

1.2 Conditional Probability

last week

Recap —

- $0 \leq P(A) \leq 1$ for each event $A \subseteq S$
- $P(A \cup B) = P(A) + P(B) - P(A \cap B)$ for any events A, B
- If A and B are mutually exclusive then $A \cap B = \emptyset$
- $P(\bar{A}) = 1 - P(A)$ for any event A

Example 1.8

- If a single fair die is tossed, then

$$P(\text{face 2 turns up}) = \frac{1}{6}$$

- If a single fair die is tossed, and it is known that the face that turns up is an even number, then

$$P(\text{face 2 turns up}) = \frac{1}{3}$$

- If a single fair die is tossed, and it is known that the face that turns up is an odd number, then

$$P(\text{face 2 turns up}) = \frac{0}{3} = 0$$

□

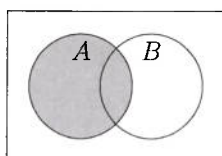
Notes —

- The probability of an event is **conditioned** by what other events we know to have occurred.
- The **sample space** that a probability is calculated with respect to has a great bearing upon the probability.

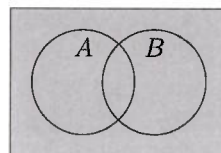
The **probability** of event A is

$$P(A) = \frac{\text{number of outcomes in } A}{\text{number of outcomes in } S}$$

} assume
equally
likely
outcomes



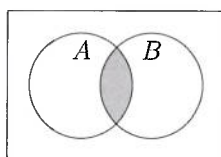
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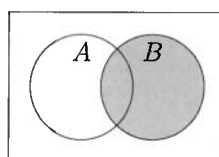
The **probability** of event A **conditional** on knowing that event B has occurred is

$$P(A|B) = \frac{\text{number of outcomes in } A \cap B}{\text{number of outcomes in } B}$$

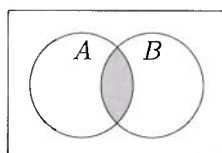
IDEA:
→ event is $A \cap B$
→ sample space
is B



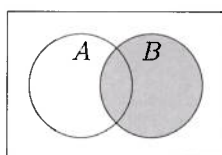
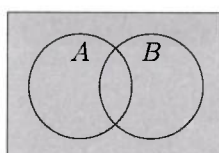
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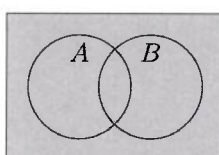
which is equivalent to



divided by



divided by



In general %

definition

■ The **conditional probability of A given B** is

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

event of
interest

1020

has already occurred
(additional information)

Example 1.9

An electronic display is equally likely to show any of the digits 1,2,3,4,5,6,7,8,9.

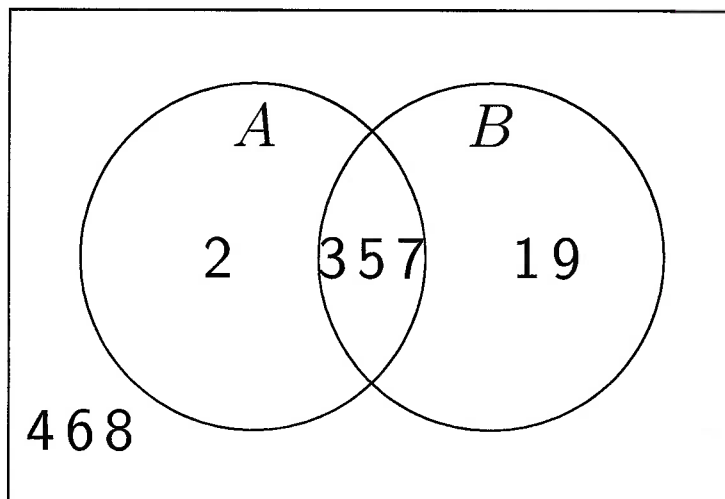
Determine the probability that it shows a prime number (i.e. one of 2, 3, 5 and 7):

- (i) given no knowledge about the number
- (ii) given the information that the number is odd

Solution.

Let A be the event "a prime number".

