#### Lecture 4: More counting examples; Kelly 1.2, A.4

## 4.1 Binomial counts and Stirling's formula

(i) Suppose there are N equiprobable outcomes in  $\Omega$ .

Suppose event E is true for R of these outcomes. Then P(E) = R/N.

(ii) An AB parent and an O parent can have an A child or a B child.

Suppose they have n children: there are  $2^n$  possible sequences of A and B children.

Assume these are equiprobable. (In fact, they are.)

 $\binom{n}{k}$  of these sequences have k A children.  $P(k \ A \ children \ out \ of \ n) = \binom{n}{k}/2^n$ .

(iii) n! can be approximated for large n by  $\sqrt{2\pi}n^{n+\frac{1}{2}}e^{-n}$ . Also  $\binom{n}{k}$  is largest when  $k \approx n/2$ .

Then, for large 
$$n$$
,  $\binom{n}{n/2} = \frac{n!}{(n/2)!(n/2)!} \approx \frac{\sqrt{2\pi n^{n+\frac{1}{2}}e^{-n}}}{\sqrt{2\pi}(n/2)^{(n/2)+\frac{1}{2}}e^{-(n/2)} \times \sqrt{2\pi}(n/2)^{(n/2)+\frac{1}{2}}e^{-(n/2)}}$   
=  $(1/\sqrt{2\pi})2^{n+1}n^{n+\frac{1}{2}-(n/2)-\frac{1}{2}} = (1/\sqrt{2\pi})(2/\sqrt{n})2^{n}$   
Or  $P((n/2) A$  children out of  $n) = \binom{n}{n/2}(\frac{1}{2})^{n} = 1/\sqrt{2\pi(n/4)}$ 

This result will have meaning later when we discuss approximating Binomial probabilities by the Normal probability distribution.

#### 4.2 The binomial theorem; Kelly A.4.7

$$(x+y)^n = \sum_{k=0}^n \left( \begin{array}{c} n \\ k \end{array} \right) x^k y^{n-k}$$

Note in each bracket we choose x or y. There are  $2^n$  sequences.

The number of sequences in which there are k choices of x is  $\binom{n}{k}$ , and each has value  $x^k y^{n-k}$ .

The case of 4.1 (ii) is a special case when  $x = y = \frac{1}{2}$ .

An alternative proof is by induction using result of 2.2 (iv).

# 4.3 Sampling with and without replacement (Kelly 1.2, P. 23)

(i) Draw 3 cards from 52-card pack.  $E = \{ \text{draw at least one face card} \}$ . Note  $E^c$  (no face card) is easier. With replacement:  $P(\text{no face card}) = (40 \times 40 \times 40)/(52 \times 52 \times 52) = 0.455.$ 

$$P(\text{at least 1 face card}) = 1 - 0.455 = 0.545.$$

Without replacement:  $P(\text{no face card}) = (40 \times 39 \times 38)/(52 \times 51) \times 50) = 0.447.$ 

(ii) The birthday problem: ignore Feb 29, and assume other days equiprobable.

In k people, E is event that at least 2 share a birthday.  $E^c$  is the event that all bithdays are different.

 $P(E^c) = 365.364.363.362....(365 - k + 1) / 365.365.365....365 = 365! / (365 - k)! \times (1/365)^k$ 

$$k = 2, P(E) = 1/365; k = 23, P(E) \approx 0.5; k = 45, P(E) = 0.94$$

(iii) Hypergeometric probabilities: N fish in a pond; n are red, N - n are blue.

Sample k fish without replacement. What is the probability x are red?

Total number of outcomes is  $\binom{N}{k}$ . Number of ways of choosing the x red fish is  $\binom{n}{x}$ , and the (k - x blue)

fish is 
$$\binom{N-n}{k-x}$$
. Probability is  $\binom{n}{x} \times \binom{N-n}{k-x} / \binom{N}{k}$ 

#### Lecture 5: Conditional probability and independence Kelly 2.2, 2.4

#### **5.1 Conditional probability** (Kelly 2.2)

(i) Idea: revise probabilities in light of restrictions or partial information.

