

Lecture 3: Probability, diagnostic testing and Bayes

Supplementary Reading: Pagano/Gauvreau; Chapter 6

Probability

Frequentist viewpoint: Large-sample limit of a proportion, i.e. a proportion of a defined set of outcomes

Bayesian viewpoint: The degree to which a person (or community) believes that a proposition is true, subject to available information.

Ultimately, both definitions relate to *information*

Definitions

experiment - any procedure that can be repeated (in principle, if not logistically) an arbitrarily large number of times and has a well defined set of possible outcomes

sample space - set of all possible outcomes

event - any set of outcomes of interest

probability - (of an event) relative frequency of this set of outcomes of interest

Conceptually (from a frequentist point of view): if an experiment is repeated n times under essentially the same conditions and if the event A occurs m times, then as n gets large the ratio m/n approaches a fixed limit p_0 that is the probability of A ; $p_0 = P(A) \approx m/n$. This is called the *law of large numbers*.

Example: Coin toss

Consider the experiment consisting of flipping a coin three times. The sample space is

$$S = \{ \text{HHH, HHT, HTH, THH, HTT, THT, TTH, TTT} \}$$

$$\text{Event } A = \{ \text{Majority of coins show heads} \}$$

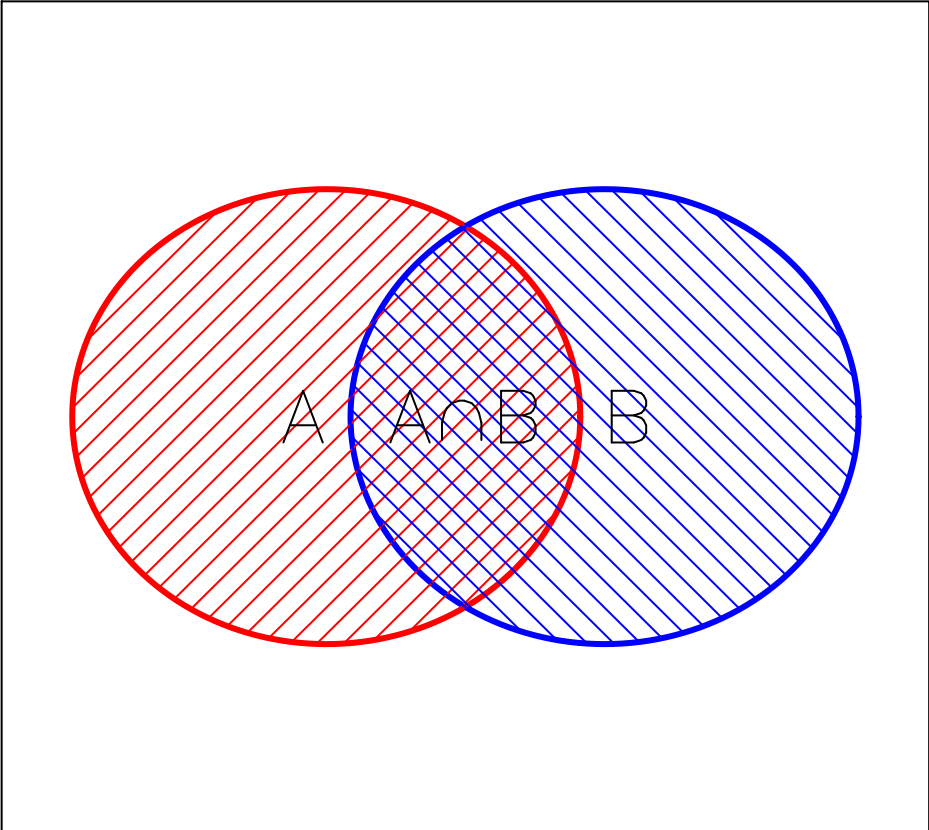
$$A = \{ \text{HHH, HHT, HTH, THH} \}$$

Basic properties of probabilities

1. The probability of an event, $P(E)$, always satisfies $0 \leq P(E) \leq 1$
2. In outcomes A and B are two events that cannot both happen at the same time, then A and B are **mutually exclusive**:
 $P(A \text{ or } B) = P(A) + P(B)$

More definitions/notation

1. $A \cup B$ is the event that either A or B occurs or they both occur - **union**
2. $A \cap B$ is the event that both A and B occur simultaneously - **intersection**
3. A^c (or \bar{A}) is the event that A does not occur - **complement**;
 $P(A^c) = 1 - P(A)$
4. Note that $P(A \cup B) = P(A) + P(B) - P(A \cap B)$. Why?



