

## Lecture 1: Sample spaces and events: Larson & Marx, 2.2; Sept 26, 2018

### 1.1 Sample spaces

The *sample space*  $\Omega$  is the set of all possible outcomes of an experiment.

One and only one outcome can occur.

### 1.2 Examples

- (i) Child is boy or girl:  $\Omega = \{\text{boy, girl}\}$
- (ii) Toss of one die:  $\Omega = \{1, 2, 3, 4, 5, 6\}$
- (iii) Number of traffic accidents:  $\Omega = \{0, 1, 2, 3, 4, \dots\} = \{0\} \cup \mathbb{Z}_+$ .
- (iv) Time waiting for the bus:  $\Omega = (0, \infty) = \mathbb{R}_+$ , the positive half line.

### 1.3 Events

Any subset  $E$  of  $\Omega$  is a *event*.

(The book says this: it is OK for countable sample spaces, but an oversimplification for a space like  $\mathbb{R}_+$ .)

### 1.4 Combining events

- (i) If  $E$  is an event, not- $E$  (the *complement* of  $E$ : written  $E^c$ ) is an event.
- (ii) If  $E$  is an event and  $F$  is an event, then “ $E$  and/or  $F$ ” is an event. “ $E$  and/or  $F$ ” is written  $E \cup F$ .
- (iii) If  $E_1, E_2, \dots$  are events then  $E_1 \cup E_2 \cup E_3 \dots$  is an event. (Countable unions.)
- (iv) If  $E$  is an event and  $F$  is an event, then “ $E$  and  $F$ ” is an event. “ $E$  and  $F$ ” is written  $E \cap F$ .
- (v) If  $E_1, E_2, \dots$  are events then  $E_1 \cap E_2 \cap E_3 \dots$  is an event. (Countable intersections.)

### 1.5 More events:

- (i) The empty set  $\Phi$  is an event: so  $\Omega = \Phi^c$  is an event.

If  $E \cap F = \Phi$ ,  $E$  and  $F$  are *disjoint* also known as *mutually exclusive*.

- (ii)  $E$  and  $E^c$  are *disjoint events*:  $E \cup E^c = \Omega$ ,  $E \cap E^c = \Phi$ .
- (iii) Events  $E_1, E_2, \dots, E_k$  are *mutually exclusive* if  $E_i \cap E_j = \Phi$  for all pairs  $(i, j)$  ( $i, j = 1, \dots, k$ ,  $i \neq j$ ).
- (iv) Events  $E_1, E_2, \dots, E_k$  are *exhaustive* if  $E_1 \cup E_2 \cup \dots \cup E_k = \Omega$ .
- (v) If  $\Omega$  is discrete, the elements of  $\Omega$  are a set of mutually exclusive and exhaustive events.

### 1.6 A genetic example: The ABO blood types.

We can be blood type  $A$ ,  $B$ ,  $AB$  or  $O$ .

Let our “experiment” be finding the blood types of two children in a family.

Then  $\Omega = \{(i, j); i, j = A, B, AB, O\}$ .

Let  $E_1$  be the first child is type  $A$ .

Let  $E_2$  be the first child is type  $O$ .

Let  $E_3$  be the second child is type  $B$ .

Let  $E_4$  be at least one child is type  $A$ .

Let  $E_5$  be at most one child is type  $A$ .

Let  $E_6$  be the two children have the same blood type.

Let  $E_7$  be the two children have different blood types.

Which pairs of events are *complements*?

Which pairs of events are *disjoint*?

Which pair of events is *mutually exclusive and exhaustive*?

What is the intersection of  $E_4$  and  $E_5$ ?

## Lecture 2: Probabilities of Events: Larson & Marx 2.3; Sept 28, 2018 (a)

### 2.1 Probability axioms

For each event  $E$  we assume we can assign a number  $P(E)$  which satisfies the following three axioms:

- (i)  $P(E) \geq 0$  for every event  $E$ .
- (ii)  $P(\Omega) = 1$
- (iii) If  $E_1, E_2, \dots$  are *mutually exclusive*  $P(E_1 \cup E_2 \cup E_3 \cup \dots) = P(E_1) + P(E_2) + P(E_3) + \dots$

Note: for a countable sample space, each outcome (element of  $\Omega$ ) has a probability, and each event is a union of outcomes, with probability the sum of the probabilities of the outcomes.

