

# 04: Conditional Probability and Bayes

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Lisa Yan and Jerry Cain  
September 21, 2020

# Quick slide reference

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3	Conditional Probability + Chain Rule	04a_conditional
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# Conditional Probability

# Dice, our misunderstood friends

Roll two 6-sided dice, yielding values  $D_1$  and  $D_2$ .

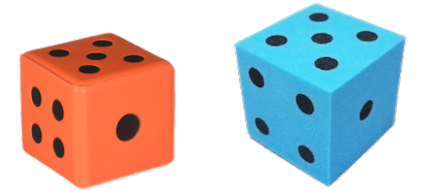
Let  $E$  be event:  $D_1 + D_2 = 4$ .

What is  $P(E)$ ?

$$|S| = 36 \quad \underline{6} \cdot \underline{6}$$

$$E = \{(1,3), (2,2), (3,1)\}$$

$$P(E) = 3/36 = 1/12$$



Let  $F$  be event:  $D_1 = 2$ .

What is  $P(E, \text{given } F \text{ already observed})$ ?

$$S = \{(2,1), (2,2), (2,3), (2,4), (2,5), (2,6)\}$$

$$E = \{(2,2)\}$$

$$P(E, \text{given } F \text{ already observed}) = \frac{1}{6}$$

# Conditional Probability

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The **conditional probability** of  $E$  given  $F$  is the probability that  $E$  occurs given that  $F$  has already occurred. This is known as conditioning on  $F$ .

Written as:	$P(E F)$
Means:	“ $P(E, \text{ given } F \text{ already observed})$ ”
Sample space $\rightarrow$	all possible outcomes consistent with $F$ (i.e. $S \cap F$ )
Event $\rightarrow$	all outcomes in $E$ consistent with $F$ (i.e. $E \cap F$ )

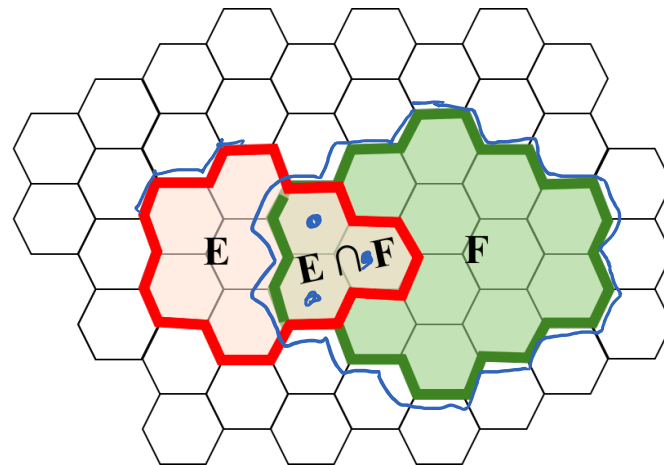
# Conditional Probability, equally likely outcomes

The **conditional probability** of  $E$  given  $F$  is the probability that  $E$  occurs given that  $F$  has already occurred. This is known as conditioning on  $F$ .

With **equally likely outcomes**:

$$P(E|F) = \frac{\# \text{ of outcomes in } E \text{ consistent with } F}{\# \text{ of outcomes in } S \text{ consistent with } F} = \frac{|E \cap F|}{|S \cap F|}$$

$$P(E|F) = \frac{|EF|}{|F|}$$



$|S| = 50$

$$P(E) = \frac{8}{50} \approx 0.16$$
$$P(E|F) = \frac{3}{14} \approx 0.21$$

# Slicing up the spam

$$P(E|F) = \frac{|EF|}{|F|}$$

Equally likely  
outcomes

24 emails are sent, 6 each to 4 users.

- 10 of the 24 emails are spam.
- All possible outcomes are equally likely.

Let  $E$  = user 1 receives  
3 spam emails.

What is  $P(E)$ ?

Let  $F$  = user 2 receives  
6 spam emails.

What is  $P(E|F)$ ?

Let  $G$  = user 3 receives  
5 spam emails.

What is  $P(G|F)$ ?



# Slicing up the spam

$$P(E|F) = \frac{|EF|}{|F|} \quad \text{Equally likely outcomes}$$

24 emails are sent, 6 each to 4 users.

- 10 of the 24 emails are spam.
- All possible outcomes are equally likely.

Let  $E$  = user 1 receives 3 spam emails.

What is  $P(E)$ ?

$$P(E) = \frac{\binom{10}{3} \binom{14}{3}}{\binom{24}{6}} \approx 0.3245$$

Let  $F$  = user 2 receives 6 spam emails. <sup>"honeypot"</sup>

What is  $P(E|F)$ ?

$$P(E|F) = \frac{\binom{4}{3} \binom{14}{3}}{\binom{18}{6}} \approx 0.0784$$

Let  $G$  = user 3 receives 5 spam emails.

What is  $P(G|F)$ ?

$$P(G|F) = \frac{\binom{4}{5} \binom{14}{1}}{\binom{18}{6}} = 0$$

No way to choose 5 spam from 4 remaining spam emails!



# Conditional probability in general

General **definition** of conditional probability:

$$P(E|F) = \frac{P(EF)}{P(F)}$$

ELD:

$$P(E|F) = \frac{|EF|}{|F|}$$
$$= \frac{|EF|/|S|}{|F|/|S|}$$

The **Chain Rule** (aka Product rule):

$$P(EF) = P(F)P(E|F)$$
$$= P(E)P(F|E)$$

Note:  $P(E \cap F) = P(F \cap E)$

These properties hold even when outcomes are not equally likely.

**NETFLIX**

# Netflix and Learn

$$P(E|F) = \frac{P(EF)}{P(F)}$$
 Definition of  
Cond. Probability

Let  $E$  = a user watches Life is Beautiful.

What is  $P(E)$ ?

Equally likely outcomes?

$S = \{\text{watch, not watch}\}$

$E = \{\text{watch}\}$

$P(E) = 1/2$  ?



$$\checkmark P(E) = \lim_{n \rightarrow \infty} \frac{n(E)}{n} \approx \frac{\# \text{ people who have watched movie}}{\# \text{ people on Netflix}}$$

$$= 10,234,231 / 50,923,123 \approx 0.20$$

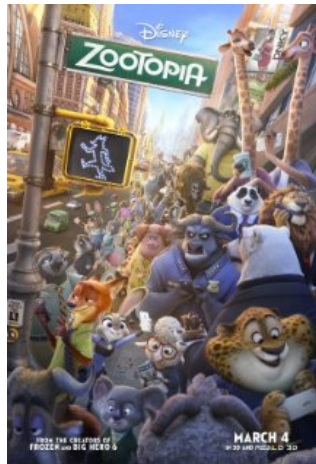
# Netflix and Learn

$$P(E|F) = \frac{P(EF)}{P(F)}$$
 Definition of  
Cond. Probability

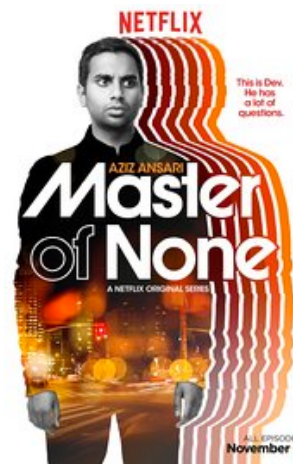
Let  $E$  be the event that a user watches the given movie.



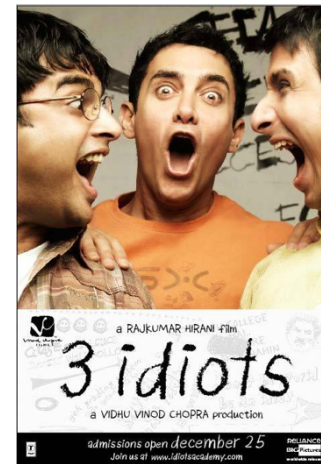
$$P(E) = 0.19$$



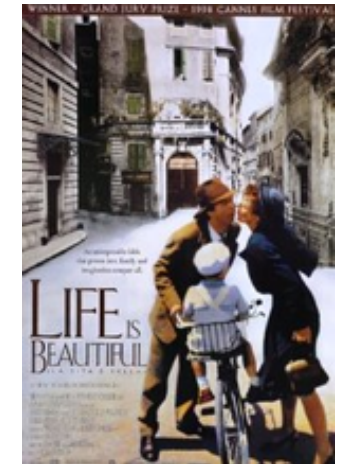
$$P(E) = 0.32$$



$$P(E) = 0.20$$



$$P(E) = 0.09$$



$$P(E) = 0.20$$

# Netflix and Learn

$$P(E|F) = \frac{P(EF)}{P(F)}$$
 Definition of Cond. Probability

Let  $E$  = a user watches Life is Beautiful.

Let  $F$  = a user watches Amelie.

What is the probability that a user watches Life is Beautiful, given they watched Amelie?

$$P(E|F)$$

$$\begin{aligned} P(E|F) &= \frac{P(EF)}{P(F)} = \frac{\frac{\# \text{ people who have watched both}}{\# \text{ people on Netflix}}}{\frac{\# \text{ people who have watched Amelie}}{\# \text{ people on Netflix}}} \\ &= \frac{\# \text{ people who have watched both}}{\# \text{ people who have watched Amelie}} \end{aligned}$$

$$\approx 0.42$$



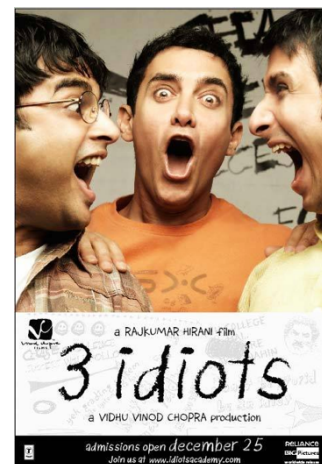
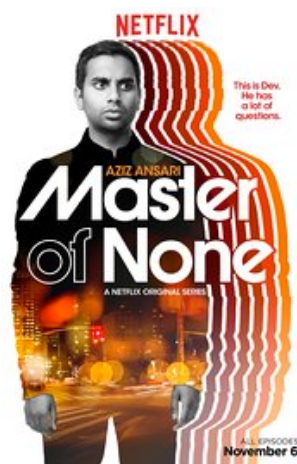
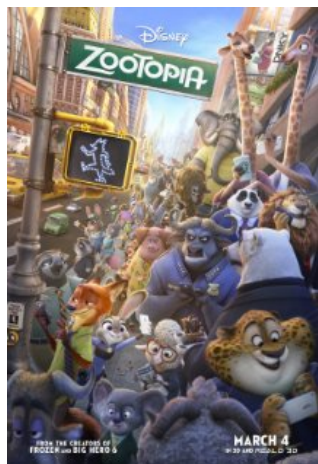
data:  
for each user,  
which movies  
do they watch?