Let B be the event "an odd number".



$$(i) \quad P(A) = \frac{|A|}{|S|} = \frac{4}{9} = P(\text{prime})$$

$$(ii) \quad P(A|B) = \frac{|A \cap B|}{|B|} = \frac{3}{5} = P(\text{prime} | \text{odd})$$

Note : $P(\text{prime} \cap \text{odd}) = \frac{3}{9}$ □

1021 \uparrow AND

Addition rule (always):

$$P(A \cup B) = P(A) + P(B) - P(A \cap B)$$

1.2.1 Independent Events

Two events are **independent** if the occurrence of either event does not change the probability of the other event occurring.

definition

■ Events A and B are **independent** if

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

(or equivalently)

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

Notes —

- Physically independent events are always statistically independent.

- Since $P(A | B) = \frac{P(A \cap B)}{P(B)}$ we have \rightarrow definition of conditional probability

$$P(A \cap B) = P(A | B) \times P(B) \rightarrow \text{often useful in general case}$$

- If A and B are independent then $P(A | B) = P(A)$ so ...

■ *Multiplication Law of Probability*

If A and B are independent events then

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

Practice Problem. Two events A and B are such that

$$P(A) = 0.5 \quad P(B) = 0.4 \quad P(A|B) = 0.3$$

(a) Are A and B independent?

If independent then equal

No since $P(A) \neq P(A|B)$

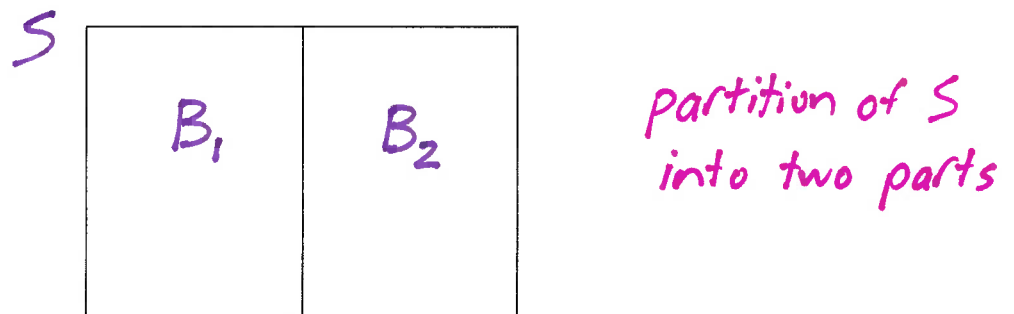
(b) Find the value of $P(A \cap B)$.

$$\begin{aligned} P(A \cap B) &= P(A|B) \times P(B) \\ &= 0.3 \times 0.4 \\ &= 0.12 \quad \rightarrow A \cap B \neq \emptyset \end{aligned}$$

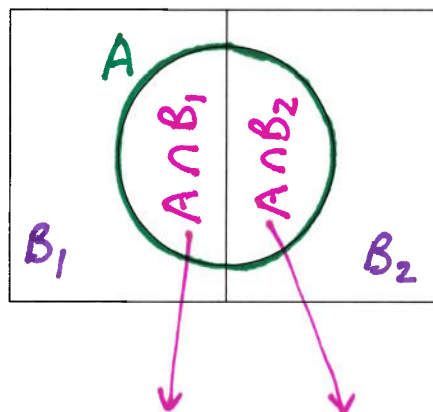
Note : $P(A) \times P(B) = 0.5 \times 0.4 = 0.2$ □

1.2.2 Law of Total Probability

Suppose $S = B_1 \cup B_2$ where $B_1 \cap B_2 = \emptyset$
i.e., B_1 and B_2 are mutually exclusive and exhaustive (see page 1012).



Now consider event A .



It is clear that

$$A = (A \cap B_1) \cup (A \cap B_2)$$

and that

$$(A \cap B_1) \cap (A \cap B_2) = \emptyset$$

i.e., A can be expressed as the union of two mutually exclusive events $(A \cap B_1)$ and $(A \cap B_2)$.

Therefore (using the addition law)

$$\begin{aligned} P(A) &= P(A \cap B_1) + P(A \cap B_2) \\ &= P(A | B_1) \times P(B_1) + P(A | B_2) \times P(B_2) \end{aligned}$$

$$P(\text{do well } M \mid \text{do well } P) = 0.8$$

Example 1.10

- Of those students who do well in Physics, 80% also do well in Mathematics.
- Of those students who do not do well in Physics, only 30% do well in Mathematics.
- If 40% of students do well in Physics, what proportion do well in Mathematics?

$$P(\text{do well } M \mid \text{do not well } P) = 0.3$$

$$P(\text{do well } P) = 0.4$$

Solution.