(ii) Definition: probability of D given E is  $P(D | E) = P(D \cap E)/P(E)$ , provided P(E) > 0. (iii) Example: Our AB and O parents have three children.

 $\Omega = \{AAA, AAB, ABA, ABB, BAA, BAB, BBA, BBB\}$  and these are equiprobable (1/8 each). Let *D* be at least 2 *A* children: P(D) = 4/8 = 1/2. Note also P(first child is A) = 4/8 = 1/2.  $P(D \mid \text{first child is } A) = P(D \cap \text{first child is } A)/P(\text{first child is } A) = (3/8)/(4/8) = 3/4$ . **5.2 The chain rule** (Kelly 2.2.10, 2.2.11) If  $P(E_1 \cap E_2 \cap ... \cap E_n) > 0$ ,

$$P(E_1 \cap E_2 \cap ... \cap E_n) = P(E_1) \cdot P(E_2 | E_1) \cdot P(E_3 | E_1 \cap E_2) \dots P(E_n | E_1 \cap E_2 \cap ... \cap E_n)$$

To prove, just write it out:

$$P(E_1).P(E_2|E_1).P(E_3 | E_1 \cap E_2)....P(E_n | E_1 \cap E_2 \cap .... \cap E_n) = P(E_1)\frac{P(E_2 \cap E_1)}{P(E_1)}\frac{P(E_3 \cap E_2 \cap E_1)}{P(E_2 \cap E_1)}... \dots \frac{P(E_n \cap ... \cap E_3 \cap E_2 \cap E_1)}{P(E_{n-1} \cap .. \cap E_2 \cap E_1)}$$

This looks messy, but in fact we have already used it in examples of sampling without replacement. Example: No face card in 3 cards, when sampling without replacement. (40/52).(39/51).(38/50).

### 5.3 Bayes' formula: Kelly 2.3

Assume P(D) and P(E) are both > 0. Then, by definition,

$$P(D \mid E) P(E) = P(D \cap E) = P(E \mid D) P(D)$$
 or  $P(D \mid E) = P(E \mid D) P(D) / P(E)$ 

Note also, from the law of total probability, we could compute P(E) as

$$P(E) = P(E \cap D) + P(E \cap D^{c}) = P(E|D)P(D) + P(E | D^{c})(1 - P(D))$$

**Example:**  $P(\text{first child is } A \mid D \equiv \text{``at least } 2A \text{ children''})$ 

$$= P(D \mid \text{first is } A)P(\text{first is } A)/P(D) = (3/4) \times (1/2) / (1/2) = 3/4.$$

For better examples see Lecture 6: P(D|A) = P(A|D) ONLY BECAUSE P(D) = P(A).

#### **5.4 Independent events** (Kelly 2.4)

(i) Definition: E and F are independent if  $P(E \cap F) = P(E) \times P(F)$ .

(ii) Interpretation: Knowing F happens does not affect the probability of E: P(E|F) = P(E).

(iii) If E and F are independent,  $P(E \cup F) = P(E) + P(F) - P(E) \cdot P(F)$  so

 $P(E^{c} \cap F^{c}) = 1 - P(E \cup F) = 1 - P(E) - P(F) + P(E) \cdot P(F) = (1 - P(E)) \cdot (1 - P(F)) = P(E^{c}) \cdot P(F^{c})$ . That is  $E^{c}$  and  $F^{c}$  are independent. (So are  $E^{c}$  and F, and E and  $F^{c}$ ).

Example: P(A) = 0.36, P(B) = 0.2, P(AB) = 0.08, P(O) = 0.36. P(have antigen A) = P(A) + P(AB) = 0.44, P(have antigen B) = P(B) + P(AB) = 0.28.

