT ARE TWO PROPERTIES OF THE "PARTITION"?

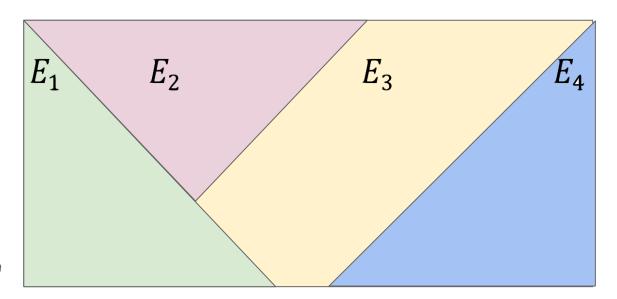


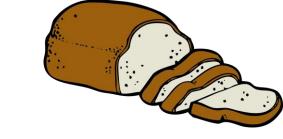




WHAT ARE TWO PROPERTIES OF THE "PARTITION"?

1. THEY "COVER" THE WHOLE SPACE.



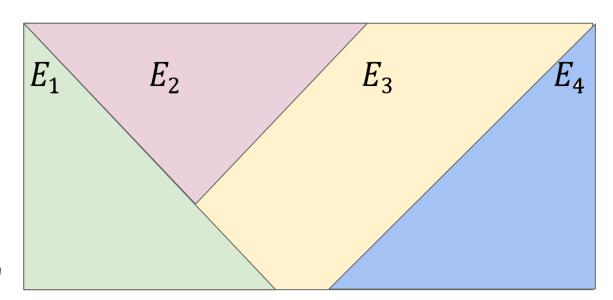




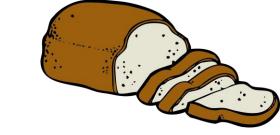
WHAT ARE TWO PROPERTIES OF THE "PARTITION"?

1. THEY "COVER" THE WHOLE SPACE.

2. THEY DON'T OVERLAP.





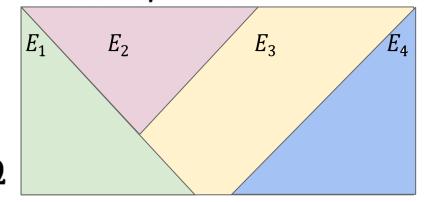


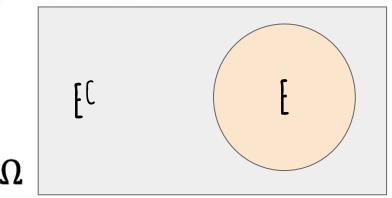
#### PARTITIONS

#### <u>Partition</u>: Non-empty events $E_1, \dots, E_n$ partition the sample space $\Omega$ if

- (Exhaustive)  $E_1 \cup E_2 \cup ... \cup E_n = \bigcup_{i=1}^n E_i = \Omega$ .
- (Pairwise Mutually Exclusive) For all  $i \neq j$ ,  $E_i \cap E_j = \emptyset$ .

Notice for any event  $E \colon E$  and  $E^C$  always partition  $\Omega$ .

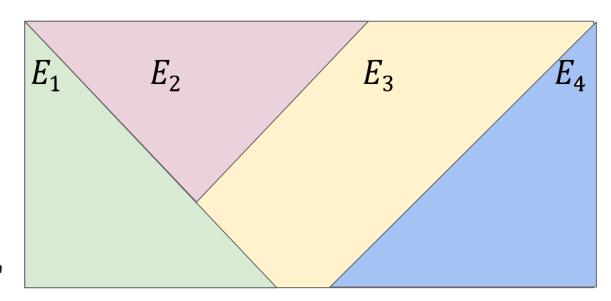




# (THE PICTURE) LAW OF TOTAL PROBABILITY

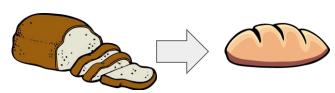


BACK TO THE OLD PICTURE.



