CSE 103: Probability and statistics

Fall 2010

Topic 2 — Multiple events, conditioning, and independence, I

2.1 Two or more events on the same sample space

Frequently we end up dealing with multiple events A, B, \ldots defined on a single sample space Ω . For instance, Ω might denote the possible outcomes of picking a card from a deck, and the events of interest could include $A = \{\omega : \omega \text{ is a number}\}$, $B = \{\omega : \omega \text{ is a spade}\}$, and so on. Instead of just the single event probabilities $\Pr(A)$ and $\Pr(B)$, we are also interested in probabilities of combinations of events, such as $\Pr(A \cap B)$ and $\Pr(A \cup B)$.

2.1.1 Human intuition about combined event probabilities is fallible

Let's start with Exercise 1.2.25 from the textbook. The famous psychologists Kahneman and Tversky spent a lot of time trying to understand the nature of human rationality and decision making. For instance, do humans think probabilistically? In one experiment, they told subjects the following:

Linda is 31, single, outspoken, and very bright. She majored in philosophy in college. As a student, she was deeply concerned with racial discrimination and other social issues, and participated in anti-nuclear demonstrations.

and then asked them to assess the likelihood of three events:

- (A) Linda is active in the feminist movement.
- (B) Linda is a bank teller.
- (C) Linda is a bank teller who is active in the feminist movement.

A significant majority of respondents deemed A to be the most likely, followed by C, and lastly B. But $Pr(C) = Pr(A \cap B) \leq Pr(B)$! The respondents exhibited what is called a *conjunction fallacy*.

2.1.2 Joint probability tables

One way to write down the probabilities of pairs of events is in a table of all possible outcomes. Let's take another exercise from the textbook, 1.2.15. Here two students, John and Mary, are taking a math course in which the possible grades are A, B, and C. The sample space of possible outcomes can be written as $\Omega = \{A, B, C\}^2$, where the first coordinate of $\omega \in \Omega$ is John's grade and the second is Mary's grade. We can write down the probabilities of the nine possible outcomes in a 3×3 table:

		Mary		
		A	B	C
	A	p_{AA}	p_{AB}	p_{AC}
John	B	p_{BA}	p_{BB}	p_{BC}
	C	p_{CA}	p_{CB}	p_{CC}

The particular problem (see book) can be solved by writing the evidence in terms of these probabilities and adding and subtracting appropriate events.

2.1.3 The union bound

As another example, suppose that in a city,

60% of people have a car,

20% of people have a bicycle, and

10% of people have a motorcycle.

Anybody who lacks all three ends up walking to work, and nobody walks unless they have to. Based on this data, what fraction of people walk to work?

We don't have enough information to give an exact answer. This is because we don't know about the overlap in ownership of different vehicles: what fraction of the people with cars also have bicycles, and so on. Nonetheless, we can give upper and lower bounds. To get started, consider the experiment of picking a random person in the city, and let Ω be the space of outcomes. We then define the four events of interest: $C = \{\text{owns a car}\}$, $B = \{\text{owns a bicycle}\}$, $M = \{\text{owns a motorcycle}\}$, and $W = \{\text{walks}\}$. Since $W = \Omega - (C \cup B \cup M)$, we need to bound $\Pr(C \cup B \cup M)$.

The least $\Pr(C \cup B \cup M)$ could possibly be is 0.6, which occurs when everyone who has a bicycle or motorcycle also owns a car. The most $\Pr(C \cup B \cup M)$ could be is 0.9, which happens when the sets C, B, M are disjoint. Therefore $0.1 \le \Pr(W) \le 0.4$.

Generalizing, we get the extremely useful **union bound**: for any events A_1, A_2, \ldots, A_k

$$\Pr(A_1 \cup \cdots \cup A_k) \leq \Pr(A_1) + \cdots + \Pr(A_k).$$

2.2 Balls in bins, or urn problems

You have m indistinguishable balls and in front of you is a row of n bins. You place each ball into a bin chosen at random.

Let's write the sample space as $\Omega = \{1, 2, ..., m\}^n$; in each outcome $\omega = (\omega_1, ..., \omega_n)$, the value ω_i represents the number of balls in the *i*th bin.

Here are some interesting questions one can ask.

1. What is the chance that the ith bin is empty? (What is $Pr(\omega_i = 0)$?)

Well, there are m balls, and we want every single one of them to miss the ith bin. The probability that the first ball misses this bin is (n-1)/n = 1 - 1/n. The probability that the second ball misses is also 1 - 1/n, as with the third, and fourth, and so on. Therefore,

$$\Pr(i\text{th bin empty}) = \left(1 - \frac{1}{n}\right)^m \le (e^{-1/n})^m = e^{-m/n},$$

where we've used the formula $e^x \ge 1 + x$ that we discussed earlier.