# Netflix and Learn

$$P(E|F) = \frac{P(EF)}{P(F)}$$

Definition of  
Cond. Probability

Let  $E$  be the event that a user watches the given movie.  
Let  $F$  be the event that the same user watches Amelie.



$$P(E) = 0.19$$

$$P(E) = 0.32$$

$$P(E) = 0.20$$

$$P(E) = 0.09$$

$$P(E) = 0.20$$

$$P(E|F) = 0.14$$

$$P(E|F) = 0.35$$

$$P(E|F) = 0.20$$

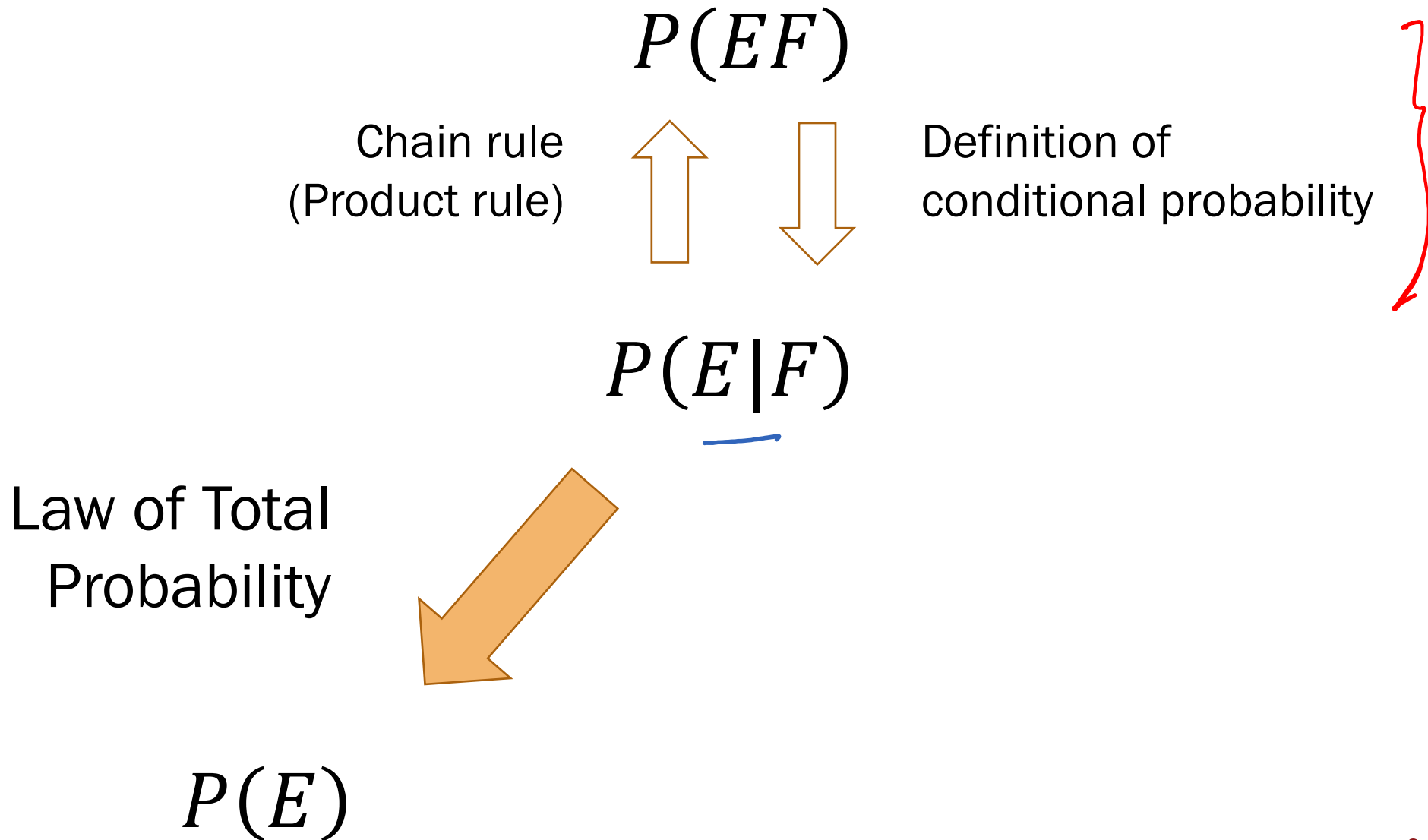
$$P(E|F) = 0.72$$

$$P(E|F) = 0.42$$

# Law of Total Probability

# Today's tasks

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# Law of Total Probability

Thm Let  $F$  be an event where  $P(F) > 0$ . For any event  $E$ ,

$$P(E) = P(E|F)P(F) + P(E|F^C)P(F^C)$$

*need:*  
 $P(E|F)$   
 $P(E|F^C)$   
 $P(F)$

Proof

1.  $F$  and  $F^C$  are disjoint s.t.  $F \cup F^C = S$

*such that*

2.  $E = (EF) \cup (EF^C)$

3.  $P(E) = P(EF) + P(EF^C)$

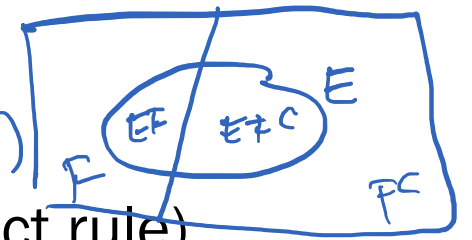
4.  $P(E) = P(\underline{E|F})P(F) + P(E|F^C)P(F^C)$

Def. of complement

(see diagram)

Additivity axiom (3)

Chain rule (product rule)



Note: disjoint sets by definition are mutually exclusive events

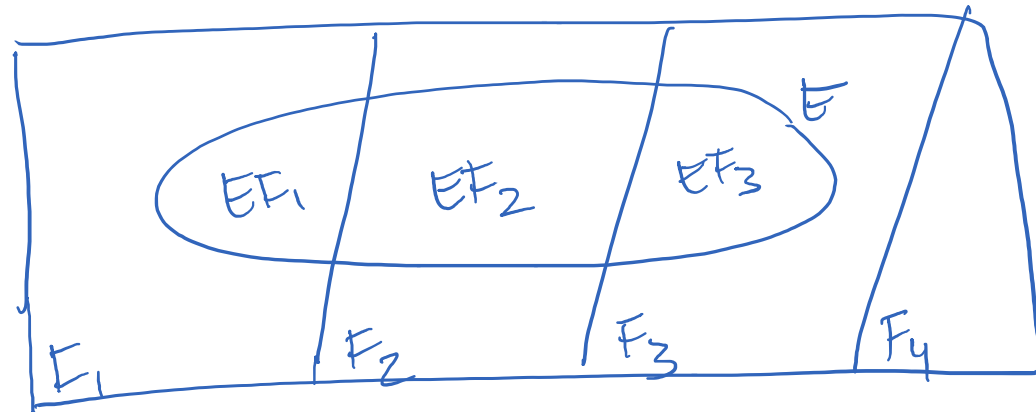
# General Law of Total Probability

Thm For **mutually exclusive events**  $F_1, F_2, \dots, F_n$   
s.t.  $F_1 \cup F_2 \cup \dots \cup F_n = S$ ,

$$P(E) = \sum_{i=1}^n P(E|F_i)P(F_i)$$

$\underbrace{\hspace{10em}}_{P(E|F_i)}$

need:  
 $P(E|F_i)$   
where  $F_1 \cup F_2 \cup \dots \cup F_n = S$   
 $P(F_i)$



# Finding $P(E)$ from $P(E|F)$

$$P(E) = P(E|F)P(F) + P(E|F^C)P(F^C)$$

Law of Total Probability

- Flip a fair coin.
- If heads: roll a fair 6-sided die.
- Else: roll a fair 3-sided die.

You win if you roll a 6. What is  $P(\text{winning})$ ?

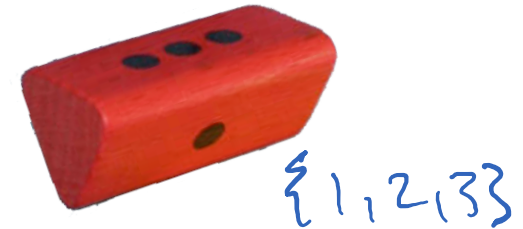


# Finding $P(E)$ from $P(E|F)$

$$P(E) = P(E|F)P(F) + P(E|F^C)P(F^C)$$

Law of Total Probability

- Flip a fair coin.
- If heads: roll a fair 6-sided die.
- Else: roll a fair 3-sided die.



You win if you roll a 6. What is  $P(\text{winning})$ ?