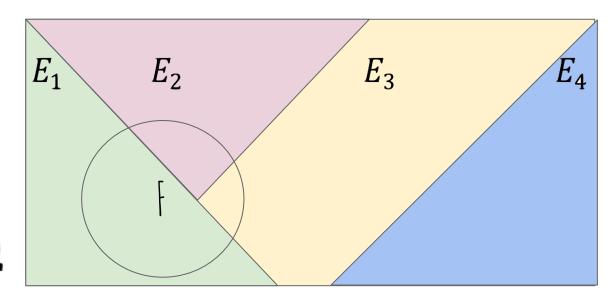


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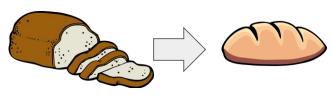
BACK TO THE OLD PICTURE. HOW CAN WE DECOMPOSE EVENT  $\mathbf{F}$ ?

$$P(F) =$$



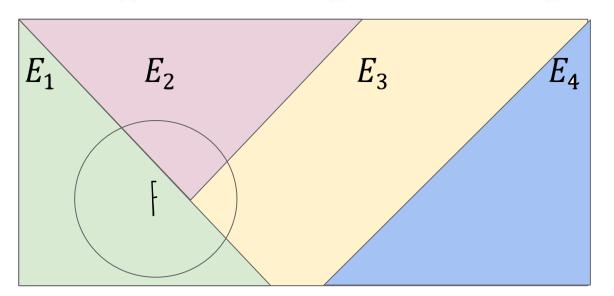


# (THE PICTURE) LAW OF TOTAL PROBABILITY



BACK TO THE OLD PICTURE. HOW CAN WE DECOMPOSE EVENT  $\mathbf{F}$ ?

$$P(F) = P(F \cap E_1) + P(F \cap E_2) + P(F \cap E_3) + P(F \cap E_4)$$





#### LAW OF TOTAL PROBABILITY (LTP)

**Law of Total Probability**: If events  $E_1, ..., E_n$  partition  $\Omega$ , then for any event F,

$$P(F) = P(F \cap E_1) + \dots + P(F \cap E_n) = \sum_{i=1}^{n} P(F \cap E_i)$$

Using the definition of conditional probability  $(P(F \cap E_i) = P(F|E_i)P(E_i))$ , we get an alternate (more useful) form

$$P(F) = P(F|E_1)P(E_1) + \dots + P(F|E_n)P(E_n) = \sum_{i=1}^{n} P(F|E_i)P(E_i)$$

$$P(F) = P(F|E_1)P(E_1) + \dots + P(F|E_n)P(E_n) = \sum_{i=1}^{n} P(F|E_i)P(E_i)$$



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IN CHEMISTRY CLASS, YOU WANT TO KNOW THE PROBABILITY YOU FAIL.

$$P(F) = P(F|E_1)P(E_1) + \dots + P(F|E_n)P(E_n) = \sum_{i=1}^{n} P(F|E_i)P(E_i)$$



- IN CHEMISTRY CLASS, YOU WANT TO KNOW THE PROBABILITY YOU FAIL.
- BUT YOU ARE RANDOMLY ASSIGNED ONE OF 3 TEACHERS. WHAT TO DO?

$$P(F) = P(F|E_1)P(E_1) + \dots + P(F|E_n)P(E_n) = \sum_{i=1}^{n} P(F|E_i)P(E_i)$$



- IN CHEMISTRY CLASS, YOU WANT TO KNOW THE PROBABILITY YOU FAIL.
- BUT YOU ARE RANDOMLY ASSIGNED ONE OF 3 TEACHERS. WHAT TO DO?
- FIRST, COMPUTE THE PROBABILITY OF FAILING IN EACH OF THE 3 CASES.

$$P(F) = P(F|E_1)P(E_1) + \dots + P(F|E_n)P(E_n) = \sum_{i=1}^{n} P(F|E_i)P(E_i)$$



- IN CHEMISTRY CLASS, YOU WANT TO KNOW THE PROBABILITY YOU FAIL.
- BUT YOU ARE RANDOMLY ASSIGNED ONE OF 3 TEACHERS. WHAT TO DO?
- FIRST, COMPUTE THE PROBABILITY OF FAILING IN EACH OF THE 3 CASES.
- THEN, WEIGHT THOSE BY THE PROBABILITY OF GETTING THAT TEACHER.