 $0.44 \times 0.28 = 0.1232 \neq 0.08 = P(AB)$ : Having antigen A is NOT independent of having antigen B.

 $P(\text{have antigen } A \mid \text{have antigen } B) = P(AB)/P(\text{have antigen } B) = 0.08/0.28 = 0.286$ 

< 0.44 = P(have antigen A)

Knowing a person has the B antigen decreases the probability they have the A antigen.

#### Lecture 6: Probability Examples

#### 6.1 More conditional probabilities

(i) Mendel discovered that every pea plant has two *factors* or *alleles* for flower color: R (red) and W (white). Each plant has RR or RW and is red, or WW and is white.

We cross two plants which we know are RW. The offspring plants are *independent* and each is RR, RW or WW with probabilities 1/4, 1/2, 1/4.

What is the probability an offspring is RR, given it is red.

Answer: P(red) = 3/4,  $P(RR \cap \text{red}) = P(RR) = 1/4$ , so  $P(RR \mid \text{red}) = (1/4)/(3/4) = 1/3$ .

(ii) Testing for a rare disease (Kelly: Exx 2.3.7)

Suppose we have a quite effective test, so P(+ | disease) = 0.99, and test is quite accurate, so P(+ | no disease) = 0.02). Now suppose the frequency of the disease is 0.001.

$$P(+ \text{ test result}) = P(+ | \text{ disease})P(\text{disease}) + P(+ | \text{ no disease})P(\text{no disease})$$
  
= 0.99 × 0.001 + 0.02 × (1 - 0.001) = 0.00099 + 0.0198 = 0.02097  
$$P(\text{disease} | +) = P(+ | \text{ disease})P(\text{disease})/P(+) = 0.99 \times 0.001/0.02097 = 0.047$$

Less than 5% of people testing positive actually have the disease!

The probability of spots given measles is large: the probability of measles given spots is small.

The probability of 4 legs given elephant close to 1: the probability of elephant given 4 legs is close to 0.

## 6.2 Updating with more information

(i) Allele frequencies in a population of pea plants.

Suppose the R allele has frequency 0.3, and W 0.7, and the types of the two alleles in an individual are independent. What is the probability a red-flowered pea-plant is RR?

Solution: By independence,  $P(RR) = 0.3 \times 0.3 = 0.09$ ,  $P(WW) = 0.7 \times 0.7 = 0.49$ , P(RW) = 1 - 0.09 - 0.49 = 0.42.

Overall P(red) = P(RR) + P(RW) = 0.09 + 0.42 = 0.51,

so then  $P(RR \mid \text{red}) = P(RR \cap \text{red})/P(\text{red}) = 0.09/0.51 = 0.176.$ 

(ii) The red pea plant is crossed to a white one (WW). The first offspring has red flowers. What now is the probability the red parent plant is RR?

Solution: P(red offspring | RR parent) = 1, P(red offspring | RW parent) = (1/2).

 $P(\text{red offspring}) = P(\text{red offspring} \mid RR)P(RR) + P(\text{red offspring} \mid RW)P(RW)$ 

$$= 1 \times 0.176 + (1/2)(1 - 0.176) = 0.588.$$

 $P(\text{red parent is } RR \mid \text{red offspring}) = P(\text{red offspring} \mid \text{red parent is } RR) \times P(\text{red parent is } RR) / P(\text{red offspring})$ =  $1 \times 0.176 / 0.588 = 0.299 \approx 0.3$ .

## 6.3 Independent events

(i) Back to the two RW parent pea plants of 6.1 (i):

Each offspring is red with probability 3/4, and white with probability 1/4, and the colors of offspring pea plants are *independent*. We grow up 10 offspring.

What is the probability all 10 are red: answer  $(3/4)^{10} = 0.0563$ 

What is the probability of at least one red: answer  $1 - P(\text{all white}) = 1 - (1/4)^{10} = 0.9999$ .