1. Define events  
& state goal

Let:  $E$ : win,  $F$ : flip heads  
Want:  $P(\text{win})$   
 $= P(E)$

2. Identify known  
probabilities

$$\begin{aligned} P(\text{win}|H) &= P(E|F) = 1/6 \\ P(H) &= P(F) = 1/2 \\ P(\text{win}|T) &= P(E|F^C) = 0 \\ P(T) &= P(F^C) = 1 - 1/2 \end{aligned}$$

3. Solve

$$\begin{aligned} P(E) &= (1/6)(1/2) \\ &\quad + (0)(1/2) \\ &= \boxed{\frac{1}{12}} \approx 0.083 \end{aligned}$$

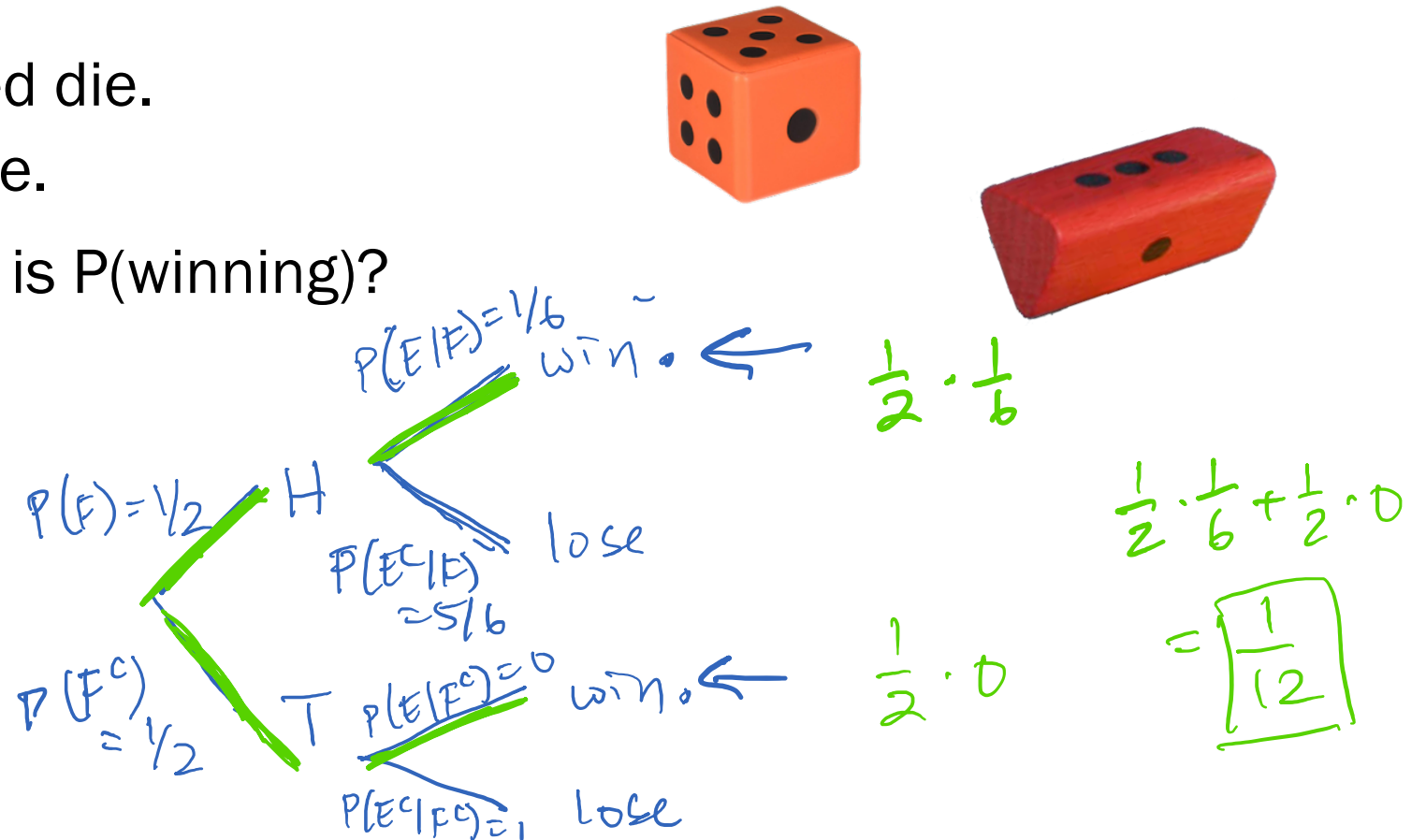
# Finding $P(E)$ from $P(E|F)$ , an understanding

- Flip a fair coin.
- If heads: roll a fair 6-sided die.
- Else: roll a fair 3-sided die.

You win if you roll a 6. What is  $P(\text{winning})$ ?

## 1. Define events & state goal

Let:  $E$ : win,  $F$ : flip heads  
Want:  $P(\text{win})$   
 $= P(E)$



“Probability trees” can help connect your understanding of the experiment with the problem statement.

# Bayes' Theorem

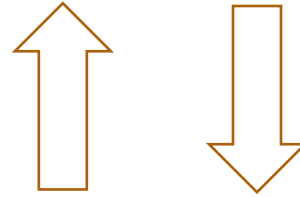
# I

# Today's tasks



Chain rule  
(Product rule)

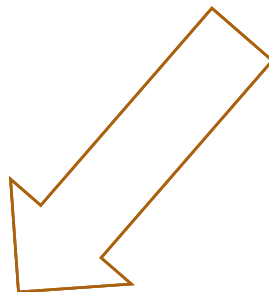
$$P(EF)$$



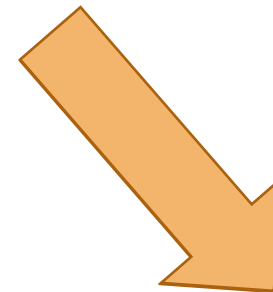
Definition of  
conditional probability

$$P(E|F)$$

Law of Total  
Probability



Bayes'  
Theorem



$$P(E)$$

$$P(F|E)$$

# Thomas Bayes

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Rev. Thomas Bayes (~1701-1761):  
British mathematician and Presbyterian minister

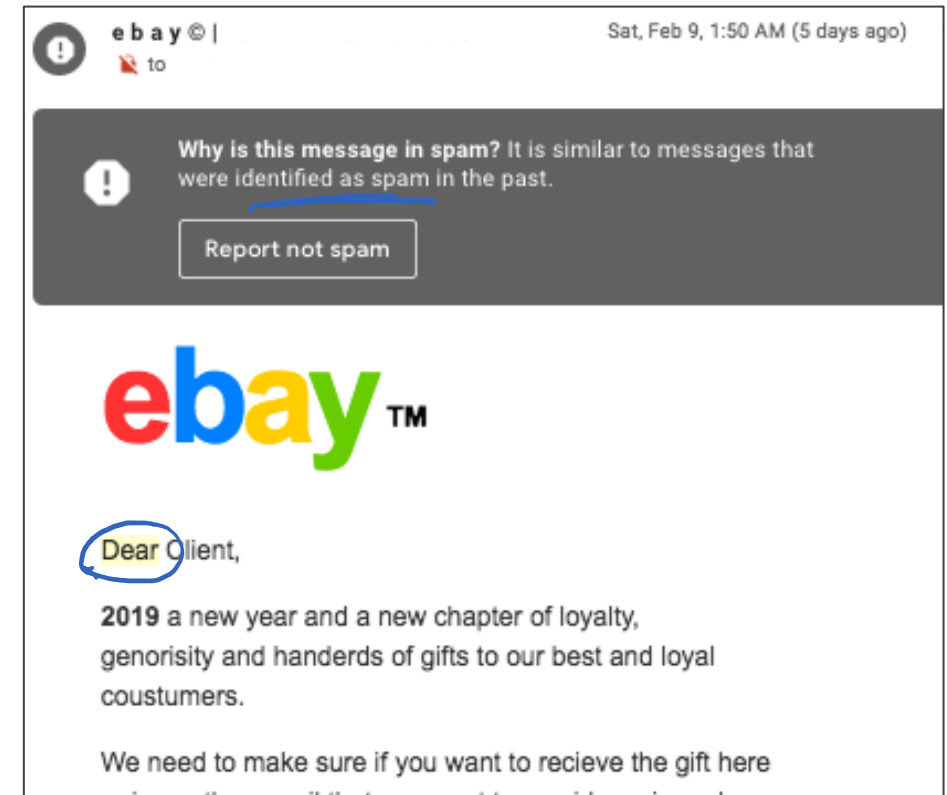
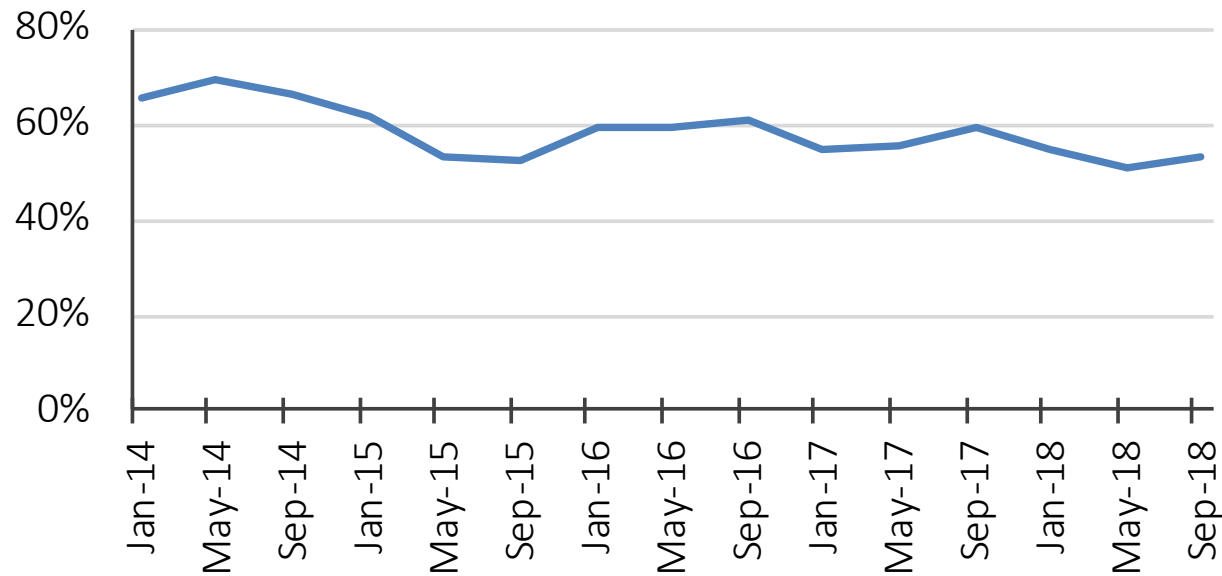


He looked remarkably similar to Charlie Sheen  
(but that's not important right now)



# Detecting spam email

Spam volume as percentage of total email traffic worldwide



We can easily calculate how many *existing* spam emails contain “Dear”:

$$P(E|F) = P(\text{“Dear”} \mid \text{Spam email})$$



But what is the probability that an *unknown* email containing “Dear” is spam?

$$P(F|E) = P(\text{Spam email} \mid \text{“Dear”})$$

(silent drumroll)

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# Bayes' Theorem

$$P(E|F) \Rightarrow P(F|E)$$

Thm For any events  $E$  and  $F$  where  $P(E) > 0$  and  $P(F) > 0$ ,

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

need:  
 $P(E|F) \cdot P(F)$   
expanded form:  $P(E|F^c)$

Proof

2 steps! See board

1.  $P(F|E) = \frac{P(EF)}{P(E)}$  definition of conditional prob.  
2.  $= \frac{P(E|F)P(F)}{P(E)}$  Chain Rule

Expanded form:

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

Proof

1 more step! See board

3.  $= \frac{P(E|F)P(F)}{P(E|F)P(F) + P(E|F^c)P(F^c)}$  Law of Total Prob.