Let A be the event "does well in Mathematics".

Let B_1 be the event "does well in Physics".

Let B_2 be the event "does not do well in Physics".

} formalise

B_1 and B_2 are mutually exclusive and exhaustive events

So
$$P(M) = P(M|P) \times P(P) + P(M|\bar{P}) \times P(\bar{P})$$

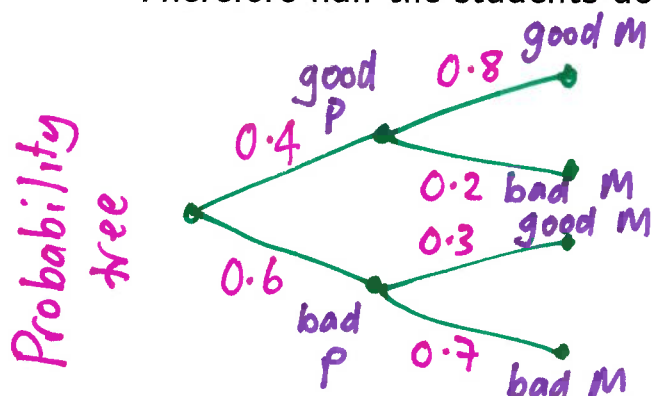
$$P(A) = P(A|B_1) \times P(B_1) + P(A|B_2) \times P(B_2)$$

$$= 0.8 \times 0.4 + 0.3 \times 0.6$$

$$= 0.32 + 0.18$$

$$= 0.5$$

Therefore half the students do well in Mathematics.



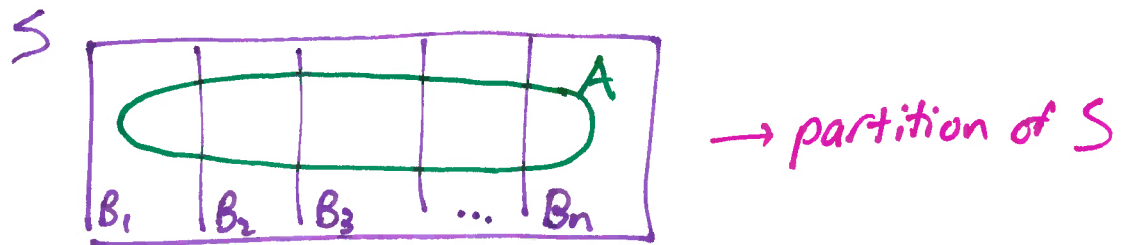
$$P(\text{good } P \cap \text{good } M) = 0.4 \times 0.8 = \boxed{0.32}$$

$$P(\text{good } P \cap \text{bad } M) = 0.4 \times 0.2 = 0.08$$

$$P(\text{bad } P \cap \text{good } M) = 0.6 \times 0.3 = \boxed{0.18}$$

$$P(\text{bad } P \cap \text{bad } M) = 0.6 \times 0.7 = 0.42$$

$$P(\text{good } M) = 0.32 + 0.18 = \boxed{0.50}$$



■ Law of Total Probability

Suppose that B_1, B_2, \dots, B_n are mutually exclusive and exhaustive events, i.e.,

- $B_i \cap B_j = \emptyset$ for every i and j where $i \neq j$
- $B_1 \cup B_2 \cup \dots \cup B_n = S$

Then

$$\begin{aligned}
 P(A) &= P(A \cap B_1) + P(A \cap B_2) + \dots + P(A \cap B_n) \\
 &= P(A | B_1) \times P(B_1) \\
 &\quad + P(A | B_2) \times P(B_2) \\
 &\quad + \dots \\
 &\quad + P(A | B_n) \times P(B_n)
 \end{aligned}$$