Operations on events

Let events

$A = \{30 \text{ year old woman lives to see her } 70^{\text{th}} \text{ birthday}\}$

$B = \{\text{her } 30 \text{ year old husband lives to see his } 70^{\text{th}} \text{ birthday}\}$

Intersection of events A and B:

$A \cap B = \{ \text{both live to } 70 \}$

Union of events A and B:

$A \cup B = \{\text{the woman lives to } 70 \text{ or the man lives to } 70, \text{ or both live to } 70\}$

Complement of event : $A^c = \{30 \text{ year old woman does not live to } 70\}$

Example :

Let $A = \text{set of integers } x \text{ for which } x^2 + 2x = 8$

Let $B = \text{set of integers } x \text{ for which } x^2 + x = 6$

Let $C = \text{set of integers } x \text{ for which } x^2 - 6x = 5$

$A = \{-4, 2\}$, $B = \{-3, 2\}$, and $C = \{1, 5\}$

$A \cap B = \{2\}$

$A \cup B = \{-4, -3, 2\}$

$A \cap C = \emptyset$ (A and C are *mutually exclusive*)

Definitions/notation (con't)

Two events A and B are **independent** if $P(A \cap B) = P(A) \times P(B)$

If two events are **dependent**, then $P(A \cap B) \neq P(A) \times P(B)$

Example: Consider all possible diastolic blood pressure (DBP) measurements from a mother and her first-born child.

Let $A =$ mothers DBP ≥ 95 and $B =$ first-born child's DBP ≥ 80

Suppose $P(A \cap B) = 0.05$; $P(A) = 0.1$; and $P(B) = 0.2$

Then, $P(A \cap B) = 0.05 > P(A) \times P(B)$

So A and B are dependent

A^c (or \bar{A}) is the event that A does not occur - **complement**;
 $P(A^c) = 1 - P(A)$

In general, if A_1, \dots, A_k are independent events, then $P(A_1 \cap \dots \cap A_k) = P(A_1) \times \dots \times P(A_k)$ - **multiplicative law of probability**

Properties of probabilities

$$P(A^c) = 1 - P(A)$$

$$P(A \cup A^c) = 1$$

$$P(A \cap A^c) = 0$$

Examples:

$$P(\text{a child is not male}) = P(\text{a child is female})$$

$$P(\text{a child is male or female}) = 1$$

$$P(\text{a child is male and female}) = 0$$

$$P(A \cup B) = P(A) + P(B) \quad \text{if } A \cap B = \emptyset$$



$$1 = P(A \cup A^c) = P(A) + P(A^c)$$



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

iff A and B are independent



Conditional probability

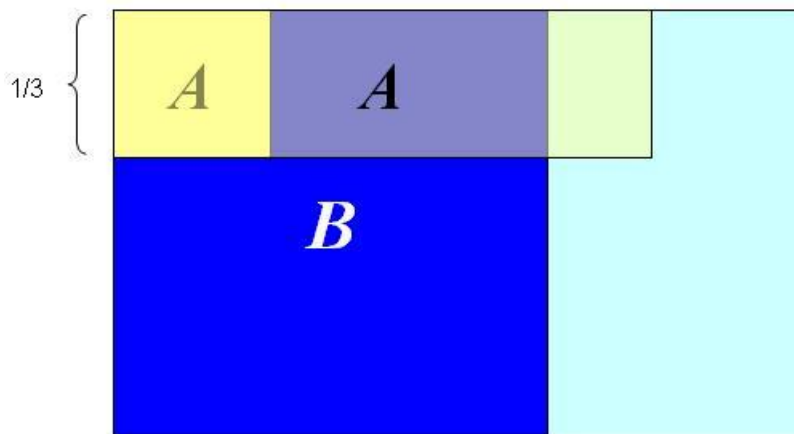
Say we are interested in finding the probability a person will survive for the next five years given that she/he has already reached the age of 60.

Interested in the probability that one event will occur given another has already taken place, i.e., the conditional probability denoted $P(A|B)$