# Detecting spam email

$$P(F|E) = \frac{P(E|F)P(F)}{P(E|F)P(F) + P(E|F^c)P(F^c)} \quad \begin{array}{l} \text{Bayes'} \\ \text{Theorem} \end{array}$$

- 60% of all email in 2016 is spam.
- 20% of <sup>known</sup> spam has the word “Dear”
- 1% of <sup>known</sup> non-spam (aka ham) has the word “Dear”

$$P(F) = 0.6$$

$$P(E|F) = 0.2$$

$$P(E|F^c) = 0.01$$

You get an <sup>unknown</sup> email with the word “Dear” in it.

What is the probability that the email is spam?

1. Define events  
& state goal

2. Identify known  
probabilities

3. Solve

Let:  $E$ : “Dear”,  $F$ : spam  
Want:  $P(\text{spam} | \text{“Dear”})$   
 $= P(F|E)$

$$P(F|E) = \frac{(0.20)(0.6)}{(0.20)(0.6) + (0.01)(0.4)} \approx 0.967$$

# Detecting spam email, an understanding

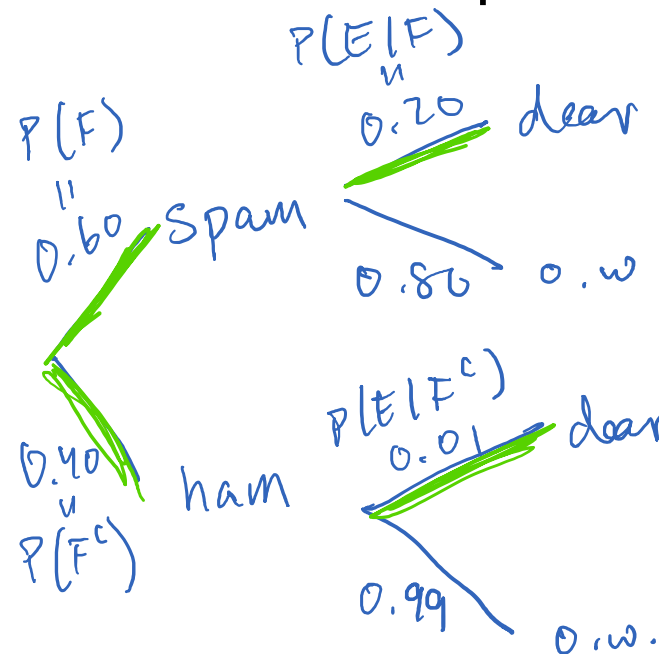
- 60% of all email in 2016 is spam.
- 20% of spam has the word “Dear”
- 1% of non-spam (aka ham) has the word “Dear”

You get an email with the word “Dear” in it.  
What is the probability that the email is spam?

## 1. Define events & state goal

Let:  $E$ : “Dear”,  $F$ : spam

Want:  $P(\text{spam} | \text{“Dear”})$   
 $= P(F | E)$ .



**Note:** You should still know how to use Bayes/ Law of Total Probab., but drawing a probability tree can help you identify which probabilities you have. The branches are determined using the problem setup.

$$\frac{(0.60)(0.20)}{(0.60)(0.20) + (0.40)(0.01)} \approx 0.967$$

# Bayes' Theorem terminology

- 60% of all email in 2016 is spam.
- 20% of spam has the word “Dear”
- 1% of non-spam (aka ham) has the word “Dear”

$P(F)$  prior

$P(E|F)$  likelihood

$P(E|F^C)$  no special term

You get an email with the word “Dear” in it.

What is the probability that the email is spam?

Want:  $P(F|E)$  posterior

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

posterior

likelihood prior

normalization constant

F: Fact

E: Evidence

(live)

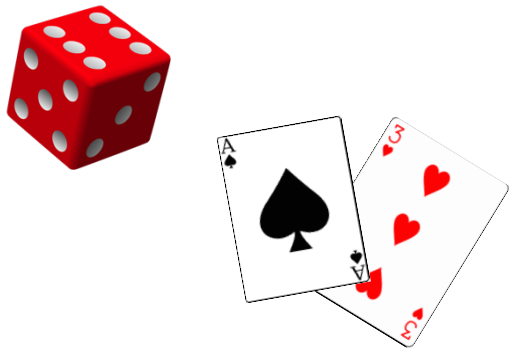
# 04: Conditional Probability and Bayes

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Lisa Yan and Jerry Cain  
September 21, 2020

# This class going forward

Last week  
Equally likely  
events



$$P(E \cap F) \quad P(E \cup F)$$

(counting, combinatorics)

Today and for most of this course  
**Not equally likely events**

$$P(E = \text{Evidence} \mid F = \text{Fact})$$

(collected from data)

Bayes'

$$P(F = \text{Fact} \mid E = \text{Evidence})$$

(categorize  
a new datapoint)



General definition of conditional probability:

$$P(E|F) = \frac{P(EF)}{P(F)}$$

F: FACT

E: EVIDENCE

The Chain Rule (aka Product rule):

$$P(EF) = P(F)P(E|F)$$

*(Handwritten red annotations: a bracket under P(EF), a bracket under P(F), a circle around P(E|F), and an equals sign followed by P(E)P(E|F) below the main equation.)*

These properties hold even when outcomes are not equally likely.

# Think, then Breakout Rooms

Then check out the question on the next slide (Slide 35). Post any clarifications here!

<https://us.edstem.org/courses/2678/discussion/128395>

Think by yourself: 1 min

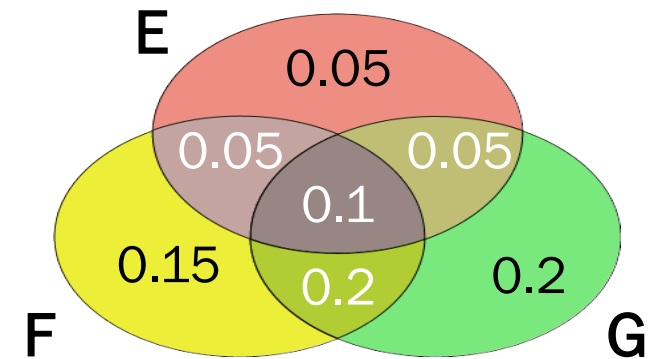
Breakout rooms: 4 min. Introduce yourself!



# Think, then groups

You have a flowering plant.

Let  $E$  = Flowers bloom  
 $F$  = Plant was watered  
 $G$  = Plant got sun



1. How would you write

- the probability that the plant got sun, <sup>G</sup> given that it was watered and flowers bloomed? <sup>EF</sup>
- the probability that the plant got sun <sup>G</sup> and flowers <sup>E</sup> bloomed given that it was watered? <sup>F</sup>

$$P(G|FE) =$$

$$P(GE|F) =$$

2. Using the Venn diagram, compute the above probabilities.

3. Chain Rule: Fill in the blanks.

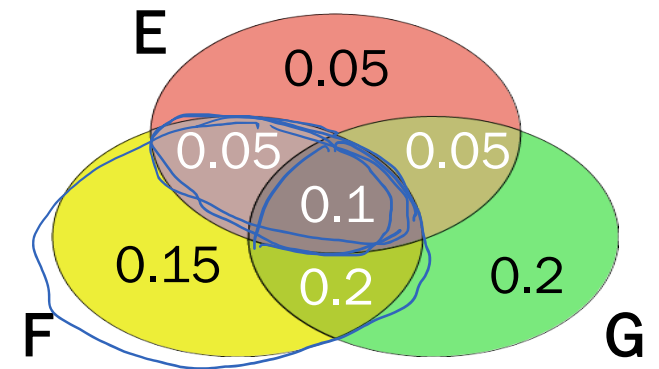
- $P(GE) = \underline{\hspace{2cm}} \cdot P(E)$
- $P(GE|F) = P(G|EF) \cdot \underline{\hspace{2cm}}$



# Think, then groups

You have a flowering plant.

