

2.4 Conditional Probability

So far, we've looked at events occurring in a vacuum (totally on their own). What if we know more info? What if another event could influence our event?

Monty Hall problem / Let's Make a Deal / 21

3 doors : 2 goats
 1 sports car

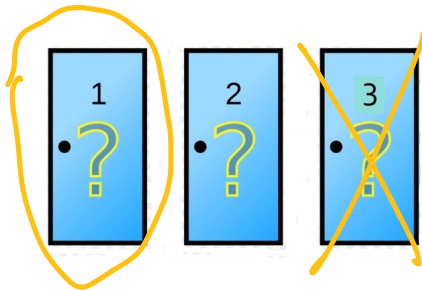
Equally likely to start:

$$P(D_1) = P(D_2) = P(D_3) = \frac{1}{3}.$$

Say you pick D_1 . The host then reveals a goat behind D_3 . Do you stick with D_1 or do you switch to D_2 ?

You may think there's now a $\frac{1}{2}$ chance the car is behind either D_1 or D_2 .

But this new information has changed the probability of switching leading to a car! (You should switch, and we'll find out why...)



Def. Let A, B be events on sample space S with $P(B) > 0$. The conditional probability of A given B is

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

Note that the formula does not work if $P(B) = 0$.
 We can't draw any conclusions about an event A based on an event B with no chance of occurring!

Back to Monty Hall...
 You chose D_1 .

If $D_1 = \text{car}$, host opens either D_2 or D_3 .

$$P(D_1) = \frac{1}{3}$$

half the time host opens $D_2 \rightarrow \frac{1}{6}$
 " " " " " $D_3 \rightarrow \frac{1}{6}$

Prize door, opened door	Probability
D_1, D_2	$\frac{1}{6}$
D_1, D_3	$\frac{1}{6}$
D_2, D_3	$\frac{1}{3}$
D_3, D_2	$\frac{1}{3}$

If $D_2 = \text{car}$, host must open $D_3 \rightarrow \frac{1}{3}$

" D_3 " " " " " $D_2 \rightarrow \frac{1}{3}$

let $A = \text{event prize is behind } D_1 \text{ (your door)}$

$B = \text{event host opens } D_3$

$$P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{\frac{1}{6}}{\frac{1}{2}} = \frac{1}{3} \quad \text{if you don't switch}$$

Let C = event prize is behind D_2 (not your door!)

$$P(C|B) = \frac{P(C \cap B)}{P(B)} = \frac{\frac{1}{3}}{\frac{1}{2}} = \frac{2}{3} \text{ if you switch!!}$$

The additional info from the host opening the door means we should change our guess!

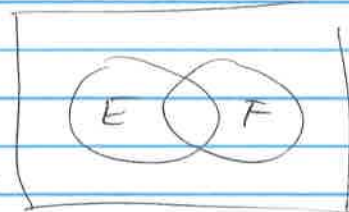
Sometimes we will use the conditional probability to compute the probability of an intersection. (sometimes $P(A|B)$ is given or easier to compute than $P(A \cap B)$).

3.3 Bayes's Formula

Let E and F be events. Then

$$E = EF \cup EF^c$$

Since EF and EF^c are mutually exclusive events



$$P(E) = P(EF) + P(EF^c) \quad \leftarrow \text{by Axiom 3}$$

$$= P(E|F) \cdot P(F) + P(E|F^c) P(F^c)$$

but $P(F^c) + P(F) = 1 \Rightarrow$

$$P(E) = P(E|F) P(F) + P(E|F^c) (1 - P(F))$$

EX

Recall Celine.

$$P(F) = 1/2$$

$$P(C) = 1/2$$

probability she takes French
probability she takes Chem.
each flip

$$P(A|C) = \frac{2}{3}$$

← given she takes Chem
probability of an A

$$P(A|F) = 1/2$$

← given she takes French
probability of an A.

Suppose after the semester she tells you she got an A.

