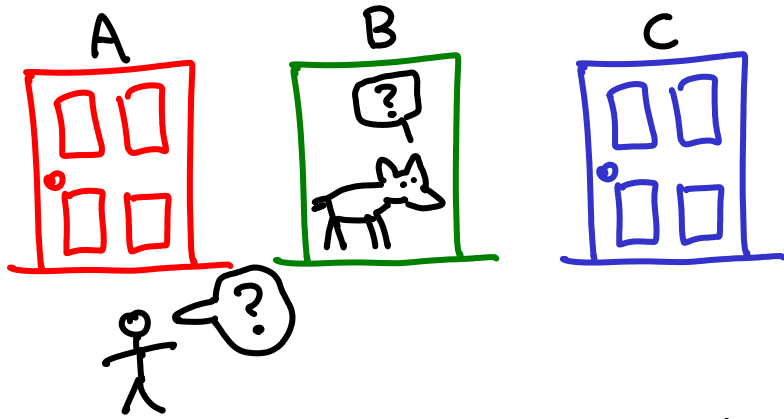


THE MONTY HALL PROBLEM



1 of these 3 doors hides a car.
The other 2 hide goats.

You get to pick a door. You randomly pick **A**.

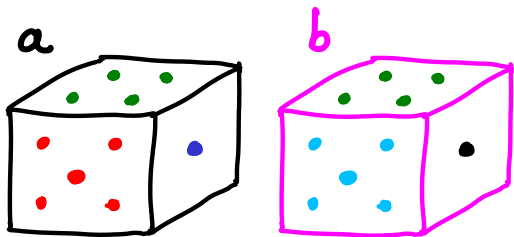
Then a door you didn't pick is opened (say, **B**) revealing a goat.

You're given the choice: **KEEP YOUR DOOR** OR **SWITCH**

?

CONDITIONAL PROBABILITY

Roll 2 dice ...



$$P(A) = P(\text{sum} = 8)$$

$$P(B) = P(\text{both are even})$$

"prob. A given B"

$$P(A|B)$$



If we knew that both are even, then what is the probability that the sum is 8?

