

# Probability Review

Survey Sampling  
Statistics 4234/5234  
Fall 2018

September 6, 2018

Example 1: Flip a coin 3 times, the sample space is

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

Definition: The *sample space* for a random experiment is the set of possible outcomes.

Assume  $\Omega$  is finite,  $\Omega = \{\omega_1, \dots, \omega_k\}$ .

Associated with each outcome  $\omega_i$  is a probability  $p_i$  satisfying

$$p_i \geq 0 \text{ for } i = 1, \dots, k \quad \text{and} \quad \sum_{i=1}^k p_i = 1$$

Definition: A collection of outcomes (any subset of the sample space) is called an *event*. The probability of an event is the sum of the probabilities of the outcomes that make up that event.

Example 1:  $p_i = \frac{1}{8}$  for  $i = 1, \dots, 8$ . Define the event  $A$  to be “exactly two heads” so  $A = \{HHT, HTH, THH\}$  and  $P(A) = \frac{3}{8}$ .

### Simple random sampling with replacement (SRSwR)

$N$  balls in urn, sample one ball  $n$  times, replacing the ball in the urn between each draw.

There are  $N^n$  possible ordered samples (permutations), each equally likely.

Example 2: Population of  $N = 5$  units, sample  $n = 2$  with replacement. The probability that unit 5 is in the sample is

$$P(\{15, 25, 35, 45, 51, 52, 53, 54, 55\}) = \frac{9}{25} = 0.36$$

## Simple random sampling without replacement (SRS)

$N$  balls in urn, draw  $n$  of them at random.

There are

$$N \times (N - 1) \times \cdots \times (N - n + 1) = \frac{N!}{(N - n)!}$$

possible ordered samples (permutations).

Ignoring the order we have

$$\frac{N!}{n!(N - n)!} = \binom{N}{n}$$

possible samples, equally likely.

Example 2: With  $N = 5$  and  $n = 2$  there are  $\binom{5}{2} = 10$  possible samples. The probability that unit 5 is in our sample is

$$P(\{15, 25, 35, 45\}) = \frac{4}{10} = 0.40$$

Example 3: An urn has 5 black balls and 3 red ones, we will draw 4 at random. Then

$$P(\text{no red}) = \frac{5 \times 4 \times 3 \times 2}{8 \times 7 \times 6 \times 5} = \frac{\binom{5}{4} \binom{3}{0}}{\binom{8}{4}} = \frac{1}{14}$$

and

$$P(\text{one red}) = \frac{\binom{5}{3} \binom{3}{1}}{\binom{8}{4}} = \frac{3}{7} \quad \text{and} \quad P(\text{two reds}) = \frac{\binom{5}{2} \binom{3}{2}}{\binom{8}{4}} = \frac{3}{7}$$

## Random variables

A *random variable* assigns a numeric value to each outcome,

$$X : \Omega \rightarrow \mathbb{R}$$

The set of possible values and their probabilities is the *probability distribution* of the random variable.

Example 3: Urn contains 5 black and 3 red balls, pick 4 at random, let  $X$  = the number of reds.

$x$	0	1	2	3
$P(X = x)$	1/14	6/14	6/14	1/14

Definition: The *expected value* of the random variable  $X$  is

$$E(X) = \sum_x xP(X = x)$$

Example 3:  $E(X) = 1.5$

Proposition: For any random variable  $X$  and function  $g(\cdot)$ , the mean of the random variable  $g(X)$  is

$$E[g(X)] = \sum_x g(x)P(X = x)$$

Definition: The *variance* of the random variable  $X$  is

$$V(X) = E[(X - EX)^2]$$

Example 3:  $V(X) = 0.5357$

Proposition: An alternative expression for the variance is

$$V(X) = E(X^2) - (EX)^2$$

## Joint distributions

Example 4: Let the random variables  $(X, Y)$  have the joint probability distribution given by the following table of  $P(X = x, Y = y)$ .