What is the probability that she took chemistry?

$$= P(C|A) = \text{probability she took Chem given she got an A} \\ (\text{different from } P(A|C)).$$

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

(67.1)

From

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

True statements that follow are...

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

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

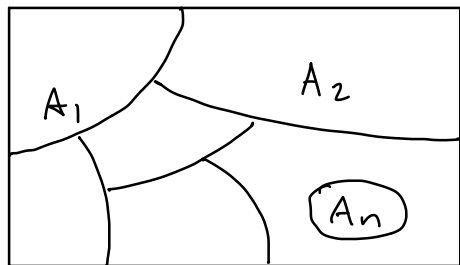
Also

$$P(F_c|E) = \frac{P(EF_c)}{P(E)}$$

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

(these are the "n=2" version of Bayes's Theorem

Let's think about more than two events. We'll think about a partition to our sample space S :



The events A_1, A_2, \dots, A_n partition S because every outcome in the sample space belongs to one and only one event A_i .

A partition is a mutually exclusive or disjoint collection of sets that together make up S .

Two theorems: 1 for unconditional probabilities,
1 for conditional probabilities.

Law of Total Probability (Thm 2.4.1 in text)

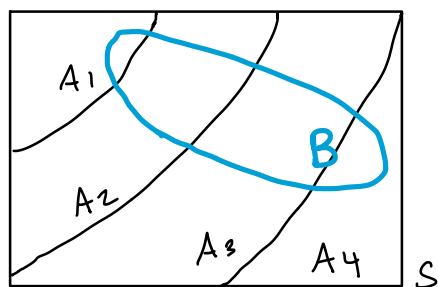
If A_1, A_2, \dots, A_n are disjoint events and

$$S = \bigcup_{i=1}^n A_i \quad (\text{the } A_i\text{'s together make up } S),$$

then for any event B ,

$$P(B) = \sum_{i=1}^n P(B|A_i) P(A_i).$$

Let's do a proof by picture when $n=4$.



How can we write B as the union of disjoint events?

$$B = (B \cap A_1) \cup (B \cap A_2) \cup (B \cap A_3) \cup (B \cap A_4)$$

↑ ↑ ↑ ↑
disjoint

Since A_1, A_2, A_3, A_4 form a partition of S .

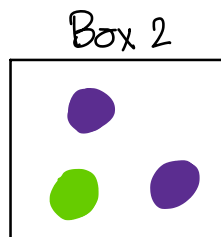
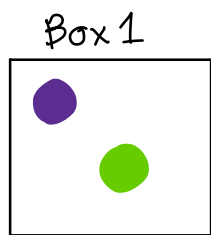
Then $P(B) = P(B \cap A_1) + P(B \cap A_2) + P(B \cap A_3) + P(B \cap A_4)$

by countable additivity (axiom 3).

Using $P(A|C) = \frac{P(A \cap C)}{P(C)} \Rightarrow P(A \cap C) = P(C)P(A|C)$

$$P(B) = P(A_1)P(B|A_1) + P(A_2)P(B|A_2) + P(A_3)P(B|A_3) + P(A_4)P(B|A_4).$$

Ex Consider an experiment with 2 boxes with marbles. You toss a coin: heads \Rightarrow select from Box 1
tails \Rightarrow select from Box 2



Let R be the event a purple marble is selected.

Let $B_i =$ event Box i is selected
because these are determined by a coin flip,

$$P(B_1) = P(B_2) = \frac{1}{2}.$$

Conditional probabilities are straightforward:

$$P(R|B_1) = \frac{1}{2}, \quad P(R|B_2) = \frac{2}{3}.$$

Note that B_1 and B_2 partition the sample space and are disjoint. Thus,

$$\begin{aligned} P(R) &= P(B_1) P(R|B_1) + P(B_2) P(R|B_2) \\ &= \left(\frac{1}{2}\right)\left(\frac{1}{2}\right) + \left(\frac{1}{2}\right)\left(\frac{2}{3}\right) = \frac{1}{4} + \frac{1}{3} = \frac{7}{12}. \end{aligned}$$

Now let's look at a theorem that allows us to switch the conditioning of events, ex. from $P(A|B)$, find $P(B|A)$.