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

$$B = \{(2,2), (2,4), (2,6), (4,2), (4,4), (4,6), (6,2), (6,4), (6,6)\}$$

$$P(A|B) = \frac{3}{9}$$

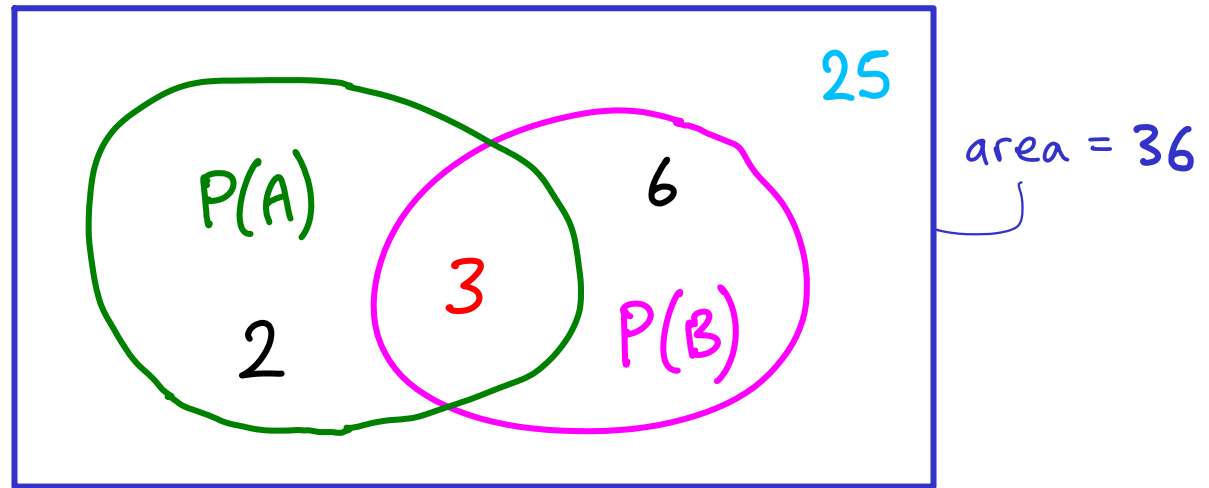
$$\neq P(A)$$

in this example

$$P(A) = P(\text{sum} = 8)$$

$$P(B) = P(\text{both are even})$$

$$P(A) = \frac{\text{green area}}{\text{blue area}} = \frac{5}{36}$$

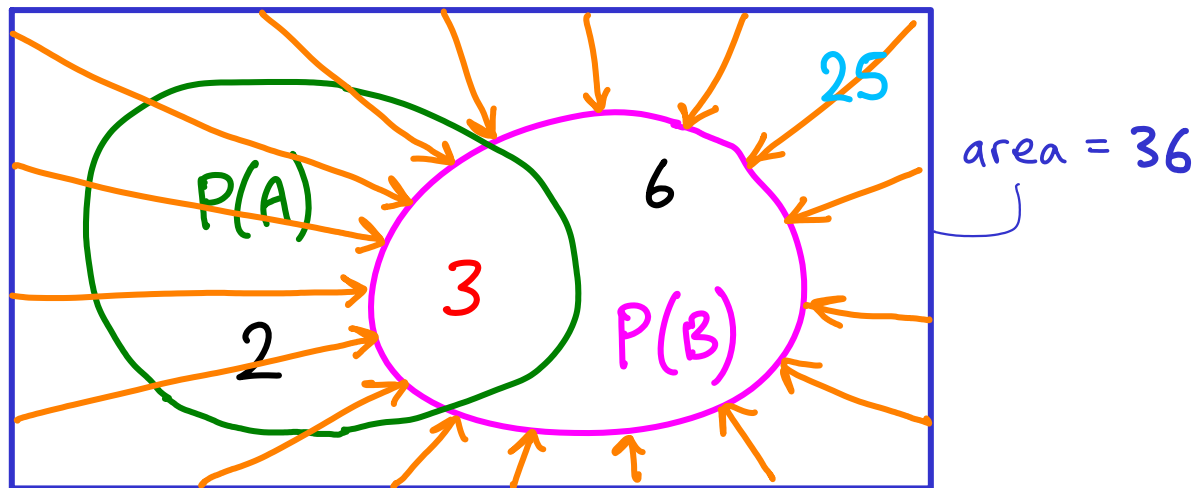


Divide all these numbers by 36, if you prefer the universe to have area 1.

$$P(A) = P(\text{sum} = 8)$$

$$P(B) = P(\text{both are even})$$

$$P(A) = \frac{\text{green area}}{\text{blue area}} = \frac{5}{36}$$



When we establish B then the universe shrinks.

The probability that A holds is normalized: $\frac{\text{remaining valid green area}}{\text{new universe (pink area)}}$

$$P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{3}{3+6}$$

another example Flip a coin 5 times. $P(\text{1st flip} = T) = \frac{1}{2}$

But what if you know that 3 of the 5 flips were H?

$$\hookrightarrow P(\text{1st flip} = T \mid 3 \cdot H) = \frac{P[(\text{1st flip} = T) \cap (3 \cdot H)]}{P(3 \cdot H)}$$

$$P(3 \cdot H) = \frac{\binom{5}{3}}{2^5} \rightarrow \begin{array}{l} \text{Ways to choose} \\ \text{3 positions for H.} \\ \text{Sample space.} \end{array} = \frac{5!}{3!2!} = \frac{5}{16}$$

$$\rightarrow \frac{2/16}{5/16} = \frac{2}{5}$$

$$P[(\text{1st flip} = T) \cap (3 \cdot H)] : \left. \begin{array}{l} T \text{ HHHT} \\ T \text{ HHTH} \\ T \text{ HTHH} \\ T \text{ THHH} \end{array} \right\} \frac{4}{32} \quad \text{OR} \quad \frac{1}{2} \cdot \frac{\binom{4}{3}}{2^4} = \frac{1}{8}$$

another example Flip a coin 5 times. $P(\text{1st flip} = T) = \frac{1}{2}$

But what if you know that 3 of the 5 flips were H?