Let  $E$  = Flowers bloom  
 $F$  = Plant was watered  
 $G$  = Plant got sun



1. How would you write

- i. the probability that the plant got sun, given that it was watered and flowers bloomed?
- ii. the probability that the plant got sun and flowers bloomed given that it was watered?

$$P(G|FE) = \frac{P(FEG)}{P(FE)} = \frac{0.1}{0.15} = \frac{2}{3}$$

$$P(GE|F) = \frac{P(FEG)}{P(F)} = \frac{0.1}{0.15} = \frac{2}{3}$$

2. Using the Venn diagram, compute the above probabilities.

3. Chain Rule: Fill in the blanks.

i.  $P(GE) = P(G|E) \cdot P(E)$

ii.  $P(\underline{GE}|F) = P(\underline{G|EF}) \cdot \underline{P(E|F)}$

$$\frac{P(\cancel{GE})}{P(F)}$$

# Bayes' Theorem II


# Why is Bayes' so important?



It links **belief** to **evidence** in probability!

$$\begin{array}{ccc} \text{posterior} & \text{likelihood} & \text{prior} \\ P(F|E) = & \frac{P(E|F)P(F)}{P(E)} \end{array}$$

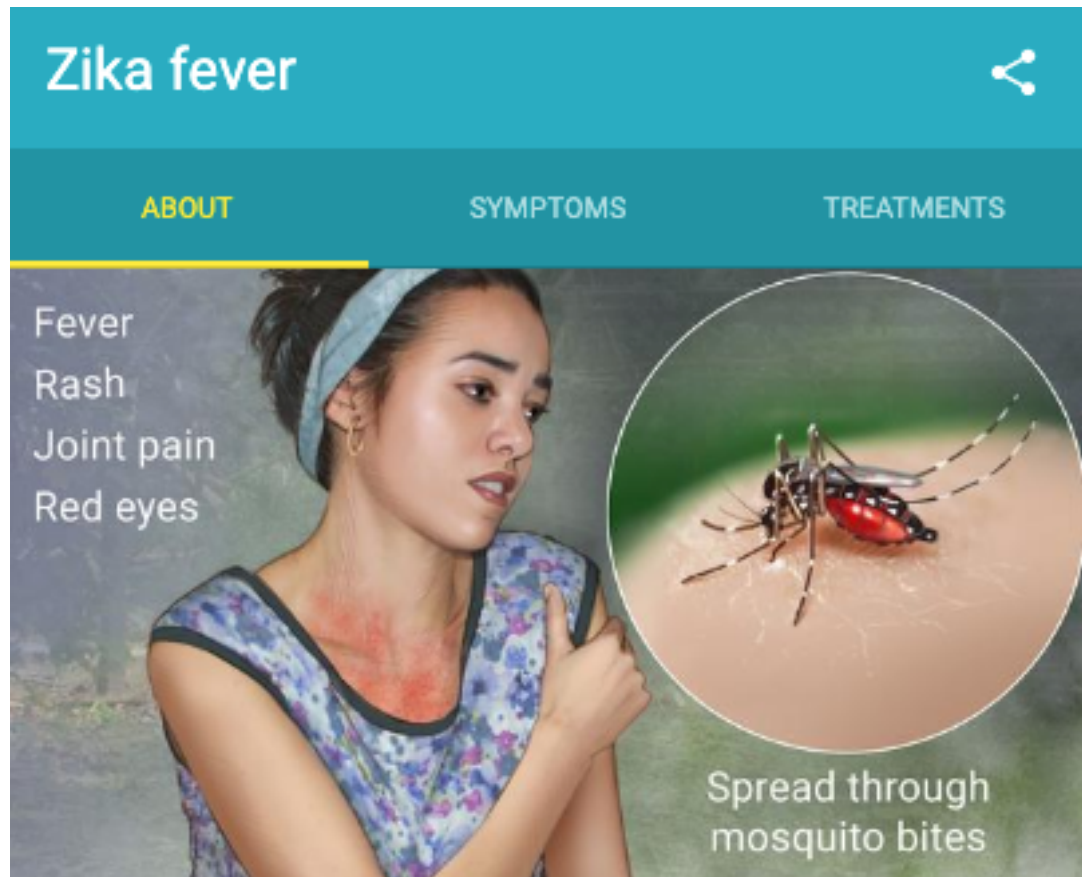
Mathematically:


$$P(E|F) \rightarrow P(F|E)$$

Real-life application:

Given new evidence  $E$ , update belief of fact  $F$   
Prior belief  $\rightarrow$  Posterior belief  
 $P(F) \rightarrow P(F|E)$

# Zika, an autoimmune disease



Ziika Forest, Uganda



Rhesus monkeys

If a test returns positive, what is the likelihood you have the disease?

A disease spread through mosquito bites. Usually no symptoms; worst case paralysis. During pregnancy: may cause birth defects



# Taking tests: Confusion matrix



Fact,  $F$  Has disease  
or  $F^C$  No disease



Evidence,  $E$  Test positive  
or  $E^C$  Test negative

		Fact	
		$F$ , disease +	$F^C$ , disease -
Evidence	$E$ , Test +	True positive $P(E F)$	False positive $P(E F^C)$
	$E^C$ , Test -	False negative $P(E^C F)$	True negative $P(E^C F^C)$

If a test returns positive, what is the likelihood you have the disease?

# Taking tests: Confusion matrix



Fact,  $F$  Has disease  
or  $F^C$  No disease



Evidence,  $E$  Test positive  
or  $E^C$  Test negative

		Fact	
		$F$ , disease +	$F^C$ , disease -
Evidence	$E$ , Test +	True positive $P(E F)$	False positive $P(E F^C)$
	$E^C$ , Test -	False negative $P(E^C F)$	True negative $P(E^C F^C)$

If a test returns positive, what is the likelihood you have the disease?

# Breakout Rooms

Check out the question on the next slide (Slide 43). Post any clarifications here!

<https://us.edstem.org/courses/2678/discussion/128395>

Breakout rooms: 5 minutes



# Zika Testing

$$P(F|E) = \frac{P(E|F)P(F)}{P(E|F)P(F) + P(E|F^c)P(F^c)} \quad \begin{array}{l} \text{Bayes' Theorem} \end{array}$$

- A test is 98% effective at detecting Zika (“true positive”).  $P(E|F)$
- However, the test has a “false positive” rate of 1%.  $P(E|F^c)$
- 0.5% of the US population has Zika.  $P(F)$

What is the likelihood you have Zika if you test positive?

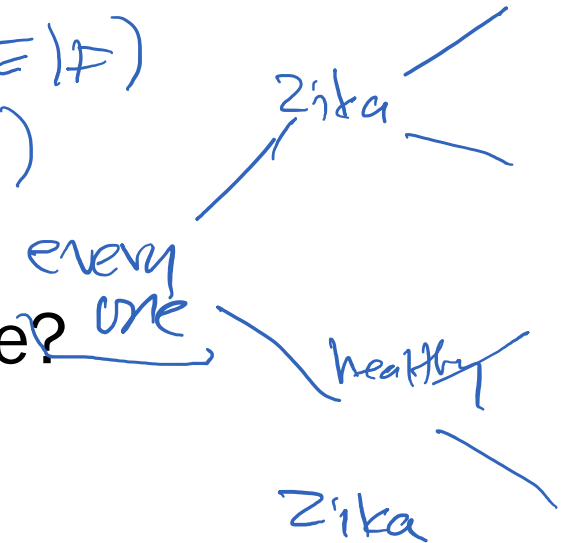
Why would you expect this number?

## 1. Define events & state goal

Let:  $E$  = you test positive *evidence*  
 $F$  = you actually have the disease *fact*

Want:

$$\begin{array}{l} P(\text{disease} \mid \text{test}+) \\ = P(F|E) \end{array}$$

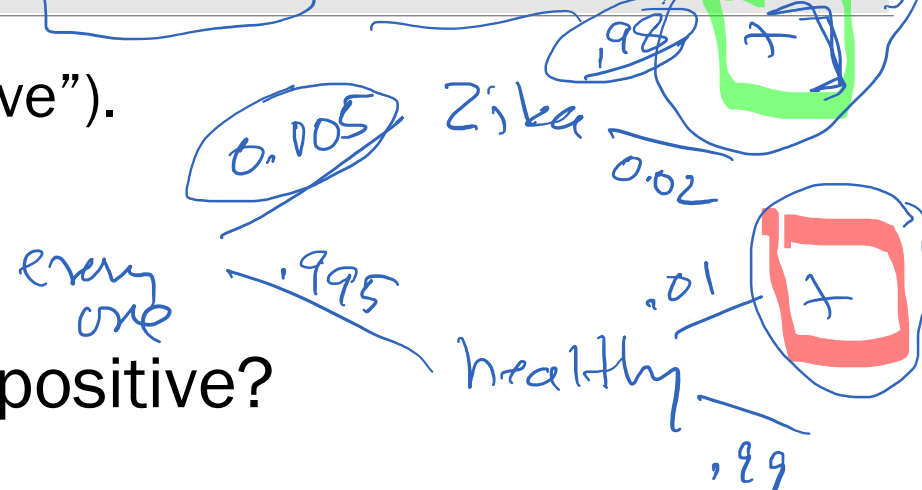


# Zika Testing

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

Bayes' Theorem

- A test is 98% effective at detecting Zika (“true positive”).
- However, the test has a “false positive” rate of 1%.
- 0.5% of the US population has Zika.



What is the likelihood you have Zika if you test positive?

Why would you expect this number?

1. Define events & state goal

2. Identify known probabilities

3. Solve

Let:  $E$  = you test positive  
 $F$  = you actually have the disease

Want:

$$P(\text{disease} \mid \text{test+}) = P(F|E)$$

$$P(F|E) = \frac{(0.98)(0.005)}{(0.98)(0.005) + (0.995)(0.01)} \approx 0.330$$

# Bayes' Theorem intuition

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Original question:

What is the likelihood you have Zika if you test positive for the disease?