Definition: $P(A|B) = P(A \cap B)/P(B)$

If A and B are independent events, then $P(A|B) = P(A) = P(A|B^c)$

If two events A and B are dependent, then $P(A|B) \neq P(A) \neq P(A|B^c)$



Example: Compute the probability of the event $B|A$ where

A = a person in the US who lives to 60 yrs

B = a person in the US who lives to 65 yrs

$\rightarrow B|A$ = a 60 yr old person in the US will live to age 65

Data: $P(A) = 0.85$; $P(B) = 0.79$

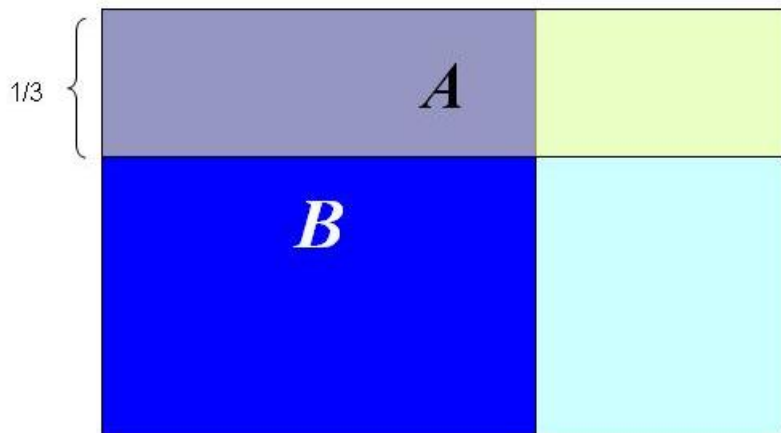
Exercise: Show $P(B) = P(B|A) \times P(A) + P(B|A^c) \times P(A^c)$

Definitions/notation (con't)

Independent events: Two events are said to be independent if $P(A|B) = P(A)$

Intuitively, the knowledge that B has occurred does not affect the probability that A will occur.

Note that $P(A \cap B) = P(A|B)P(B) = P(A)P(B)$ – multiplicative rule from before.



Example: Diagnostic testing

Notation:

D = has disease

D^c = does not have disease

T = test positive

T^c = test negative

Goal: To determine the probability that individual has the disease of interest, given they tested positive ($\rightarrow P(D|T)$)

More notation:

$P(D)$ = **prevalence** of the disease

$P(T|D)$ = **sensitivity** of the test

$P(T|D^c)$ = probability of a false positive test

$P(T^c|D)$ = probability of a false negative test

$P(T^c|D^c)$ = **specificity** of the test

Ideally, for a diagnostic test to be effective at predicting “disease” (or a procedure to be effective at predicting an “event”), we want both sensitivity and specificity to be high. There are usually tradeoffs.

Bayes' Rule

Let T = symptom (positive test) and D = disease

$$P(D|T) = \frac{P(T|D) \times P(D)}{P(T|D) \times P(D) + P(T|D^c) \times P(D^c)}$$
$$P(D|T) = \frac{\text{sensitivity} \times \text{prevalence}}{\text{sensitivity} \times \text{prevalence} + (1 - \text{specificity}) \times (1 - \text{prevalence})}$$

$P(D|T)$ – **predictive value of a positive test**

$P(D^c|T^c)$ – **predictive value of a negative test**

Example: X-Ray screening for TB

| X-Ray | Tuberculosis | | Total |
|----------|--------------|-----|-------|
| | No | Yes | |
| Negative | 1739 | 8 | 1747 |
| Positive | 51 | 22 | 73 |
| Total | 1790 | 30 | 1820 |

Question: What is $P(D|T)$ – What is the probability someone has TB given they had a positive X-Ray?

Prevalence of TB: $P(D) = 9.3$ in 100,000

$$P(D) = 0.000093 \text{ (given)}$$

$$P(D^c) = 1 - P(D) = 0.999907$$

$$P(T|D) =$$

$$P(T^c|D) =$$

$$P(T|D^c) =$$

$$P(T^c|D^c) =$$

So,

$$P(D|T) = \text{_____}$$

Bayes

Revising previous beliefs based on new “evidence” (Bayesian concept)

$$P(D) = 0.000093 \text{ prior probability}$$
$$P(D|T) = 0.00239 \text{ posterior probability}$$

$P(D|T)$ is updated probability of TB based on new “evidence”

Consider the ratio of the prior and posterior probabilities: $\frac{0.00239}{0.000093} = 25.7$. In essence, updating our knowledge has reduced our degree of uncertainty about the prevalence of TB by roughly 26 fold.

Example: Accuracy of home pregnancy test kits

Background

2 kinds of pregnancy tests (blood and urine); test detects levels of hCG (human chorionic gonadotropic)– a hormone secreted by at different times during pregnancy and builds up over time; hCG generally doubles every 2 days during the first few weeks of pregnancy

Home tests: most sensitive 20 mIU/ml of hCG in urine (250 mIU/ml)

$$A_1 = \{\text{woman pregnant}\}$$

$$A_2 = \{\text{woman NOT pregnant}\}$$

$$T = \{\text{positive test result}\}$$

$$\text{Sensitivity} = 80\%$$

$$\text{Specificity} = 68\%$$

$$\text{Prevalence of pregnancy} = 60\%$$

1. What is the probability that a woman with a positive home pregnancy test result is actually pregnant?
2. What is the probability that a woman with a negative home pregnancy test result is NOT pregnant?