2. What is the chance that there is an empty bin if m = n?

Intuitively, one would expect this to be pretty close to 1 (that is, pretty much certain), because the complementary event – no empty bins – would occur only if every single bin received exactly one ball. Let's analyze this latter probability.

$$\Pr(\text{every bin gets a ball}) \ = \ \left(1 - \frac{1}{n}\right) \left(1 - \frac{2}{n}\right) \cdots \frac{1}{n} \ = \ \frac{n!}{n^n}.$$

This is miniscule; for instance, it is less than $1/2^{n/2}$.

3. What is the chance that there is an empty bin if $m = 2n \ln n$?

When m is increased from n to $2n \ln n$, the chance of an empty bin drops from ≈ 1 to ≈ 0 . To see this, let A_i be the event that the ith bin is empty. We proved above that $\Pr(A_i) \leq \exp(-m/n) = \exp(-2 \ln n) = 1/n^2$. Therefore

$$\Pr(\text{some bin is empty}) = \Pr(A_1 \cup \dots \cup A_n) \leq \Pr(A_1) + \dots + \Pr(A_n) \leq n \cdot \frac{1}{n^2} = \frac{1}{n}$$

(the first inequality is the union bound). Thus with high probability (at least 1 - 1/n), every bin gets at least one ball.

4. The coupon collector problem. Many probability questions turn out to be thinly disguised balls-and-bins problems. Here's an example. Suppose that each cereal box contains one of k action figures (chosen uniformly at random). How many cereal boxes should you buy in order to collect all k figures? If you make the following associations:

 $\operatorname{bin} \equiv \operatorname{action figure}$ $\operatorname{throw a ball} = \operatorname{buy a box of cereal}$ $\operatorname{every bin gets a ball} = \operatorname{you collect all the figures},$

then you see that what we are really asking is, how many balls do you need to throw into n = k bins in order to hit all of them? And we've already seen that $m = 2k \ln k$ balls (cereal boxes) will suffice.

5. What is the probability that some bin gets two or more balls?

This is another situation in which it is easier to study the complementary event, that every bin gets at most one ball.

$$\Pr(\text{every bin gets 0 or 1 ball}) \ = \ \left(1 - \frac{1}{n}\right) \left(1 - \frac{2}{n}\right) \cdots \left(1 - \frac{m-1}{n}\right) \ = \ \frac{n!}{(n-m)!n^m}.$$

Although this is exact, it is a little difficult to fathom, and so let's try an approximation instead, using $1 - i/n \le e^{-i/n}$:

Pr(every bin gets 0 or 1 ball)
$$= \left(1 - \frac{1}{n}\right) \left(1 - \frac{2}{n}\right) \cdots \left(1 - \frac{m-1}{n}\right)$$

$$\leq \exp(-1/n) \exp(-2/n) \cdots \exp(-(m-1)/n)$$

$$= \exp\left(-\frac{1}{n}\left(1 + \cdots + (m-1)\right)\right)$$

$$= \exp\left(-\frac{m(m-1)}{2n}\right).$$

This is an excellent approximation when $m \ll n$. It tells us that if $m \ge c\sqrt{n}$ (for some small constant c), then this probability is at most 1/2. That is, if you throw (roughly) \sqrt{n} (or more) balls into n bins, then chances are that some bin will get two or more balls.

6. Birthday paradox. We saw earlier than in a room with 23 people, chances are two of them will share the same birthday.

This is yet another balls-and-bins problem:

bin \equiv day of the year

throw a ball \equiv select birthday of a person in the room

bin with 2 balls \equiv two people with the same birthday.

The number of bins is n = 365 and the number of balls is m = 23. Using the formula we just obtained,

Pr(two people share the same birthday) = 1 - Pr(every bin has at most one ball)

$$\geq 1 - \exp\left(-\frac{m(m-1)}{2n}\right) \geq \frac{1}{2}.$$

7. Coin tossing.

Even the tossing of a fair coin, m times, can be modeled using balls and bins. Imagine there are just n=2 bins (call one bin H and the other bin T).

2.3 Conditional probability

You meet a stranger, a random citizen of the United States. What's the chance that he (or, equally likely, she) votes for the Democratic party? Who knows, but it's probably close to 50%.

$$\Omega = \{ \text{citizens of the US} \}, E = \{ \omega \in \Omega : \text{votes Dem} \}.$$

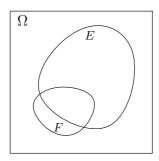
Now what if you find out a little bit more information: this stranger likes spicy food.

$$F = \{ \omega \in \Omega : \text{likes spicy food} \}.$$

How does change the odds? What is Pr(E|F) ("probability of E given F")? Again, it's hard to say, but for illustrative purposes here is a (definitely incorrect) set of probabilities:

$$Pr(E) = 0.5, \quad Pr(F) = 0.2, \quad Pr(E \cap F) = 0.15.$$

The Venn diagram for this scenario is



If you stare at it a little bit, you will probably conclude, correctly, that Pr(E|F) is the chance that a point drawn from the small blob lies inside the large blob, namely 0.15/0.2 = 0.75. By reasoning this way, you are implicitly using the formula for conditional probability

$$\Pr(E|F) = \frac{\Pr(E \cap F)}{\Pr(F)}.$$

We will make heavy use of this and of the equivalent

$$Pr(E \cap F) = Pr(E|F)Pr(F).$$

2.3.1 A few examples

1. Pregnancy test. How accurate is a pregnancy test? Here we make up some numbers and compute the relevant conditional probabilities.