Horrific! → use probability tree instead

→ multiply probabilities
along the branches

1.2.3 Tree Diagrams

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

Since $B \cap A = A \cap B$ this can be rearranged as

$$P(A \cap B) = P(A) \times P(B|A) \rightarrow \text{"chain" rule}$$

Repeating this idea gives the **chain rule** for conditional probabilities

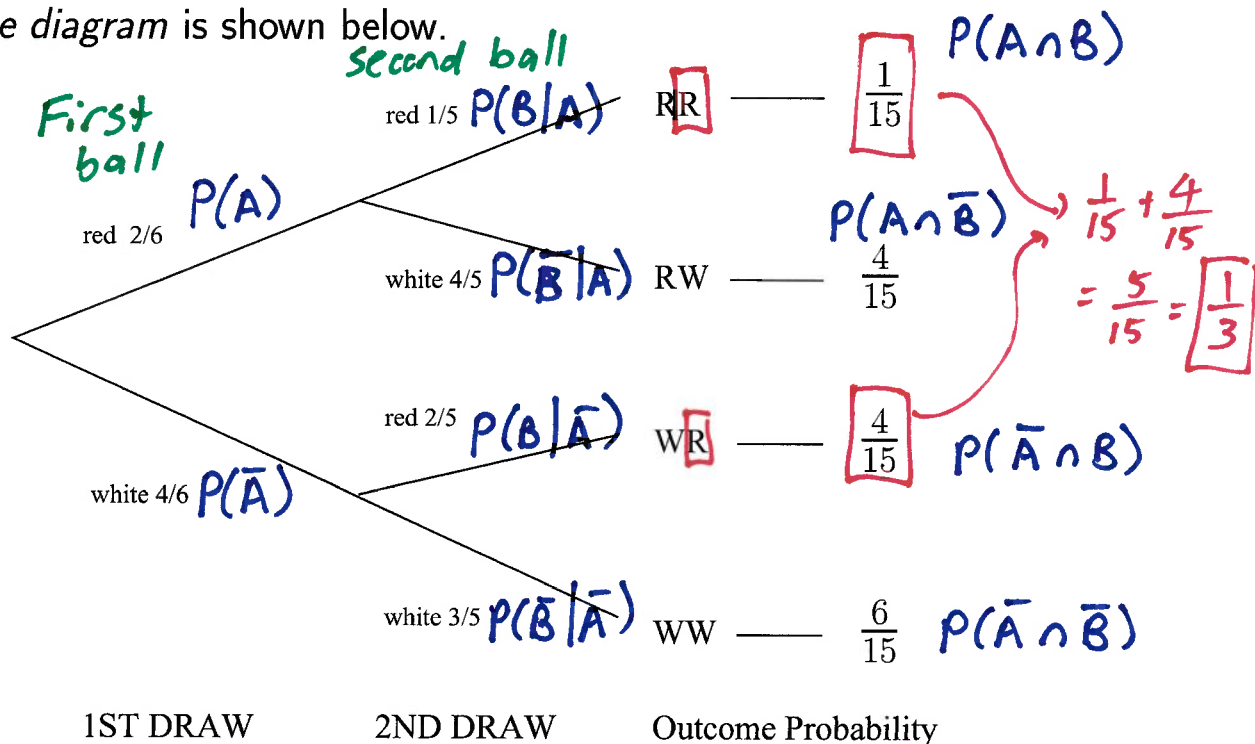
$$P(A \cap B \cap C) = P(A) \times P(B|A) \times P(C|A \cap B)$$

Example 1.11 A bucket contains 2 red balls and 4 white balls. Two balls are drawn in sequence, *without replacement*. What is the probability that the second ball drawn is red?

$A = \text{First ball red.}$
 $B = \text{Second ball red.}$

Solution.

A tree diagram is shown below.



pr003

- The probabilities along the branches are the conditional probabilities for the next stage, given the results of the previous stages.
- Multiplying the probabilities along each branch gives

(by the **chain rule**)

the probability of the outcome that branch represents.

$$P(RR) = \frac{2}{6} \times \frac{1}{5} = \frac{1}{15}$$

$$P(RW) = \frac{2}{6} \times \frac{4}{5} = \frac{4}{15}$$

$$P(WR) = \frac{4}{6} \times \frac{2}{5} = \frac{4}{15}$$

$$P(WW) = \frac{4}{6} \times \frac{3}{5} = \frac{6}{15}$$

- Check that the total probability is 1, i.e.,

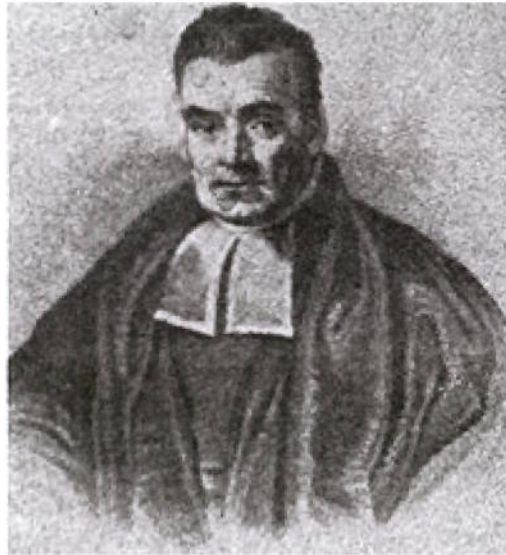
$$\frac{1}{15} + \frac{4}{15} + \frac{4}{15} + \frac{6}{15} = \frac{1+4+4+6}{15} = \frac{15}{15} = 1$$