↪ think of having a biased coin: 60-40 vs 50-50

then the first flip has 40% for T $\rightarrow \frac{2}{5}$

$P(\geq 2 \text{ people in a group of } k \text{ have same birthday})$

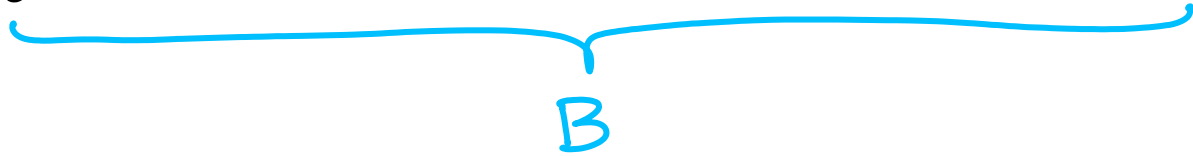
$= 1 - P(\text{all } k \text{ have distinct birthdays})$

$\hookrightarrow P(\text{2nd person has different bday than 1st})$

$$= \frac{364}{365} = P(A)$$

$\cdot P(\text{3rd } \dots \dots \dots \text{1st \& 2nd})$

$$\rightarrow \frac{363}{365} = P(B|A)$$



assuming 1st & 2nd differ

$\cdot P(\text{4th } \dots \dots \dots (1-3))$

$$\rightarrow \frac{362}{365} = P(C | (A \cap B))$$

etc

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

INDEPENDENCE

Flip a coin x3 : $P(3\text{rd} = T \mid 1\text{st} = H) =$

$$= \frac{P[(3\text{rd} = T) \wedge (1\text{st} = H)]}{P(1\text{st} = H)} = \frac{\frac{2}{8}}{\frac{1}{2}} \begin{array}{l} \longrightarrow \# \text{ outcomes} \\ \longrightarrow \text{sample space} \end{array} = \frac{1}{2}$$

Notice $P(3\text{rd} = T) = \frac{1}{2}$ so knowledge of $(1\text{st} = H)$ was useless.

A & B are independent if $P(A) = P(A \mid B)$
if $P(B) = P(B \mid A)$ [equivalent]