Bayes Rule (Thm. 2.4.2 in text)

Let A_1, A_2, \dots, A_n be disjoint and partition the sample space S (also referred to as exhaustive.)

Let B be any event. Then for any A_i ,

$$P(A_i|B) = \frac{P(A_i) P(B|A_i)}{\underbrace{P(B|A_1)P(A_1) + P(B|A_2)P(A_2) + \dots + P(B|A_n)P(A_n)}_{P(B) \text{ from law of total probability}}}$$

so this reduces to

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

Ex: Dr. Bedekar loves to cook but sometimes she is too busy.

If there is traffic due to construction, there is a 50% chance she will order takeout.

If there is traffic not due to construction, there's a 30% chance she will order takeout.

If there is no traffic, there's a 10% chance she will order takeout.

On any given day, assume there is a 60% chance of no traffic, 20% chance of traffic, and a 20% chance of traffic from construction.

If Dr. Bedekar orders takeout, what's the probability

(a) there is traffic due to construction? $P(A_1|B)$

(b) traffic not due to construction? $P(A_2|B)$

(c) no traffic? $P(A_3|B)$

Let B = event she gets takeout.

A_1 = event there is traffic (C)

A_2 = " " " " (no C)

A_3 = " " " no traffic.

We have $P(A_1) = 0.2$, $P(A_2) = 0.2$, $P(A_3) = 0.6$

and

$P(B|A_1) = 0.5$, $P(B|A_2) = 0.3$, $P(B|A_3) = 0.1$

Using Bayes' rule:

$$P(A_1|B) = \frac{P(B|A_1)P(A_1)}{\sum_{i=1}^3 P(B|A_i)P(A_i)}$$
$$= \frac{(0.5)(0.2)}{(0.5)(0.2) + (0.3)(0.2) + (0.1)(0.6)} = \frac{0.1}{0.22} \approx 45.5\%$$

Check :

$$P(A_2|B) \approx 27.3\%, \quad P(A_3|B) \approx 27.3\%$$

Did not do any of the examples after this—all fair game!

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$$P(C|A) = \frac{P(A|C)P(C)}{P(A|C)P(C) + P(A|C^c)P(C^c)}$$

note $C^c = F$

so

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

$$= \frac{(\frac{2}{3})(\frac{1}{2})}{(\frac{2}{3})(\frac{1}{2}) + (\frac{1}{2})(\frac{1}{2})} = \frac{\frac{2}{3}}{\frac{2}{3} + \frac{1}{2}} = \frac{1}{1 + \frac{3}{4}} = \frac{4}{7}$$

probability she took chemistry

Ch. 3 Problem 3.18

46% of voters in a city classify themselves as independents.

30% of liberals

24% " " " " conservatives.

In a recent election...

35% of independents voted

62% of liberals voted

58% of conservatives voted.

A voter is chosen at random. Given that this person voted...

What is the probability that he/she is

a) independent? b) liberal? c) conservative?

d) What percentage of voters participated in the election?

a) ans. $P(I|V)$ = probability of I = indep. given that they voted, V (69)

$$P(I|V) = \frac{P(IV)}{P(V)}$$

$$P(V) = P(IV) + P(LV) + P(CV)$$

$$(\text{from } V = IV \cup LV \cup CV)$$

union of mutually exclusive events

$$\text{So } P(I|V) = \frac{P(IV)}{P(IV) + P(LV) + P(CV)}$$

$$P(IV) = P(V|I) P(I) = (.35)(.46) = 0.161$$

$$P(LV) = P(V|L) P(L) = (.62)(.3) = 0.186$$

$$P(CV) = P(V|C) P(C) = (.58)(.24) = 0.1392$$

$$\underline{0.4862}$$

$$P(I|V) = \frac{0.161}{0.161 + 0.186 + 0.1392} = \frac{0.161}{0.4862} \approx .331$$

$$b) P(L|V) = \frac{P(LV)}{P(V)} = \frac{0.186}{0.4862} \approx .383$$

$$c) P(C|V) = \frac{P(CV)}{P(V)} = \frac{0.1392}{0.4862} \approx .286$$

$$d) P(V) = 0.4862$$

9.25.2018

(70)

Generalized Version of Bayes's Formula.