# Bayes' Theorem intuition

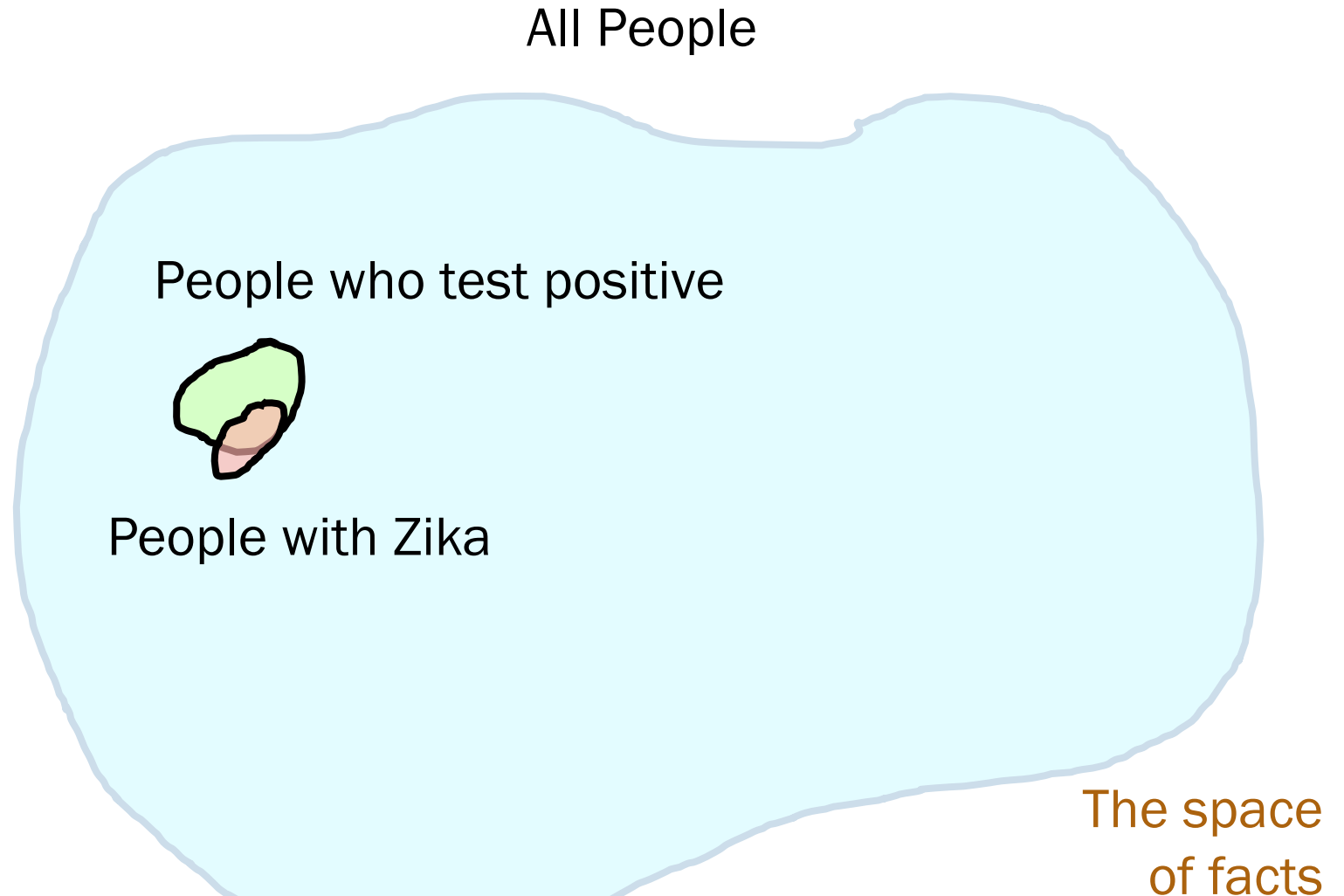
Original question:

What is the likelihood you have Zika if you test positive for the disease?

Interpret

Interpretation:

Of the people who test positive, how many actually have Zika?



# Bayes' Theorem intuition

Original question:

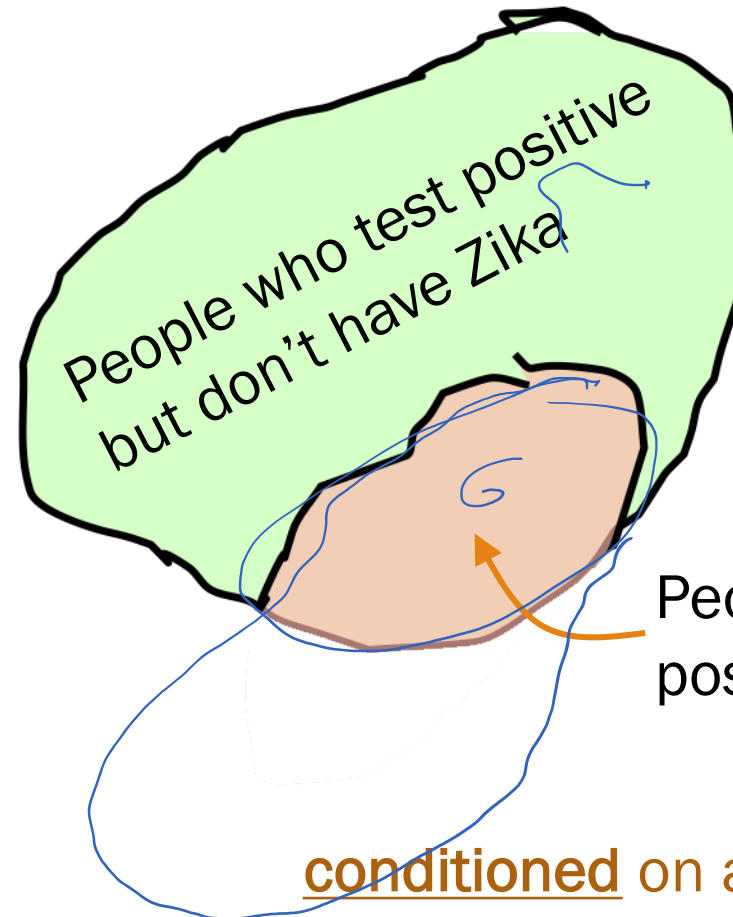
What is the likelihood you have Zika if you test positive for the disease?

Interpret

Interpretation:

Of the people who test positive, how many actually have Zika?

People who test positive





# Zika Testing

---

- A test is 98% effective at detecting Zika (“true positive”).
- However, the test has a “false positive” rate of 1%.
- 0.5% of the US population has Zika.

What is the likelihood you have Zika if you test positive?

Say we have 1000 people:



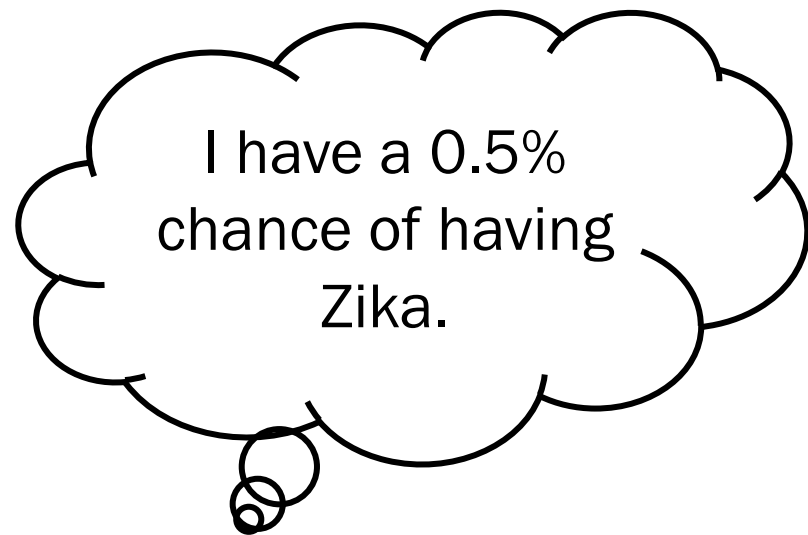
5 have Zika  
and test positive  
985 do not have Zika  
and test negative.  
10 do not have Zika  
and test positive.

$\approx 0.333$

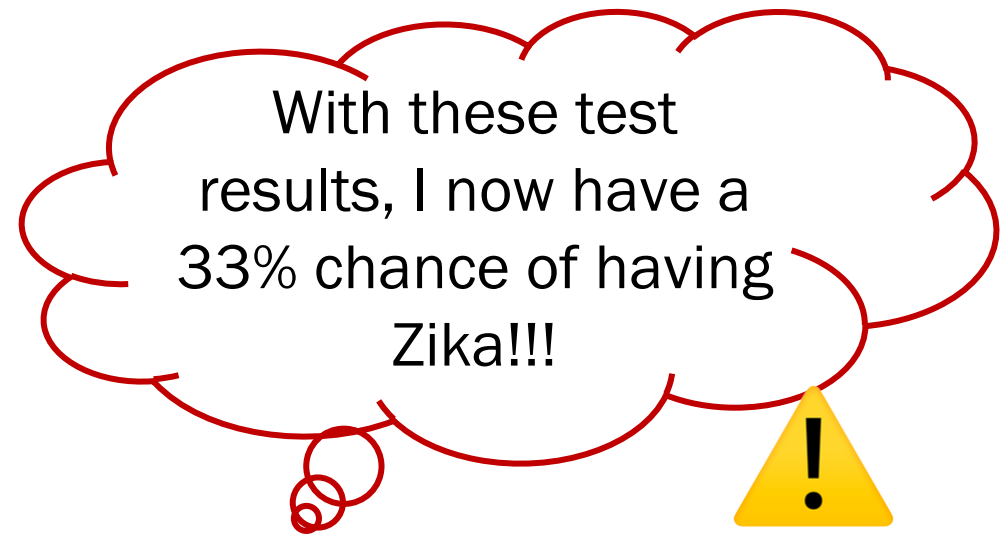
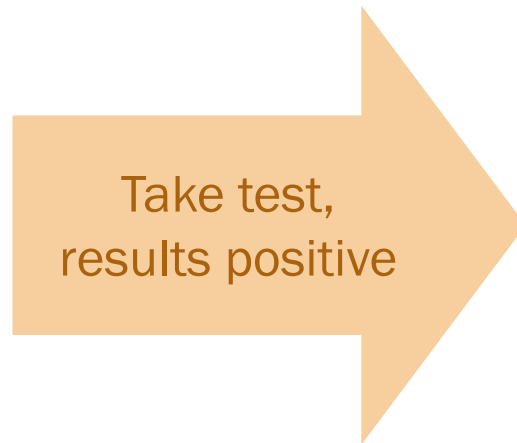
# Update your beliefs with Bayes' Theorem

$E$  = you test positive for Zika

$F$  = you actually have the disease



$P(F)$



$P(F|E)$

# Topical probability news: Bayes for COVID-19 testing

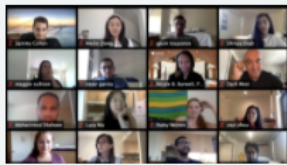


## Trial tests antibody drug as COVID-19 treatment

Stanford Medicine has joined a multisite clinical trial testing antibodies designed to block the coronavirus from infecting human cells and shorten the course of the illness.