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



always



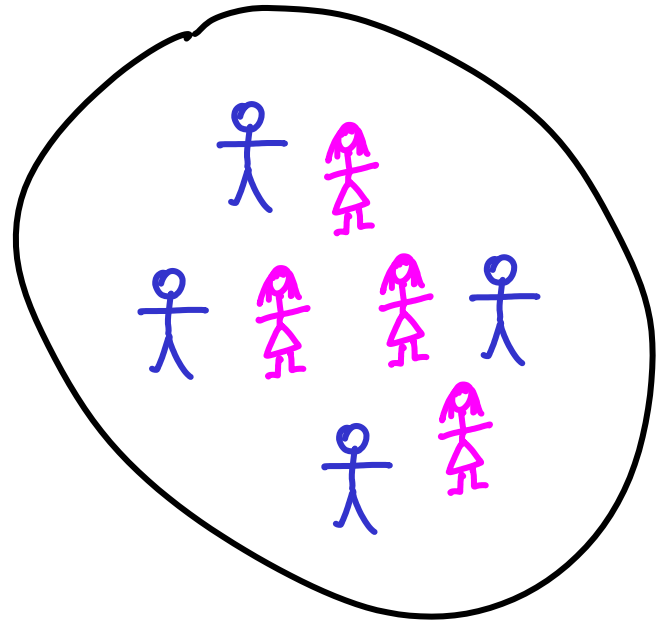
if A & B independent



alternate definition: A & B are independent if

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

another example



4 boys & 4 girls

Select 2 people
from this group

A: 1st person is a girl

B: 2nd person is a girl

dependent

$$P(A) = \frac{4}{8}$$

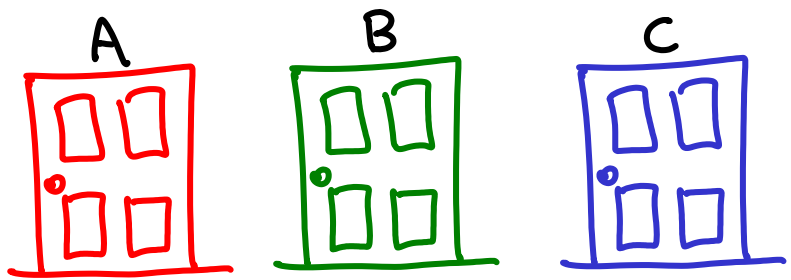
$$P(B) = \frac{1}{2} \text{ (by symmetry)}$$

(or: sample space = 8 · 7
& for each girl = 2nd, #outcomes = 7)

≠

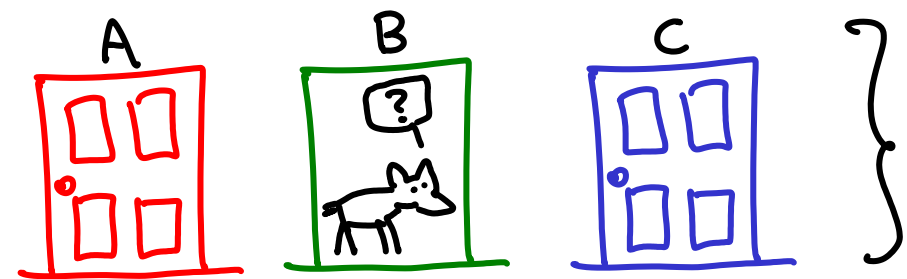
$$P(B|A) = \frac{3}{7}$$

BACK TO MONTY HALL



$$P(\text{car @ A}) = P(A) = \frac{1}{3} = P(B) = P(C)$$

Now we know the car is not at B
What is the probability it's at A?

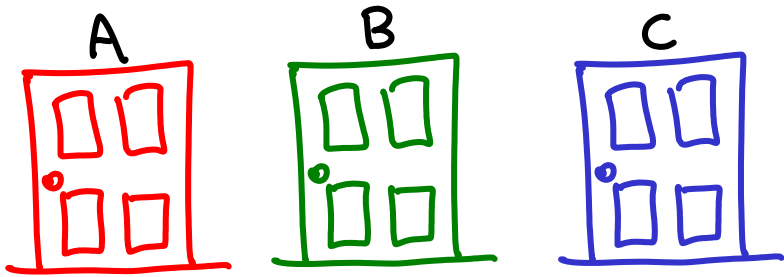


$$P(A | \bar{B}) = \frac{P(A \cap \bar{B})}{P(\bar{B})}$$

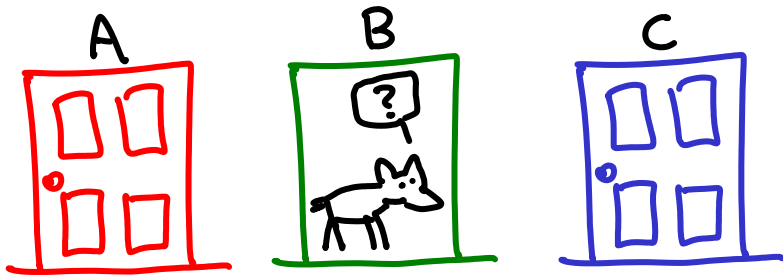
$$\dots \text{because } P(\bar{B} | A) = \frac{P(A \cap \bar{B})}{P(A)} \dots$$

$$= \frac{P(\bar{B} | A) \cdot P(A)}{P(\bar{B})} = \frac{1 \cdot P(A)}{P(\bar{B})} = \frac{\frac{1}{3}}{\frac{2}{3}} = \frac{1}{2}$$