Suppose events F_1, F_2, \dots, F_n are mutually exclusive
(i.e. $F_i \cap F_j = \emptyset$ if $i \neq j$)
and together they make up the whole

sample space. So

$$\bigcup_{i=1}^n F_i = S$$

Then we can write

$$E = EF_1 \cup EF_2 \cup \dots \cup EF_n$$

and then, since EF_i are mutually exclusive for $i \neq j$
and EF_j

$$P(E) = P(EF_1) + P(EF_2) + \dots + P(EF_n)$$

$$= \sum_{i=1}^n P(EF_i)$$

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

Ross
ch. 3.3
p. 69
of (3.11)

So

$$P(F_j|E) = \frac{P(EF_j)}{P(E)}$$

$$= \frac{P(EF_j)}{\sum_{i=1}^n P(EF_i)}$$

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

Prop. 3.1
Bayes in Ross
p. 69

$$P(E|F_j)P(F_j)$$

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

EXAMPLE 3a

Insurance company believes there are two classes of people, ~~some~~ accident prone and not accident prone.

Statistics \rightarrow

- accident prone person has probability 0.4 of having an accident in a 1-year period
- non-accident prone person has probability 0.2 of having an accident in a 1-year period.

⊗ Assume the 30% of the population is accident prone.

Q: What is the probability that a new policy holder will have an accident within a year of purchasing a policy

$$\begin{array}{ccccccc}
 P(A_1) & = & P(A_1|A) & P(A) & + & P(A_1|A^c) & P(A^c) \\
 \uparrow & & \underbrace{\quad} & \underbrace{\quad} & & \underbrace{\quad} & \underbrace{\quad} \\
 \text{prob. of} & & 0.4 & 0.3 & & 0.2 & 0.7 \\
 \text{new policy} & & & & & & \\
 \text{holder having} & & & & & & \\
 \text{accident} & & & & & & \\
 & = & 0.12 & + & 0.14 & = & \boxed{0.26}
 \end{array}$$

Q: A new policy holder has an accident in first year. What is the probability that he/she is accident prone?

$$\begin{aligned}
 P(A|A_1) &= \frac{P(AA_1)}{P(A_1)} = \frac{P(A_1|A)P(A)}{P(A_1)} = \frac{(0.4)(0.3)}{0.26} \\
 &= \frac{0.12}{0.26} = \frac{12}{26} = \frac{6}{13}
 \end{aligned}$$

~~RECEIVED~~

Note:

(74)

$$P(A^c | A_1) = \frac{P(A^c A_1)}{P(A_1)} = \frac{P(A_1 | A^c) P(A^c)}{P(A_1)}$$

probability of
being not accident
prone given that
the person had an
accident in the
first year.

$$= \frac{(0.2)(0.7)}{0.26} = \frac{0.14}{0.26}$$

$$= \frac{14}{26} = \left(\frac{7}{13}\right) = \text{i.e. } 1 - \frac{6}{13}$$

Q: What if we make a different assumption about
the % of the population that is accident prone?
(call it p_A)