### 2.2 Probability interpretation as a limiting frequency

A useful *interpretation* of  $P(E)$  is that it is the proportion of times an outcome in  $E$  occurs in a large number of repetitions of the same experiment with outcomes in the sample space  $\Omega$ .

**Example:** Sampling an individual from a very large population.

$\Omega = \{A, B, AB, O\}$ .

$P(A)$  can be interpreted as the proportion of  $A$  blood-type individuals in the population. If we repeat the sampling of an individual again, and again, the proportion of times we observe the individual to have blood type  $A$  is  $P(A)$ .

For the USA population, roughly,  $P(A) = 0.41$ ,  $P(B) = 0.16$ ,  $P(AB) = 0.07$ , and  $P(O) = 0.36$ .

$P(\text{antigen A on red blood cells}) = P(\{A\} \cup \{AB\}) = P(A) + P(AB) = 0.48$  for this example.

### 2.3 Basic probability formulae

- (i)  $\Omega = E \cup E^c$ ,  $E \cap E^c = \Phi$ , so  $P(E^c) + P(E) = P(\Omega) = 1$ , or  $P(E^c) = 1 - P(E)$ .

This also shows  $P(E) \leq 1$ , since all probabilities are non-negative.

- (ii) If  $E \subset F$ ,  $F = E \cup (F \cap E^c)$ ;  $P(F) = P(E) + P(F \cap E^c) \geq P(E)$ .

- (iii)  $E \cup F = E \cup (E^c \cap F)$ , so  $P(E \cup F) = P(E) + P(E^c \cap F)$ .

So  $P(E \cup F) + P(E \cap F) = P(E) + P(E^c \cap F) + P(E \cap F) = P(E) + P(F)$ ,

or  $P(E \cup F) = P(E) + P(F) - P(E \cap F)$ . **Note:**  $P(E \cup F) \leq P(E) + P(F)$  **always**.

### 2.4 Two important probability formulae

#### (i) Law of total probability

Suppose  $E_1, E_2, \dots$ , form a *partition* of  $\Omega$ . That is,  $E_1, E_2, \dots$  are mutually exclusive and exhaustive.

That is,  $E_i \cap E_j = \Phi$  (disjoint), and  $\Omega = E_1 \cup E_2 \cup \dots$

Then for any event  $F$ ,  $F = \bigcup_i (F \cap E_i)$ ,  $P(F) = \sum_i P(F \cap E_i)$ .

Special case: if  $e_i$  is  $i$ th outcome in a countable  $\Omega$ , and  $E_i = \{e_i\}$ ,

$$F \cap E_i = E_i \text{ or } F \cap E_i = \Phi, \text{ and } P(F) = \sum_{e_i \in F} P(E_i).$$

#### (ii) The inclusion and exclusion formula

$$P(D \cup E) = P(D) + P(E) - P(D \cap E).$$

$$P(C \cup D \cup E) = P(C) + P(D) + P(E) - P(C \cap D) - P(D \cap E) - P(C \cap E) + P(C \cap D \cap E).$$

$$\begin{aligned} P(E_1 \cup E_2 \cup \dots \cup E_k) &= P(E_1) + P(E_2) + \dots + P(E_k) \\ &\quad - P(E_1 \cap E_2) - \text{all the other 2-way} \\ &\quad + P(E_1 \cap E_2 \cap E_3) + \text{all the other 3-way} \\ &\quad - P(E_1 \cap E_2 \cap E_3 \cap E_4) - \text{all the other 4-way} \dots \pm P(E_1 \cap E_2 \cap \dots \cap E_k). \end{aligned}$$

**Note:** This sum of positive and negative terms may not be well defined as  $k \rightarrow \infty$ : see 4.3

### Lecture 3: Permutations and combinations: Larson & Marx 2.6; Sept 28, 2018 (b)

#### 3.1 Basic principle of counting

If an experiment has  $k$  steps, and if earlier choices do NOT limit later ones, then if step-1 can be done in  $n_1$  ways, step-2 in  $n_2$  ways, ... step- $k$  in  $n_k$  ways,

then there are  $n_1 \times n_2 \times \dots \times n_k$  possible outcomes for (step-1, ..., step- $k$ ).