BACK TO MONTY HALL



$$P(\text{car @ A}) = P(A) = \frac{1}{3} = P(B) = P(C)$$



~~$P(A|\bar{B})$~~ ← back up

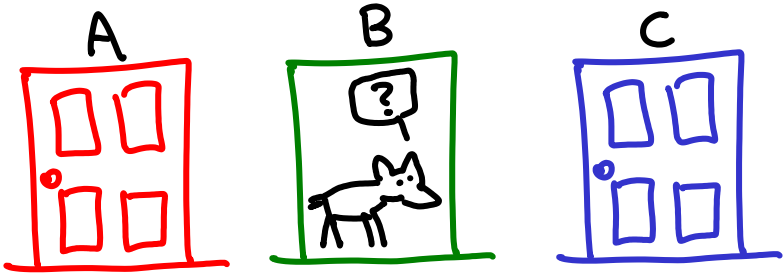
What we actually want is

$$P(A | (\text{door B was opened} \cap \text{we chose A}))$$

↳ not = \bar{B}

↳ extra info

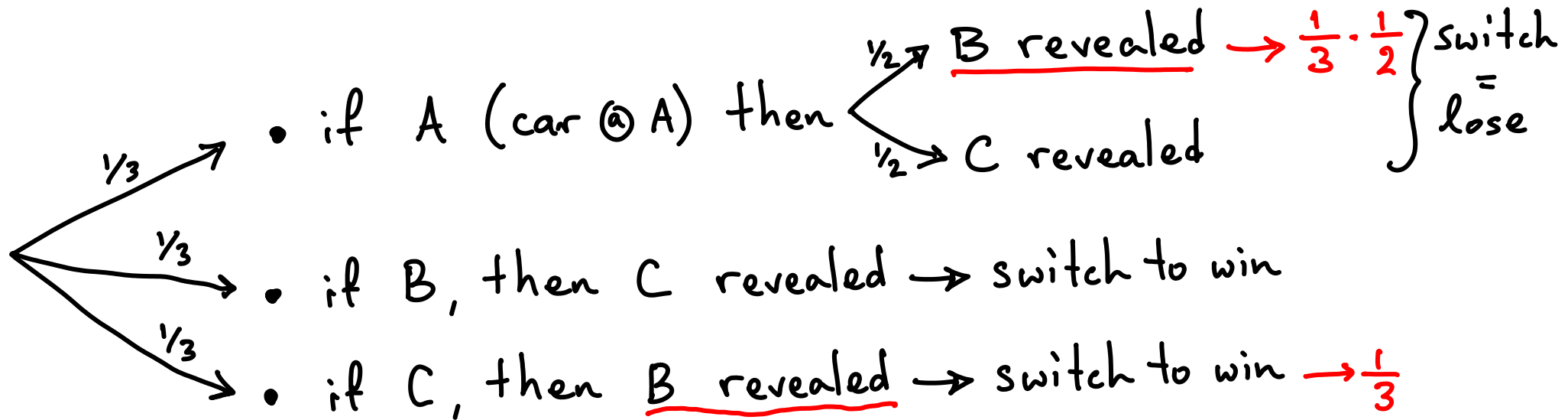
BACK TO MONTY HALL : intuition

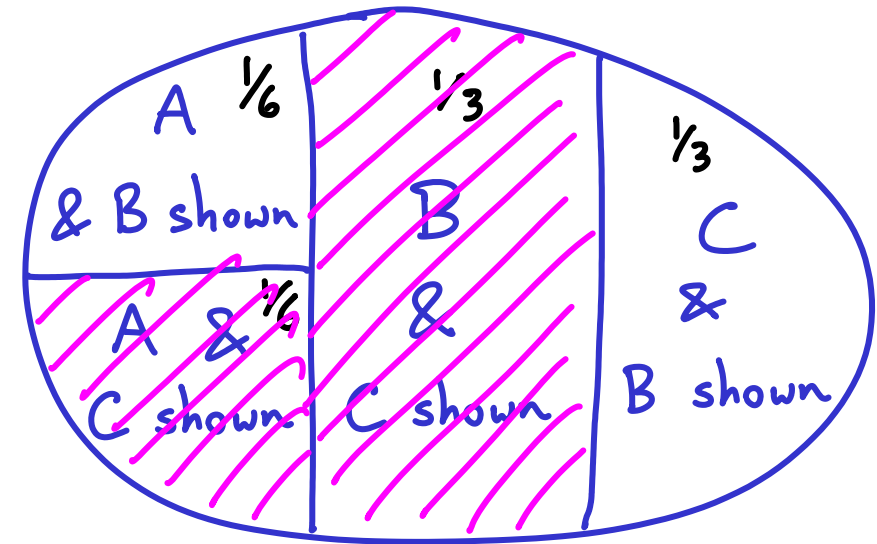
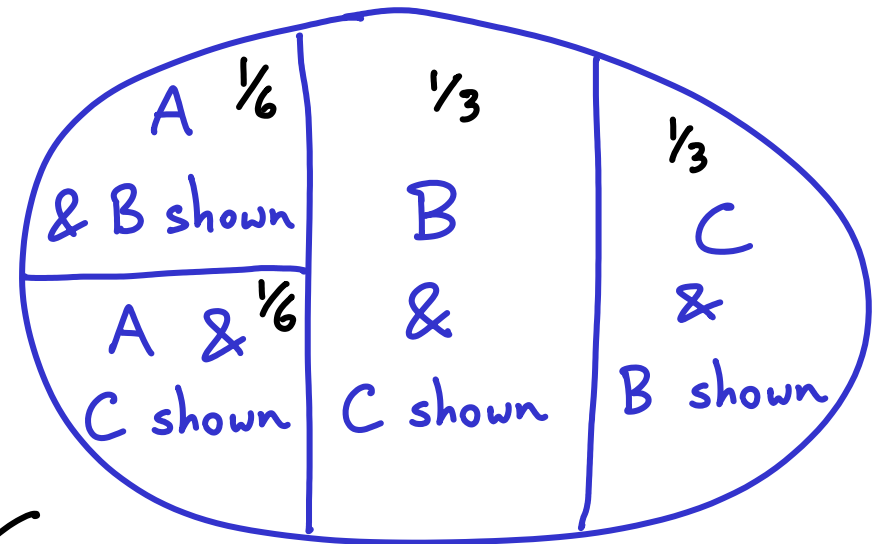
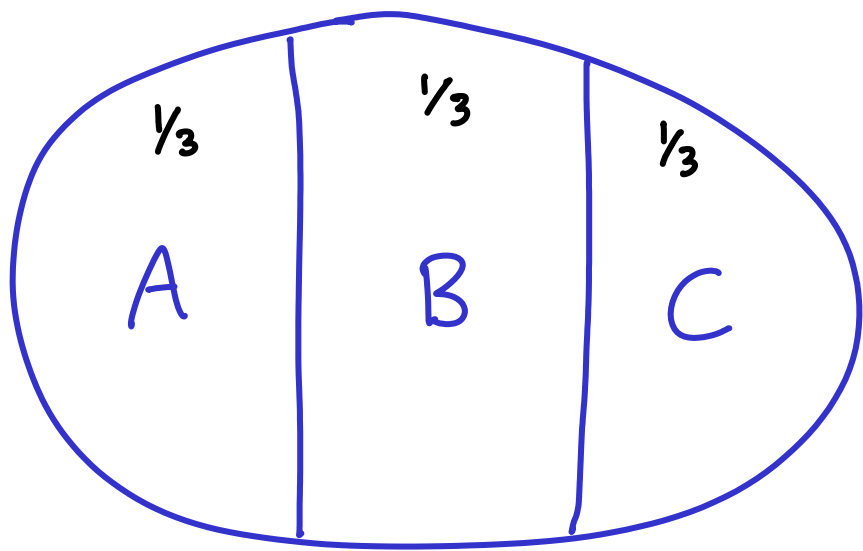


w.l.o.g. guess A
& B is shown.