- Then add the probabilities contributing to the event of interest, i.e.,

$$\begin{aligned} P(\text{second ball is red}) &= P(RR) + P(WR) \\ &= \frac{2}{6} \times \frac{1}{5} + \frac{4}{6} \times \frac{2}{5} \\ &= \frac{1}{15} + \frac{4}{15} = \frac{5}{15} = \boxed{\frac{1}{3}} \end{aligned}$$

□

1.2.4 Bayes' Theorem



Thomas Bayes (1701–1761)

18th century British clergyman and mathematician

<http://www.bayesian.org/bayesian/bayes.html>

Idea of conditional probability —

Given that event B has happened in the past, what is the probability that event A will occur?

$$P(A | B)$$

$$P(\text{wet} | \text{forecast rain})$$

Reverse question —

Given that event A has just occurred, what is the probability that it was preceded by the event B ?

$$P(B | A)$$

$$P(\text{forecast rain} | \text{wet})$$

definition of conditional probability

- Since $P(A|B) = \frac{P(A \cap B)}{P(B)}$ we have

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

- Since $P(B|A) = \frac{P(B \cap A)}{P(A)}$ we have

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

same

- But $P(A \cap B) = P(B \cap A)$ so we have

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

■ Bayes' Theorem

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

Sometimes called "Bayesian inversion"

Terminology (in machine learning) —

- $P(B)$ is known as the **prior** or **hypothesis** (before evidence is seen)
- $P(A)$ is known as the **evidence**
- $P(A|B)$ is known as the **likelihood** (indicates the compatibility of the evidence with the hypothesis)
- $P(B|A)$ is known as the **posterior** (after evidence is seen)

used in Forensic Science
(and legal cases)

Example 1.12

H or T

H or H

An engineer has a fair coin and a double-headed coin. She chooses one of the coins at random and tosses it. She obtains a head. Determine the probability that the coin that she tossed was double-headed.

Solution.

Let A be the event that "a head is obtained".

Let B_1 be the event that "the fair coin was chosen".

Let B_2 be the event that "the double-headed coin was chosen".

B_1 and B_2 are mutually exclusive and exhaustive events.

We know

$$P(B_1) = \frac{1}{2} \quad P(B_2) = \frac{1}{2} \quad P(A | B_1) = \frac{1}{2} \quad P(A | B_2) = 1$$

$$\begin{aligned} P(A) &= P(A | B_1) \times P(B_1) + P(A | B_2) \times P(B_2) \\ &= \frac{1}{2} \times \frac{1}{2} + 1 \times \frac{1}{2} = \frac{3}{4} \end{aligned}$$