#### EXAMPLE (LTP)

$$P(F) = P(F|E_1)P(E_1) + \dots + P(F|E_n)P(E_n) = \sum_{i=1}^{n} P(F|E_i)P(E_i)$$

	Mrs. Mean $(E_1)$	Mr. Nice $(E_2)$	Ms. $IDC(E_3)$
Probability of	6/8	1/8	1/8
Teaching			
Probability of			
Failing You			

#### EXAMPLE (LTP)

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Probability of	6/8	1/8	1/8
Teaching			
Probability of	1	0	1/2
Failing You			

#### EXAMPLE (LTP)

$$P(F) = P(F|E_1)P(E_1) + \dots + P(F|E_n)P(E_n) = \sum_{i=1}^{n} P(F|E_i)P(E_i)$$

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Probability of	6/8	1/8	1/8
Teaching			
Probability of	1	0	1/2
Failing You			

HOW ARE YOU LIKING YOUR CHANCES???

# EXAMPLE (REVERSAL)



	Mrs. Mean $(E_1)$	Mr. Nice $(E_2)$	Ms. $IDC(E_3)$
Probability of	6/8	1/8	1/8
Teaching			
Probability of	1	0	1/2
Failing You			
$P(F) = P(F E_1)P(E_1) + P(F E_2)P(E_2) + P(F E_3)P(E_3) = 1 \cdot \frac{6}{8} + 0 \cdot \frac{1}{8} + \frac{1}{2} \cdot \frac{1}{8} = \frac{13}{16}$			

### EXAMPLE (REVERSAL)



		Ms. $IDC(E_3)$
Probability of		1/8
Teaching	$P(E_3 F) =$	
Probability of		1/2
Failing You		

 $P(F) = P(F|E_1)P(E_1) + P(F|E_2)P(E_2) + P(F|E_3)P(E_3) = 1 \cdot \frac{6}{8} + 0 \cdot \frac{1}{8} + \frac{1}{2} \cdot \frac{1}{8} = \frac{13}{16}$ 

WHAT'S THE PROBABILITY THAT YOU HAD MS. IDC, GIVEN THAT YOU FAILED?

#### BAYES THEOREM WITH LAW OF TOTAL PROBABILITY

<u>Bayes Theorem with LTP</u>: Let  $E_1, ..., E_n$  be a partition of the sample space, and F an event. Then,

$$P(E_1|F) = \frac{P(F|E_1)P(E_1)}{P(F)} = \frac{P(F|E_1)P(E_1)}{\sum_{i=1}^{n} P(F|E_i)P(E_i)}$$

(Simple Partition) In particular, if E is an event with nonzero probability, then

$$P(E|F) = \frac{P(F|E)P(E)}{P(F)} = \frac{P(F|E)P(E)}{P(F|E)P(E) + P(F|E^{C})P(E^{C})}$$



# PROBABILITY 2.3 INDEPENDENCE

#### AGENDA

- CHAIN RULE
- INDEPENDENCE
- CONDITIONAL INDEPENDENCE

#### CHAIN RULE (IDEA)

HAVE A STANDARD 52-CARD DECK.

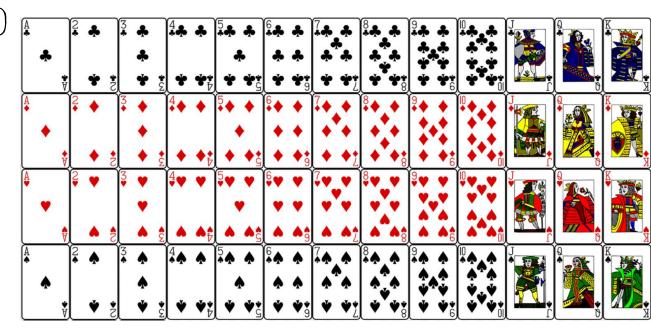
- 4 SUITS (CLUBS,
   DIAMONDS, HEARTS,
   SPADES)
- 13 RANKS (A, 2, 3, ..., 9, 10, J, Q, K)