September 9, 2020

[NEWS FEATURE →](#)



## New medical students intent on research

More than a third of the students starting medical school at Stanford plan to conduct research. The unprecedented number reflects an effort by the school to turn out more physician-scientists.

September 3, 2020

[NEWS FEATURE →](#)

**New tests (Calculated)**  
Total test results (mixed units)



**New cases (Calculated)**



**Current hospitalizations (Notes)**



**New deaths (Calculated)**



[Chart information and data](#) ↑



**A. Jul 10:**  
Florida started reporting hospitalizations of people with a "primary diagnosis of COVID-19."

*How representative are today's testing rates?*

*How do we know if a positive test is a true positive or a false positive?*

**Reasonable Question: Why test if there are errors?**

<https://covidtracking.com/data>

<http://med.stanford.edu/news.html>

# Think

Slide 53 is a question to think over by yourself.

We'll go over it together afterwards.

Post any clarifications here!

<https://us.edstem.org/courses/2678/discussion/128395>

Think by yourself: 2 minutes



# Why it's still good to get tested

$$P(F|E) = \frac{P(E|F)P(F)}{P(E|F)P(F) + P(E|F^C)P(F^C)} \quad \text{Bayes' Theorem}$$

- A test is 98% effective at detecting Zika (“true positive”).
- However, the test has a “false positive” rate of 1%.
- 0.5% of the US population has Zika.

Let:  $E$  = you test positive  
 $F$  = you actually have the disease

Let:  $E^C$  = you test **negative** for Zika with this test.

	$F$ , disease +	$F^C$ , disease -
$E$ , Test +	True positive $P(E F) = 0.98$	False positive $P(E F^C) = 0.01$

What is  $P(\underline{F}|\underline{E^C})$ ?



# Why it's still good to get tested

$$P(F|E) = \frac{P(E|F)P(F)}{P(E|F)P(F) + P(E|F^C)P(F^C)} \quad \text{Bayes' Theorem}$$

- A test is 98% effective at detecting Zika (“true positive”).
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 $F$  = you actually have the disease

Let:  $E^C$  = you test **negative** for Zika with this test.

	$F$ , disease +	$F^C$ , disease -
$E$ , Test +	True positive $P(E F) = 0.98$	False positive $P(E F^C) = 0.01$

What is  $P(F|E^C)$ ? =  ~~$1 - P(E|E)$~~  not true  
 =  $1 - P(F^C|E^C)$

# Why it's still good to get tested

$$P(F|E) = \frac{P(E|F)P(F)}{P(E|F)P(F) + P(E|F^C)P(F^C)} \quad \text{Bayes' Theorem}$$

- A test is 98% effective at detecting Zika (“true positive”).
- However, the test has a “false positive” rate of 1%.
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Let:  $E$  = you test positive  
 $F$  = you actually have the disease

Let:  $E^C$  = you test **negative** for Zika with this test.

	$F$ , disease +	$F^C$ , disease -
$E$ , Test +	True positive $P(E F) = 0.98$	False positive $P(E F^C) = 0.01$
$E^C$ , Test -	False negative $P(E^C F) = 0.02$	True negative $P(E^C F^C) = 0.99$

What is  $P(F|E^C)$ ?

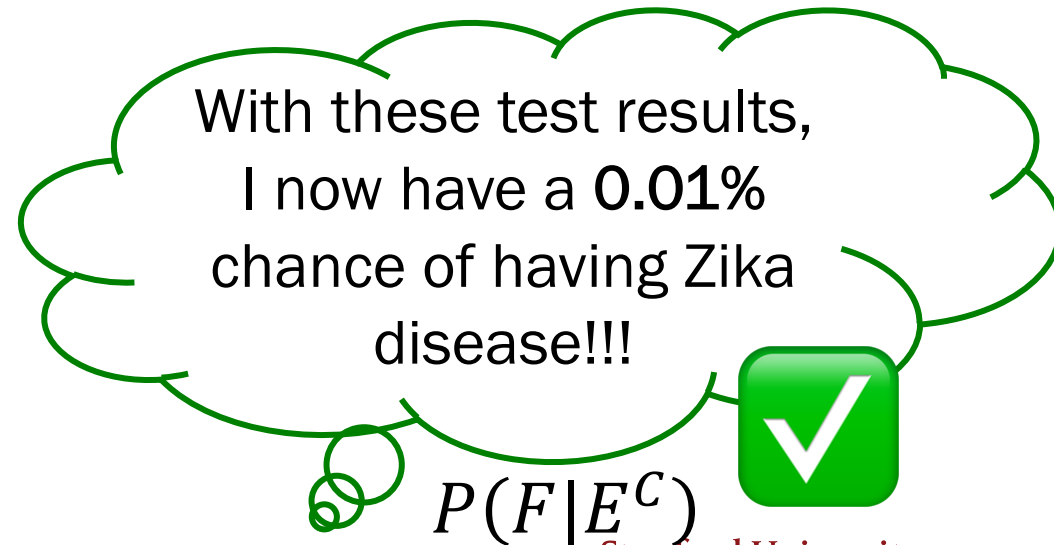
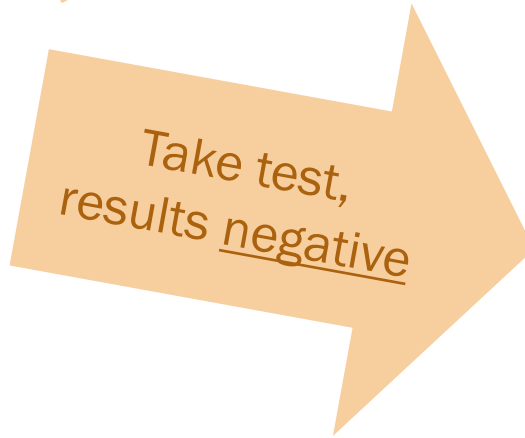
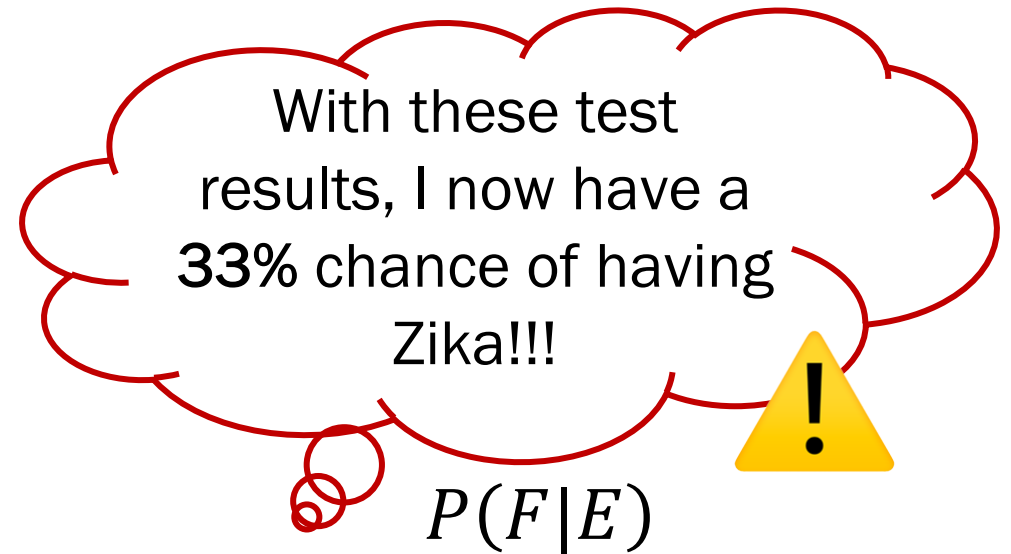
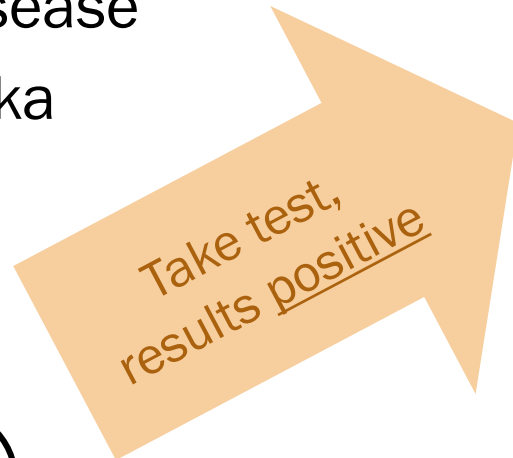
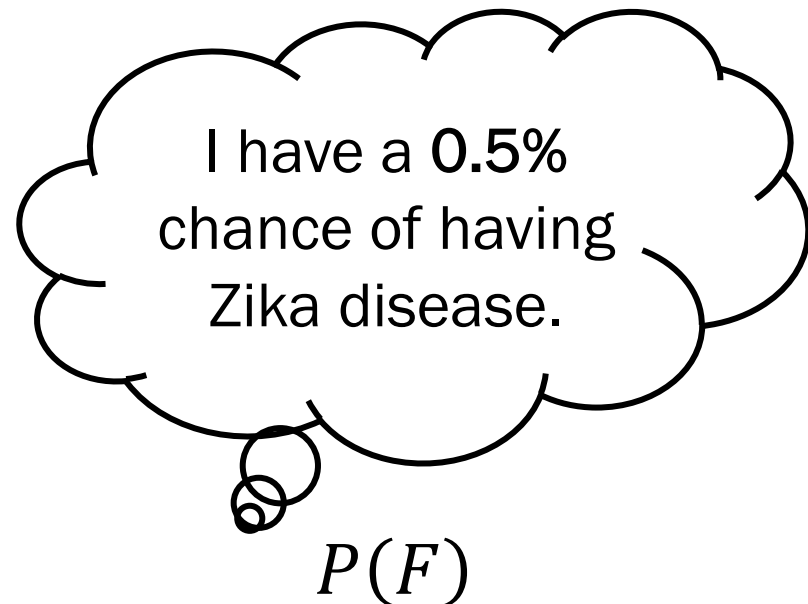
$$P(F|E^C) = \frac{P(E^C|F)P(F)}{P(E^C|F)P(F) + P(E^C|F^C)P(F^C)} \quad \Bigg] = 0.0001$$