$$P(A_1) = P(A_1 | A) P(A) + P(A_1 | A^c) P(A^c)$$

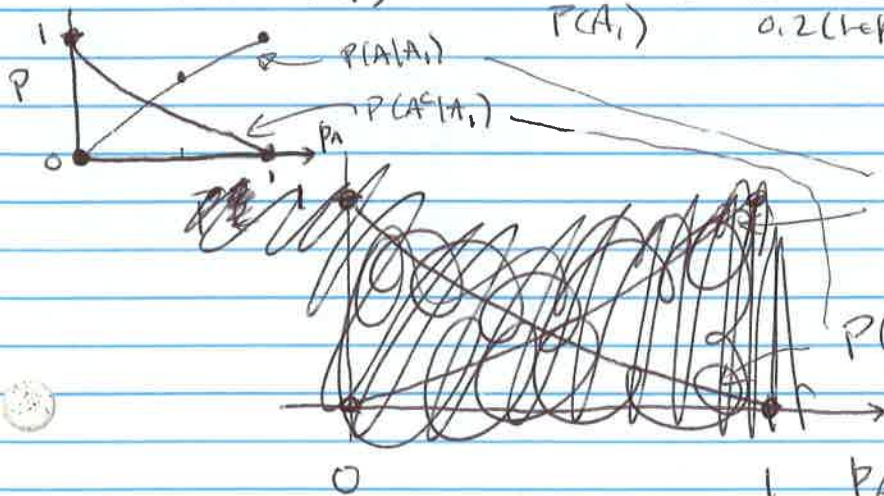
$$= (0.4) p_A + 0.2 (1 - p_A)$$

$$= 0.2 + 0.2 p_A = 0.2 (1 + p_A)$$

note: the
accident
prone
proportion
of the
population
is relatively
small in
this
assumption.
(30%)

$$P(A | A_1) = \frac{P(A A_1)}{P(A_1)} = \frac{P(A | A) P(A)}{0.2 (1 + p_A)} = \frac{0.4 p_A}{0.2 (1 + p_A)} = \frac{2 p_A}{1 + p_A}$$

$$P(A^c | A_1) = \frac{P(A^c A_1)}{P(A_1)} = \frac{P(A^c | A^c) P(A^c)}{0.2 (1 + p_A)} = \frac{0.2 (1 - p_A)}{0.2 (1 + p_A)} = \frac{1 - p_A}{1 + p_A}$$



$P(A | A_1)$ = probability they are
accident prone given
an accident

$P(A^c | A_1)$ = probability they are
not accident prone
given an accident

$p_A = \frac{\%}{100}$ of population that
is accident prone

EXAMPLE 3d P. 69

- A lab blood test is 95% effective at identifying a disease when it is actually present.
- "False Positive" tests occur for 1% of healthy people tested.
- Suppose 0.5% of the population has the disease.

Q: What is the probability that a person has the disease given that the test result is positive?

~~P(D|t_{pos})~~ let D = event that person tested has the disease

let t_{pos} = event that the test is positive

$$P(D|t_{pos}) = \frac{P(D \cdot t_{pos})}{P(t_{pos})}$$

D^c = event they do not have disease

$$= \frac{P(t_{pos}|D)P(D)}{P(t_{pos}|D)P(D) + P(t_{pos}|D^c)P(D^c)}$$

$$= \frac{(0.95)(0.005)}{(0.95)(0.005) + (0.01)(.995)}$$

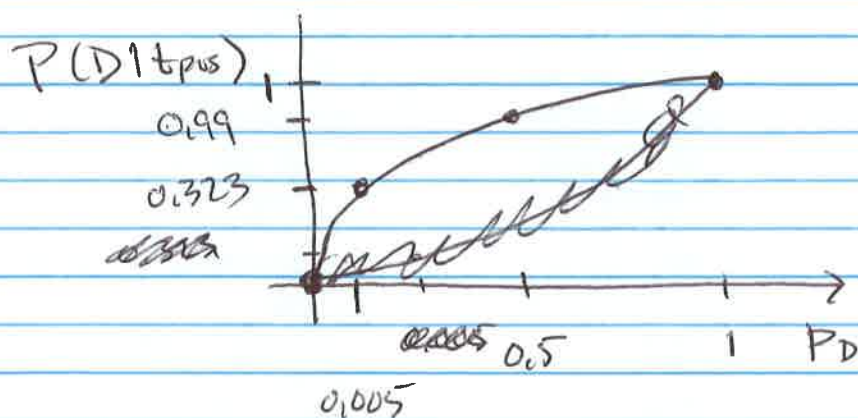
$$= \frac{0.00475}{0.00475 + 0.00995} = \frac{0.00475}{0.0147} \approx 0.323$$

(relatively low it would seem...)