## CHAIN RULE (IDEA)



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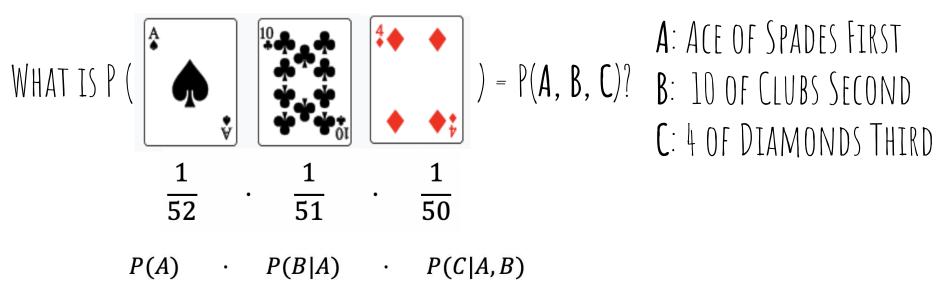


# CHAIN RULE (IDEA)



HAVE A STANDARD 52-CARD DECK. SHUFFLE IT, AND DRAW THE TOP 3 CARDS.

FORGET EVERYTHING YOU'VE LEARNED SO FAR ABOUT PROBABILITY.



#### CHAIN RULE

<u>Chain Rule:</u> Let  $A_1, ..., A_n$  be events with nonzero probability. Then,

$$P(A_1, ..., A_n) = P(A_1)P(A_2|A_1)P(A_3|A_1, A_2) ... P(A_n|A_1, ..., A_{n-1})$$

In the case of two events A, B,

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

An easy way to remember this formula: we need to do n tasks, so we can perform them one at a time, conditioning on what we've done so far.



#### THE NEED FOR INDEPENDENCE



Quick question: In general, is

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

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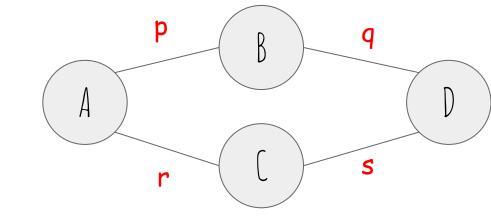
$$P(A,B) = P(A)P(B|A)$$

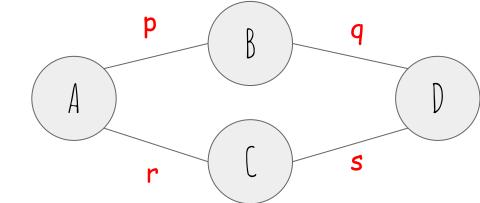
So no, unless the special case when P(B|A) = P(B). This case is so important it has a name.

#### INDEPENDENCE

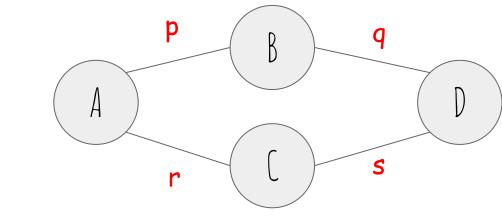
<u>Independence</u>: Events A, B are independent if any of the three equivalent conditions hold:

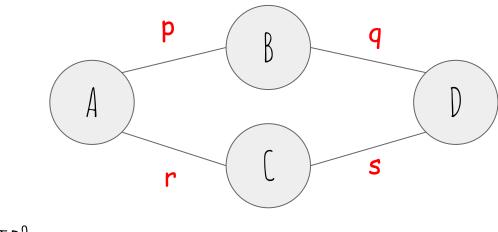
- 1. P(A|B) = P(A)
- **2**. P(B|A) = P(B)
- 3. P(A, B) = P(A)P(B)