$$P(\text{car @ A}) = P(A) = \frac{1}{3} = P(B) = P(C)$$

4 events

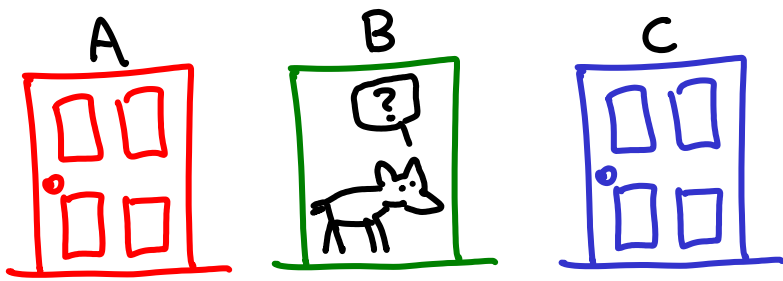




When we establish that B is shown, the universe shrinks.

$$P(A \cap B \text{ shown}) = \frac{\frac{1}{6}}{\frac{1}{6} + \frac{1}{3}} = \frac{1}{3}$$

$$P(C \cap B \text{ shown}) = \frac{\frac{1}{3}}{\frac{1}{6} + \frac{1}{3}} = \frac{2}{3}$$



$$P[A \mid (\text{we chose A} \cap \text{door B was opened})]$$

$$= \frac{P[A \cap (\text{we chose A} \cap \text{door B was opened})]}{P(\text{we chose A} \cap \text{door B was opened})}$$

$$= \frac{P[(A \cap \text{we chose A}) \cap \text{door B was opened}]}{P(\text{we chose A} \cap \text{door B was opened})}$$

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

just moving parentheses

again, for numerator

$$= \frac{P[\text{door B was opened} \mid (A \cap \text{we chose A})] \cdot P(A \cap \text{we chose A})}{P(\text{we chose A} \cap \text{door B was opened})}$$

So far,

$$P[A \mid (\text{we chose A} \cap \text{door B was opened})]$$

$$= \frac{P[\text{door B was opened} \mid (A \cap \text{we chose A})] \cdot P(A \cap \text{we chose A})}{P(\text{we chose A} \cap \text{door B was opened})}$$

$$= \frac{\underbrace{\frac{1}{2}}_{\text{host picks randomly}} \cdot \overbrace{P(A) \cdot P(\text{we chose A})}^{\text{independent}}}{P(\text{we chose A} \cap \text{door B was opened})}$$

$$= \frac{\frac{1}{2} \cdot (\frac{1}{3} \cdot \frac{1}{3})}{P(\text{we chose A} \cap \text{door B was opened})}$$

So far,

The 2 terms in the denominator are "mutually exclusive", which means that they have no intersection. We can see that because one asks for A to happen, but the other asks for C to happen (among other things). In fact we should also include the term +P(we chose A AND door B was opened AND B) but this term is equal to zero. See full notes for a little more discussion on this.

$$P[A | (\text{we chose A} \cap \text{door B was opened})]$$

$$= \frac{1/2 \cdot 1/3 \cdot 1/3}{P(\text{we chose A} \cap \text{door B was opened})}$$

could not have
door B opened
AND
car at B

$$= \frac{1/18}{\left[P(\text{we chose A} \cap \text{door B was opened} \cap A) + P(\text{we chose A} \cap \text{door B was opened} \cap C) \right]}$$

$$\left. \begin{aligned} P(\text{open B} | (C \cap \text{choose A})) &= 1 \\ &= \frac{P(C \cap \text{open B} \cap \text{choose A})}{P(C \cap \text{choose A})} \end{aligned} \right\} =$$

$$\frac{1/18}{P(\text{we chose A} \cap \text{door B was opened} \cap A) + \frac{1}{3} \cdot \frac{1}{3}}$$