# Why it's still good to get tested

$E$  = you test positive for Zika

$F$  = you actually have the disease

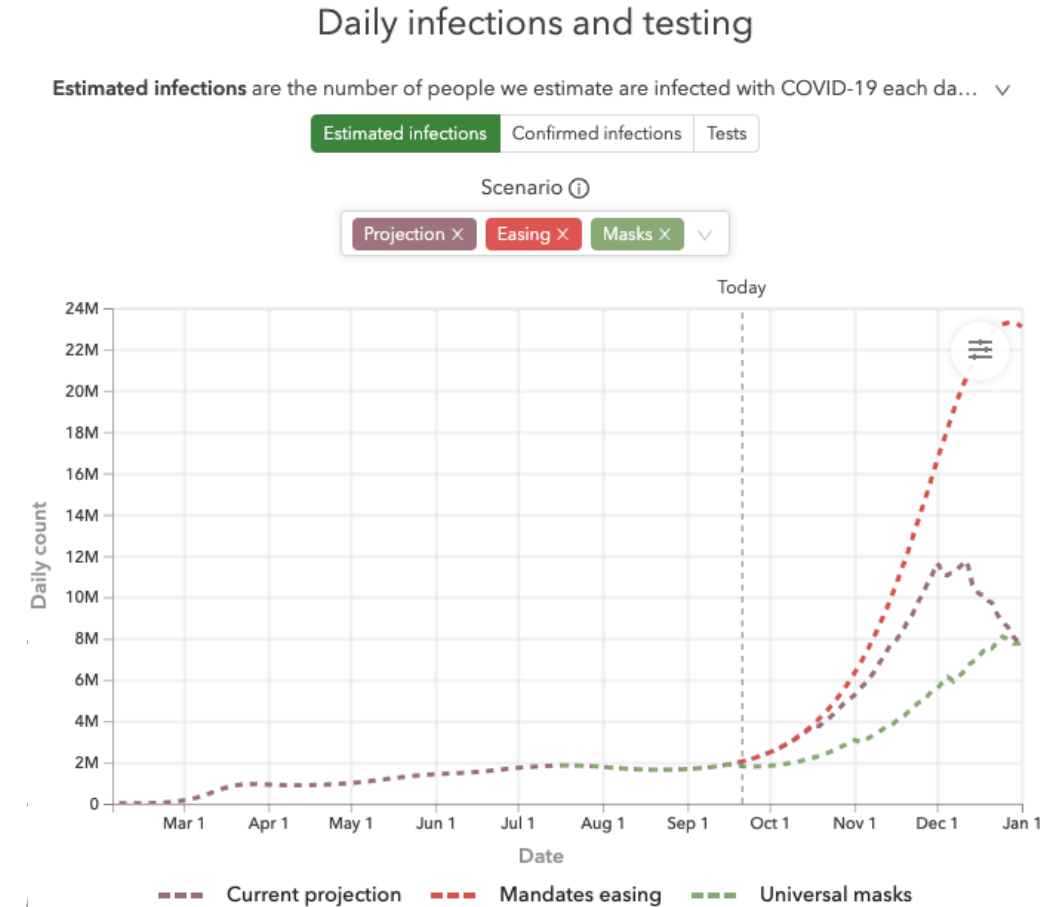
$E^C$  = you test **negative** for Zika





# Topical probability news: Bayes for COVID-19 testing

- Antibody tests (blood samples) have higher false negative, false positive rates than RT-PCR tests (nasal swab). However, they help explain/identify our body's reaction to the virus.
- The real world has many more “**givens**” (current symptoms, existing medical conditions) that improve our belief **prior** to testing.
- Most importantly, testing gives us a noisy signal of the spread of a disease.



*Why test if there are errors?*

# Topical probability news: Sources

## COVID-19 Projections

<http://covid19.healthdata.org/>

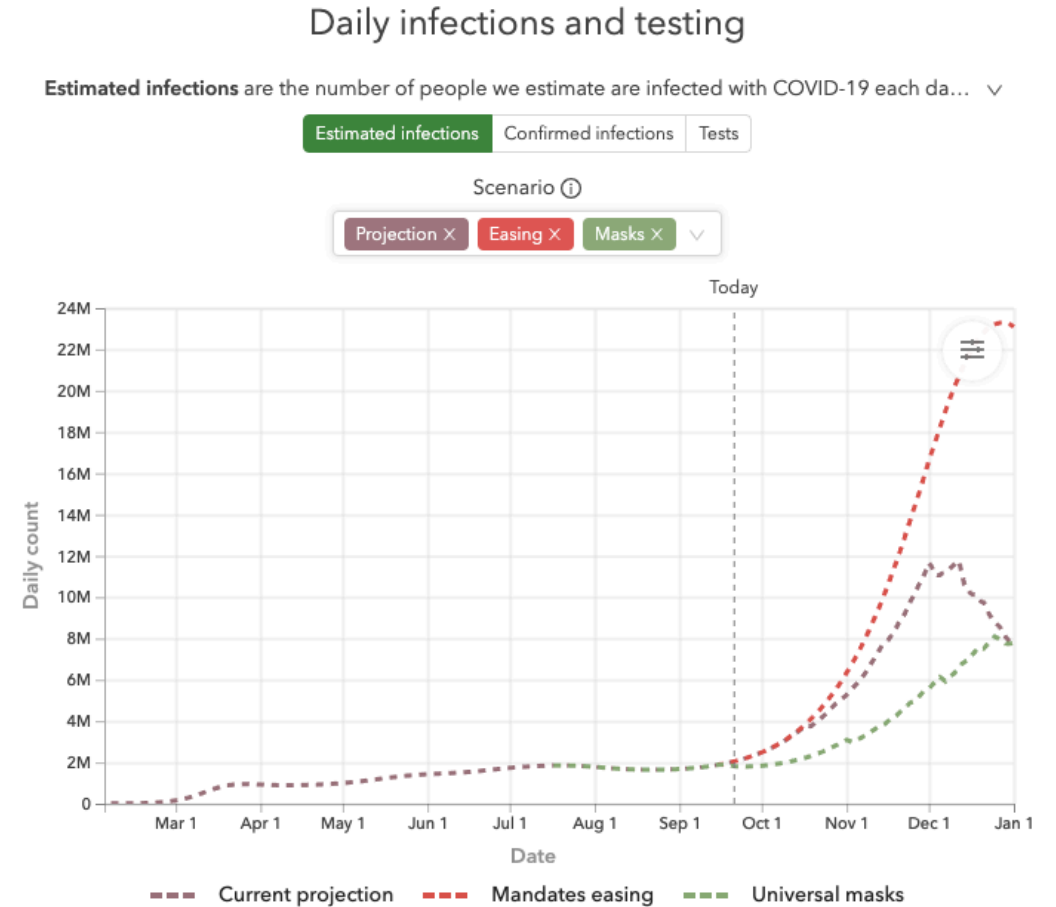
## Stanford Medicine (Sept 9 2020)

<http://med.stanford.edu/news/all-news/2020/09/researchers-test-antibodies-as-covid-19-treatment.html>

## Overview of different testing types

<https://www.globalbiotechinsights.com/articles/20247/the-worldwide-test-for-covid-19>

Compilation of scientific publications on COVID-19 [https://rega.kuleuven.be/if/corona\\_covid-19](https://rega.kuleuven.be/if/corona_covid-19)



# Monty Hall Problem

# Monty Hall Problem and Wayne Brady



# Monty Hall Problem aka Let's Make a Deal

Behind one door is a prize (equally likely to be any door).

Behind the other two doors is nothing

1. We choose a door
2. Host opens 1 of other 2 doors, revealing nothing
3. We are given an option to change to the other door.

Should we switch?

Note: If we don't switch,  $P(\text{win}) = 1/3$  (random)

We are comparing  $P(\text{win})$  and  $P(\text{win} | \text{switch})$ .



Doors A,B,C



(by yourself)

# If we switch

Without loss of generality, say we pick A (out of Doors A,B,C).

1/3

1/3

1/3

A = prize

- Host opens B or C
- We switch
- We always lose

$P(\text{win} \mid \text{A prize, picked A, switched}) = 0$

B = prize

- Host must open C
- We switch to B
- We always win

$P(\text{win} \mid \text{B prize, picked A, switched}) = 1$

C = prize

- Host must open B
- We switch to C
- We always win

$P(\text{win} \mid \text{C prize, picked A, switched}) = 1$

$$P(\text{win} \mid \text{picked A, switched}) = 1/3 * 0 + 1/3 * 1 + 1/3 * 1 = 2/3$$

***You should switch.***

# Monty Hall, 1000 envelope version

Start with 1000 envelopes (of which 1 is the prize).

1. You choose 1 envelope.

$$\left\{ \begin{array}{l} \frac{1}{1000} = P(\text{envelope is prize}) \\ \frac{999}{1000} = P(\text{other 999 envelopes have prize}) \end{array} \right.$$

2. I open 998 of remaining 999 (showing they are empty).

$$\begin{aligned} \frac{999}{1000} &= P(998 \text{ empty envelopes had prize}) \\ &\quad + P(\text{last other envelope has prize}) \\ &= P(\text{last other envelope has prize}) \end{aligned}$$

3. Should you switch?

$$\text{No: } P(\text{win without switching}) = \frac{1}{\text{original \# envelopes}}$$

$$\text{Yes: } P(\text{win with new knowledge}) = \frac{\text{original \# envelopes} - 1}{\text{original \# envelopes}}$$