$$P(\text{coin was double-headed}) = P(B_2 | A)$$

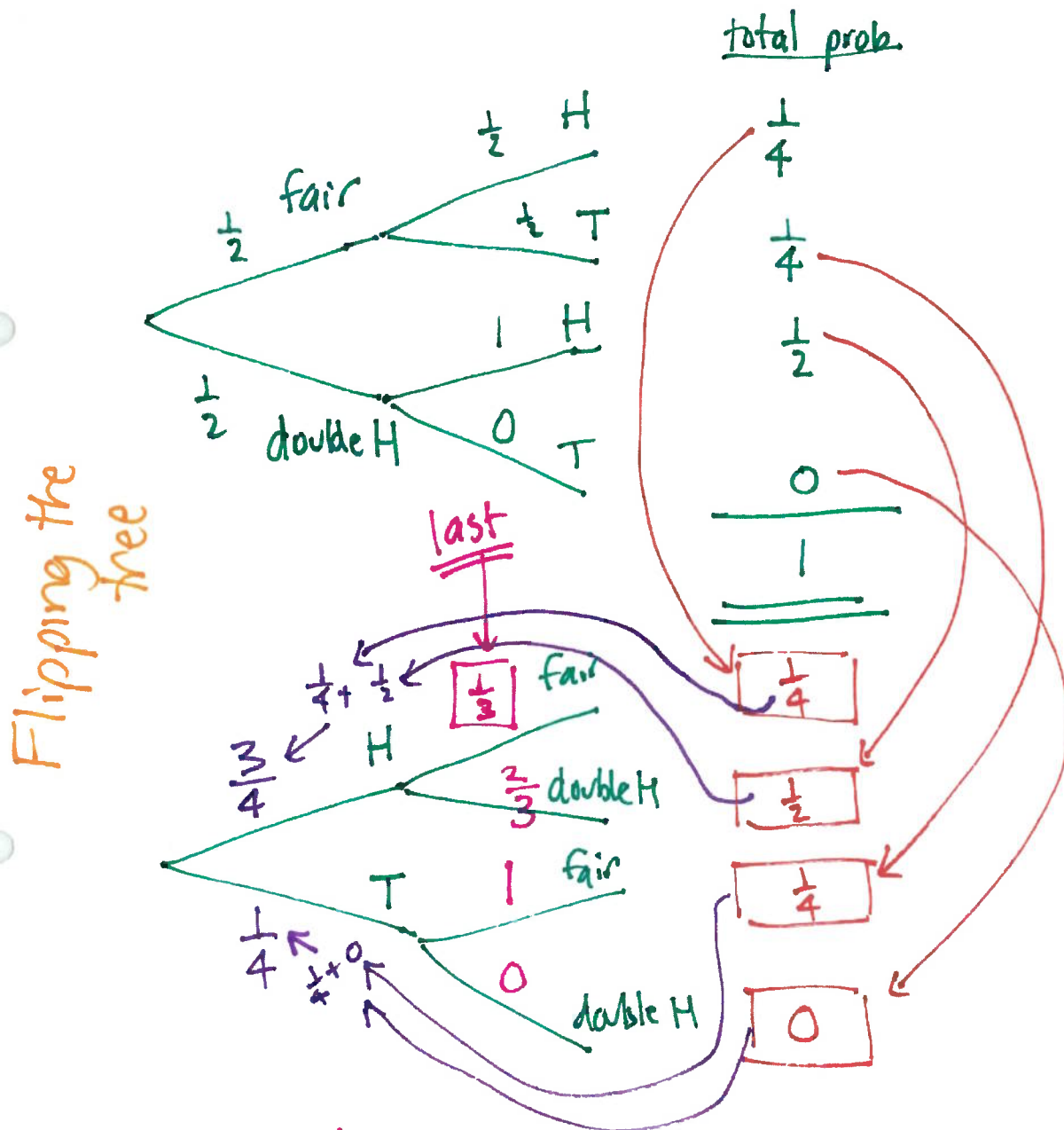
$$= \frac{P(A | B_2) \times P(B_2)}{P(A)}$$

$$= \frac{1 \times \frac{1}{2}}{\frac{3}{4}} = \frac{2}{3}$$

Bayes' theorem

► A good thing about Bayes' Theorem is that, once the events have been carefully defined, we do not need to think!

- We can often perform all the calculations on two probability trees.



$$P(\text{fair} | H) = \frac{1}{3}$$

$$P(\text{double H} | H) = \frac{2}{3} \leftarrow \text{solution to the problem}$$

$$P(\text{fair} | T) = 1$$

$$P(\text{double H} | T) = 0$$

Summary

- The *conditional probability* of event A , given that event B is known to have occurred, is

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

then we can also write

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

- Events A and B are *independent* if $P(A|B) = P(A)$
(or equivalently) $P(B|A) = P(B)$
(or equivalently) $P(A \cap B) = P(A) \times P(B)$

- *Law of total probability* —

If B_1, B_2, \dots, B_n are mutually exclusive and exhaustive then

$$\begin{aligned} P(A) &= P(A \cap B_1) + P(A \cap B_2) + \dots + P(A \cap B_n) \\ &= P(A|B_1)P(B_1) + P(A|B_2)P(B_2) \\ &\quad + \dots + P(A|B_n)P(B_n) \end{aligned}$$

- *Chain rule* — multiply probability along the branches in tree diagrams

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

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

- *Bayes' Theorem*

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

- Use probability tree wherever possible.
- "Flipping the tree."