The sample space in this case consists of women who use the test; call this Ω . There are two events on this space that we care about: $P = \{\text{actually pregnant}\}\$ and $T = \{\text{test says pregnant}\}\$. Suppose that the following are determined (warning: these are fabricated!):

$$T \subset P$$
, $Pr(P) = 0.4$, $Pr(T) = 0.3$.

There are two conditional probabilities we'd like to compute.

- What is the chance of pregnancy if the test comes out positive? Well, since $T \subset P$, we have $\Pr(P|T) = 1.0$. Algebraically, $\Pr(P \cap T) = \Pr(T)$, so $\Pr(P|T) = \Pr(P \cap T)/\Pr(T) = 1.0$.
- What is the chance of pregnancy if the test comes out negative? Let $T^c = \Omega - T$ be the event that the test is negative. Then $\Pr(T^c) = 1 - \Pr(T) = 0.7$ and $\Pr(P \cap T^c) = \Pr(P) - \Pr(P \cap T) = 0.4 - 0.3 = 0.1$. These are the two probabilities we need.

$$\Pr(P|T^c) = \frac{\Pr(P \cap T^c)}{\Pr(T^c)} = \frac{0.1}{0.7} \approx 0.14.$$

- 2. Rolls of a die. You roll a die twice. What is the probability that the sum of the two rolls is ≥ 10 if:
 - The first roll is 6?
 We could use the conditional probability formula, but that seems like overkill in so straightforward a situation.

$$\Pr(\text{sum} \ge 10 \mid \text{first} = 6) = \Pr(\text{second} \ge 4) = \frac{1}{2}.$$

• The first roll is ≥ 3 ?

Okay, this is not so trivial anymore. The sample space is $\Omega = \{1, 2, 3, 4, 5, 6\}^2$, with each of the 36 outcomes equally likely.

$$\begin{array}{lll} \Pr(\operatorname{sum} \geq 10 \mid \operatorname{first} \geq 3) & = & \frac{\Pr(\operatorname{sum} \geq 10 \text{ and } \operatorname{first} \geq 3)}{\Pr(\operatorname{first} \geq 3)} \\ & = & \frac{\Pr(\{(4,6),(5,5),(5,6),(6,4),(6,5),(6,6)\})}{4/6} \ = & \frac{6/36}{4/6} \ = & \frac{1}{4}. \end{array}$$

• The first roll is < 6?

$$\begin{array}{lll} \Pr(sum \geq 10 \mid first < 6) & = & \dfrac{\Pr(sum \geq 10 \; \text{And} \; first < 6)}{\Pr(first < 6)} \\ & = & \dfrac{\Pr(\{(5,5),(5,6),(4,6)\})}{5/6} \; = \; \dfrac{3/36}{5/6} \; = \; \dfrac{1}{10}. \end{array}$$

3. To be invisible or to fly? A few years ago, the question was going round: "which super power would you rather possess: the ability to make yourself invisible or to fly?" It was perhaps an urban legend that men usually wanted to fly while women usually wanted to be invisible; and there was much speculation

about the psychological reasons for this. But suppose the following statistics were obtained (warning: these numbers are fabricated!):

$$Pr(fly|male) = 0.8$$
, $Pr(invisible|female) = 0.6$.

Then what is the overall fraction of people who would choose the ability to fly?

This cannot be answered unless it is known what fraction of the population is male and what fraction is female. Suppose it is 60-40. Then

$$\begin{array}{lll} \Pr(\text{fly}) & = & \Pr(\text{fly AND male}) + \Pr(\text{fly AND female}) \\ & = & \Pr(\text{fly}|\text{male}) \Pr(\text{male}) + \Pr(\text{fly}|\text{female}) \Pr(\text{female}) \\ & = & (0.8)(0.6) + (0.4)(0.4) = 0.64. \end{array}$$

That is, 64% would like to fly.

2.3.2 The summation rule

In the last example, we implicitly used a **summation formula**. Suppose $A_1, \ldots, A_k \subset \Omega$ are disjoint events whose union is Ω (that is, exactly one of them will occur). Then for any event $E \subset \Omega$,

$$\Pr(E) = \sum_{i=1}^{k} \Pr(E \cap A_i) = \sum_{i=1}^{k} \Pr(E|A_i) \Pr(A_i).$$

(If the A_i are not disjoint, simply replace the first equality with a \leq .) We will often write $\Pr(E \cap A_i)$ as $\Pr(E, A_i)$.