**Corollary:** There are  $2^k$  subsets of a set size  $k$ .

**Proof:** Each element  $i$ ,  $i = 1, \dots, k$  can be chosen, or not:  $n_i = 2$ ,  $i = 1, \dots, k$ .

So total possible is  $2 \times 2 \times \dots \times 2 = 2^k$ .

Note: for proper (not  $\Omega$ ), non-empty (not  $\Phi$ ) subsets, there are  $2^k - 2$ .

#### 3.2 Permutations and combinations

(i) The number of ways of ordering  $n$  distinct objects is  $n(n-1)(n-2)\dots 3.2.1 = n!$  ( $n$ -factorial).

(ii) The number of ways of choosing  $k$  distinct objects, in order, from  $n$  is  $n(n-1)\dots(n-k+1) = n!/(n-k)!$ .

(iii) If we do not care about the order in which the  $k$  objects are selected, there are  $k!$  selections that give the same *combination*.

That is there are  $n!/((n-k)k!)$  distinct *combinations*: this is often written  ${}_nC_k$  or  $\binom{n}{k}$ .

(iv) A useful formula: (L & M, P.88)  $\binom{n}{k} = \binom{n-1}{k-1} + \binom{n-1}{k}$

Consider the number of choices that do and do not contain the particular object "Fred".

#### 3.3 The binomial theorem; L&M 2.6 P.87

$$(x+y)^n = \sum_{k=0}^n \binom{n}{k} x^k y^{n-k}$$

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  $x = y = 1$  gives  $\binom{n}{k}$  as the number of size- $k$  subsets of  $n$  objects.

The case of 4.4 (ii) (next page) is a special case with  $x = y = \frac{1}{2}$ .

#### 3.4 Multinomial combinations; L & M Theorem 2.6.2, P.86

Number of ways of arranging  $n_1$  objects type-1,  $n_2$  objects type-2, ...  $n_k$  objects type- $k$ ,

where  $n_1 + n_2 + \dots + n_k = n$ :

Choose the  $n_1$  positions for type 1:  $\binom{n}{n_1} = n!/(n_1!(n-n_1)!)$ .

Now out of the remaining  $(n-n_1)$  positions choose  $n_2$  for type-2:

number of ways =  $\binom{n-n_1}{n_2} = (n-n_1)!/(n_2!(n-n_1-n_2)!)$ . etc. ...

Total number of ways is

$$\frac{n!}{n_1!(n-n_1)!} \frac{(n-n_1)!}{n_2!(n-n_1-n_2)!} \frac{(n-n_1-n_2)!}{n_3!(n-n_1-n_2-n_3)!} \dots \frac{(n-n_1-n_2-\dots-n_{k-1})!}{n_k!0!} = \frac{n!}{n_1! n_2! \dots n_k!}$$

**Example:** Twelve students go to donate blood: 5 are type  $A$ , 2 are type  $B$ , one is  $AB$ , and 4 are type  $O$ . How many different orderings of the types of blood in the 12 blood donation tubes are there?

Answer:  $12!/(5! \times 2! \times 1! \times 4!) = (12.11.10.9.8.7.6)/(2.4.3.2) = 12.11.10.9.7 = 914,760$ .

## 4: Some additional notes on sets, probabilities, and combinatorics

### 4.1 The collection of all events

For (finite or) countable  $\Omega$ , events are all subsets of  $\Omega$ , but this does not work for  $\Omega = \mathfrak{R}$ .

More generally,  $\Omega$  is event,  $E$  an event  $\Rightarrow E^c$  an event, and  $E_1, E_2, \dots$  events  $\Rightarrow \bigcup_{i=1}^{\infty} E_i$  an event.

(Such collections, closed under complements and countable unions, are called  $\sigma$ -fields.)

Note  $\Phi = \Omega^c$  is an event, and  $\bigcap E_i = (\bigcup E_i^c)^c$  are then also events.

### 4.2 Increasing and decreasing sequences of events

(i)  $A_1, A_2, A_3, \dots$  are nested increasing sets if  $A_1 \subset A_2 \subset A_3 \subset \dots$ . Then  $\bigcup_1^n A_i = A_n$  and  $\bigcap_1^n A_i = A_1$ .