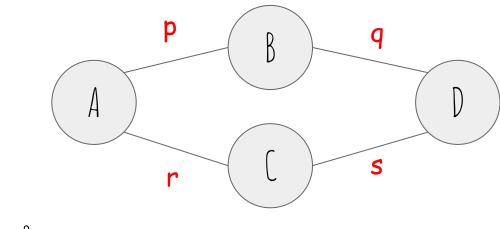


$$P(top) = P(AB \cap BD) =$$

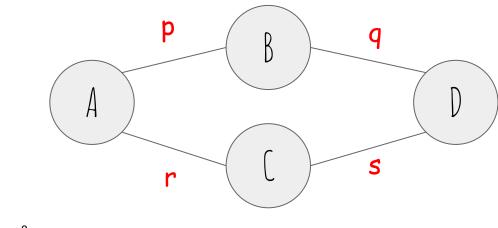




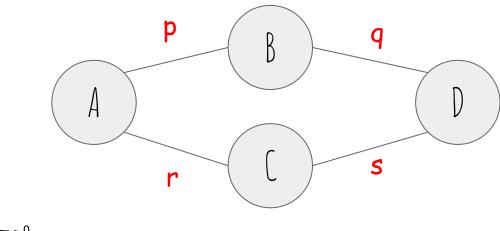
$$P(top) = P(AB \cap BD) = P(AB)P(BD) =$$
 INDEPENDENCE



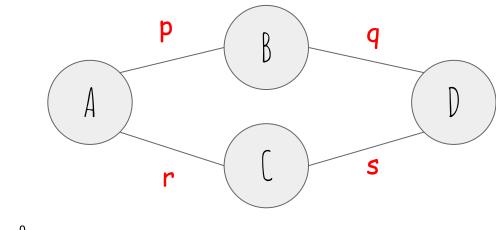
$$P(top) = P(AB \cap BD) = P(AB)P(BD) = pq$$



$$P(top) = P(AB \cap BD) = P(AB)P(BD) = pq$$
  
 $P(bottom) = P(AC \cap CD) =$ 

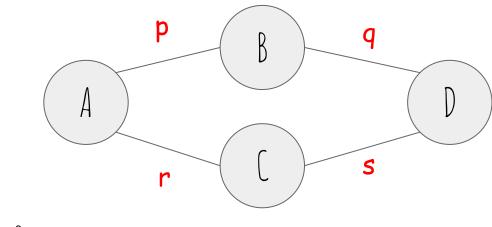


$$P(top) = P(AB \cap BD) = P(AB)P(BD) = pq$$
  
 $P(bottom) = P(AC \cap CD) = P(AC)P(CD) = rs$  INDEPENDENCE



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$$P(top \cup bottom) = P(top) + P(bottom) - P(top \cap bottom)$$



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=  $P(top) + P(bottom) - P(top)P(bottom)$  INDEPENDENCE

EACH LINK WORKS WITH THE PROBABILITY GIVEN, INDEPENDENTLY. WHAT'S THE PROBABILITY A AND D CAN COMMUNICATE?

$$P(top) = P(AB \cap BD) = P(AB)P(BD) = pq$$
  
 $P(bottom) = P(AC \cap CD) = P(AC)P(CD) = rs$ 

S

$$P(top \cup bottom) = P(top) + P(bottom) - P(top \cap bottom)$$

$$= P(top) + P(bottom) - P(top)P(bottom)$$

$$= pq + rs - pqrs$$

#### CONDITIONAL INDEPENDENCE

<u>Conditional Independence:</u> Events A, B are (conditionally) independent given C if any of the three equivalent conditions hold:

- 1. P(A|B,C) = P(A|C)
- **2**. P(B|A,C) = P(B|C)
- 3. P(A,B|C) = P(A|C)P(B|C)

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**Independence:** Events A, B are independent if any of the three equivalent conditions hold:

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### COIN FLIPPING

Suppose there is a coin  $C_1$  with P(head) = 0.3 and a coin  $C_2$  with P(head) = 0.9. We pick one randomly with equal probability and flip that coin 3 times independently. What is the probability we get all heads?