So far,

$$P[A \mid (\text{we chose A} \cap \text{door B was opened})]$$

$$= \frac{1/18}{P(\text{we chose A} \cap \text{door B was opened} \cap A) + \frac{1}{3} \cdot \frac{1}{3}}$$

$$= \frac{1/18}{P[(A \cap \text{we chose A}) \cap \text{door B was opened}] + 1/9}$$

$$= \frac{1/18}{\underbrace{P[\text{door B was opened} \mid (A \cap \text{we chose A})]}_{\text{host's random choice}} \cdot \underbrace{P(A \cap \text{we chose A})}_{\text{independent}} + 1/9}$$

$$= \frac{1/18}{\frac{1}{2} \cdot (\frac{1}{3} \cdot \frac{1}{3}) + \frac{1}{9}} = \frac{1/18}{1/18 + 2/18} = \boxed{\frac{1}{3}}$$

TESTING FOR A DISEASE

- Suppose 1% of the population has a disease
- there is a diagnostic test, that finds it 80% of the time
(assuming the subject has it)
- the test also produces false positives, at a rate of 9.6%
(you're fine, but the test says you're not)

If someone tests positive,

what are the odds that they have the disease?

4 events

	1% Have disease	99% Don't have disease
Test ☹	80%	9.6%
Test ☺	20%	90.4%

$$P(\text{disease} | \text{test } \text{☹}) = \frac{P(\text{disease} \cap \text{test } \text{☹})}{P(\text{test } \text{☹})} = \frac{P(\text{test } \text{☹} | \text{disease}) \cdot P(\text{disease})}{P(\text{test } \text{☹})}$$

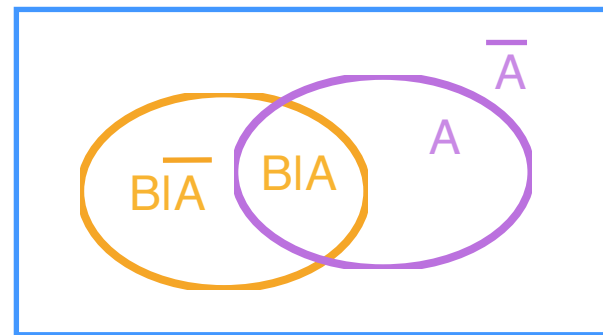
$$P(\text{test } \text{☹}) = \begin{cases} P(\text{test } \text{☹} | \text{disease}) \cdot P(\text{disease}) = 0.8 \cdot 0.01 = 0.008 \text{ (0.8\%)} \\ + \\ P(\text{test } \text{☹} | \text{no disease}) \cdot P(\text{no disease}) = 0.096 \cdot 0.99 \approx 0.095 \text{ (9.5\%)} \end{cases}$$

$$P(\text{disease} | \text{test } \text{☹}) = \frac{0.008}{0.008 + 0.095} \sim 7.8\%$$

Bayes theorem

$$\begin{aligned}P(A \cap B) &= P(A|B) \cdot P(B) \\ &= P(B|A) \cdot P(A)\end{aligned}$$

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



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

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

A: have disease

B: test :)

1% A	99% \bar{A}
B A = 80%	B \bar{A} = 9.6%

$$P(A) = 0.01$$
$$P(\bar{A}) = 0.99$$

$$P(B|A) = 0.8$$

$$P(B|\bar{A}) = 0.096$$

$$P(A|B) = \frac{0.8 \cdot 0.01}{0.8 \cdot 0.01 + 0.096 \cdot 0.99}$$