(ii)  $A_1, A_2, A_3, \dots$  are nested decreasing sets if  $A_1 \supset A_2 \supset A_3 \supset \dots$ . Then  $\bigcup_1^n A_i = A_1$  and  $\bigcap_1^n A_i = A_n$ .

**Example:** In a sequence of tries (maybe not independent), let  $A_n$  be event of no successes in  $n$  tries, a decreasing sequence. So  $\lim_{n \rightarrow \infty} A_n$  is event of no success ever:  $D_n = A_n^c$  (success by try  $n$ ) is increasing.

**4.3 Nested sets Theorem:** Let  $A_1, A_2, \dots$  be any events in  $\Omega$ .

(i) If  $A_1 \subset A_2 \subset A_3 \subset \dots$ ,  $P(A_1 \cup A_2 \cup A_3 \dots) = \lim_{n \rightarrow \infty} P(A_n)$ .

(ii) If  $A_1 \supset A_2 \supset A_3 \supset \dots$ ,  $P(A_1 \cap A_2 \cap A_3 \dots) = \lim_{n \rightarrow \infty} P(A_n)$ .

**Proof:** (i) Let  $B_i = A_i \cap A_{i-1}^c$ ; Then  $B_i$  are disjoint and  $B_1 \cup B_2 \cup \dots \cup B_n = A_1 \cup A_2 \cup \dots \cup A_n = A_n$ , so  $P(A_1 \cup A_2 \cup A_3 \dots) = P(B_1 \cup B_2 \cup B_3 \dots) = \sum_1^{\infty} P(B_i) = \lim_{n \rightarrow \infty} (\sum_1^n P(B_i)) = \lim_{n \rightarrow \infty} P(B_1 \cup B_2 \cup \dots \cup B_n) = \lim_{n \rightarrow \infty} P(A_n)$ .

(ii) Let  $D_i = A_i^c$ , so from (i)  $P(D_1 \cup D_2 \cup D_3 \dots) = \lim_{n \rightarrow \infty} P(D_n)$ .

But  $P(D_1 \cup D_2 \cup D_3 \dots) = P((A_1 \cap A_2 \cap \dots)^c) = 1 - P(A_1 \cap A_2 \cap \dots)$ .

So  $P(A_1 \cap A_2 \cap \dots) = 1 - P(D_1 \cup D_2 \cup D_3 \dots) = \lim_{n \rightarrow \infty} (1 - P(D_n)) = \lim_{n \rightarrow \infty} P(A_n)$ .

**Continuing above example:** In a sequence of trials, let  $A_n = \{ \text{no success by trial } n \}$  and  $D_n = A_n^c$ .

(i) If trials are **independent**, with constant probability of success  $p > 0$ ,

$P(A_n) = (1-p)^n \rightarrow 0$ . Eventually, with probability 1, a success occurs.

(ii) Suppose the probability of success on try  $k$  is  $p_k$ . Then  $P(D_n) \leq \sum_{k=1}^n p_k$ .

If  $p_k$  decreases fast (e.g.  $p_k = 0.1/k^2$ ) then  $\lim P(D_n) < 1$ ; eventual success is not certain.

### 4.4 Binomial counts and Stirling's formula L&M 2.6, P. 76-77

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

(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  $n$ ,  
$$\binom{n}{n/2} = \frac{n!}{(n/2)!(n/2)!} \approx \frac{\sqrt{2\pi}n^{n+\frac{1}{2}}e^{-n}}{\sqrt{2\pi}(n/2)^{(n/2)+\frac{1}{2}}e^{-(n/2)} \times \sqrt{2\pi}(n/2)^{(n/2)+\frac{1}{2}}e^{-(n/2)}} \\ = (1/\sqrt{2\pi})2^{n+1}n^{n+\frac{1}{2}-(n/2)-\frac{1}{2}-(n/2)-\frac{1}{2}} = (1/\sqrt{2\pi})(2/\sqrt{n})2^n$$

Or  $P((n/2) \text{ A children out of } n) = \binom{n}{n/2}(\frac{1}{2})^n \approx 1/\sqrt{2\pi(n/4)}$

This result will come back in approximating Binomial probabilities by the Normal probability distribution.