What if $p_D (\times 100\%)$ is the ~~percentage~~ percentage of the population that has the disease?

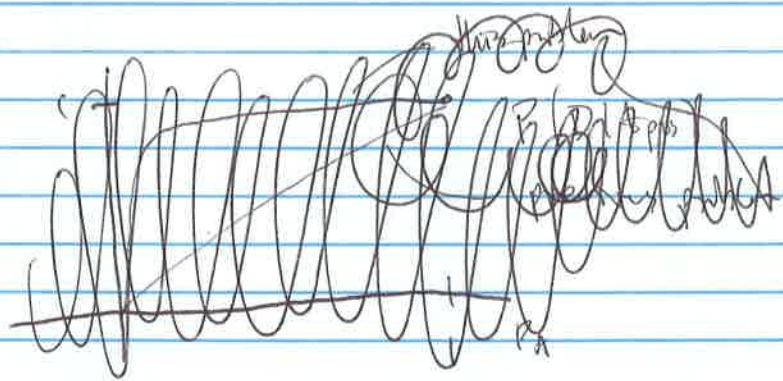
$$P(D|t_{pos}) = \frac{(0.95) p_D}{0.95 p_D + 0.01 (1-p_D)}$$

$$= \frac{95 p_D}{95 p_D + 1 - p_D} = \frac{95 p_D}{1 + 94 p_D}$$



• if $p_D = 0.005$
 $P(D|t_{pos}) = 0.323$

• if $p_D = 0.5$
 $P(D|t_{pos}) = 0.99$



EX

Suppose you have two coins in your pocket (one fair coin and one two-headed coin). Suppose you pick a coin out at random and flip it.

If the coin flip shows H (heads) what is the probability that the coin was 2-headed?

Let $Z =$ event coin was two-headed
 $F =$ " " " " fair
 $H =$ " " coin flip was heads
 $T =$ " " " " " tails

$$\begin{aligned} P(Z|H) &= \frac{P(ZH)}{P(H)} = \frac{P(H|Z)P(Z)}{P(H|Z)P(Z) + P(H|F)P(F)} \\ &= \frac{(1)(\frac{1}{2})}{(1)(\frac{1}{2}) + (\frac{1}{2})(\frac{1}{2})} = \frac{\frac{1}{2}}{\frac{3}{4}} = \left(\frac{2}{3}\right) \end{aligned}$$

What if the coin taken out of the pocket was flipped twice and both times showed Heads. What is the probability it is the two-headed coin?

$$\begin{aligned} P(Z|HH) &= \frac{P(ZHH)}{P(HH)} = \frac{P(HH|Z)P(Z)}{P(HH|Z)P(Z) + P(HH|F)P(F)} \\ &= \frac{(1)(\frac{1}{2})}{(1)(\frac{1}{2}) + (\frac{1}{4})(\frac{1}{2})} = \left(\frac{4}{5}\right) \end{aligned}$$

- What if the coin was flipped k times and showed heads k times. What is the probability it is the 2-headed coin?

$$P(Z|kH) = \frac{P(kH|Z)P(Z)}{P(kH)}$$

$$P(kH|Z)P(Z) + P(kH|F)P(F)$$

$$P(kH)$$

$$= \frac{(1)(\frac{1}{2})}{(1)(\frac{1}{2}) + (\frac{1}{2})^k - (\frac{1}{2})} = \frac{1}{1 + (\frac{1}{2})^k}$$

$$\text{as } k \rightarrow \infty \quad P(Z|kH) \rightarrow 1.$$

- What if the coin was flipped ~~2000~~ $k+1$ times and showed heads k times and tails 1 time. What is the probability it is the two headed coin?

$$P(Z|kH1T) = 0 \quad \leftarrow \text{two headed coin can never be tails.}$$