$x$	$y$		
	1	2	3
1	$1/6$	$1/6$	$1/6$
2	$1/12$	0	$1/12$
3	0	$1/3$	0

Proposition: Given the pair of random variables  $(X, Y)$ , and a function  $g(\cdot, \cdot)$ , the expected value of the random variable  $g(X, Y)$  is

$$E[g(X, Y)] = \sum_x \sum_y g(x, y) P(X = x, Y = y)$$



Example 4: Find  $E(X)$  and  $E(Y)$  and  $E(XY)$ .

OK.

$$E(X) = 1 \left( \frac{1}{2} \right) + 2 \left( \frac{1}{6} \right) + 3 \left( \frac{1}{3} \right) = \frac{11}{6}$$

and  $E(Y) = 2$ , by inspection, and

$$E(XY) = 1 \left( \frac{1}{6} \right) + 2 \left( \frac{1}{4} \right) + 3 \left( \frac{1}{6} \right) + 6 \left( \frac{5}{12} \right) = \frac{11}{3}$$

Definition: The *covariance* between the random variables  $X$  and  $Y$  is

$$\text{Cov}(X, Y) = E[(X - EX)(Y - EY)]$$

and the *correlation* is

$$\text{Corr}(X, Y) = \frac{\text{Cov}(X, Y)}{\sqrt{V(X) \cdot V(Y)}}$$

Proposition:  $\text{Cov}(X, Y) = E(XY) - (EX)(EY)$ .

Definition: The random variables  $X$  and  $Y$  are *independent* if and only if

$$P(X = x, Y = y) = P(X = x)P(Y = y)$$

for all  $x$  and  $y$ .

Proposition: If  $X$  and  $Y$  are independent then  $\text{Cov}(X, Y) = 0$ .

The converse of this proposition is not true (see Example 4).

## Conditional probability

Definition: The *conditional probability* of  $B$  given  $A$ , where  $P(A) > 0$ , is

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

Proposition: If the events  $A_1, \dots, A_k$  form a partition of the sample space, that is if  $A_1 \cup \dots \cup A_k = \Omega$  and  $A_i \cap A_j = \emptyset$  for all  $i \neq j$ , then

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

Example 5: Suppose we have three urns,  $U_1 = \{4, 6, 7, 9\}$  and  $U_2 = \{6, 8\}$  and  $U_3 = \{5\}$ . We first randomly select an urn, then randomly select a number from that urn.

Then

$$P(4) = P(U_1)P(4|U_1) + 0 + 0 = \left(\frac{1}{3}\right) \left(\frac{1}{4}\right) = \frac{1}{12}$$

and

$$P(5) = 0 + 0 + P(U_3)P(5|U_3) = \frac{1}{3}(1) = \frac{1}{3}$$

and

$$\begin{aligned} P(6) &= P(U_1)P(6|U_1) + P(U_2)P(6|U_2) \\ &= \left(\frac{1}{3}\right) \left(\frac{1}{4}\right) + \left(\frac{1}{3}\right) \left(\frac{1}{2}\right) = \frac{1}{4} \end{aligned}$$

## Conditional expectation

Given discrete random variables  $X$  and  $Y$ , and a number  $x$  with  $P(X = x) > 0$ , the *conditional distribution* of  $Y$  given  $X = x$  is defined by

$$P(Y = y|X = x) = \frac{P(X = x, Y = y)}{P(X = x)}$$

The *conditional expected value* of  $Y$  given  $X = x$  is

$$E(Y|X = x) = \sum_y yP(Y = y|X = x)$$

and the *conditional variance* of  $Y$  given  $X = x$  is

$$V(Y|X = x) = E \left\{ [Y - E(Y|X = x)]^2 | X = x \right\}$$

For each value of  $x$  with  $P(X = x) > 0$  we can define the functions

$$g(x) = E(Y|X = x) \quad \text{and} \quad h(x) = V(Y|X = x)$$

and thus have defined the random variables

$$g(X) = E(Y|X) \quad \text{and} \quad h(X) = V(Y|X)$$

which have some interesting properties:

1.  $E(Y) = E[E(Y|X)]$
2.  $V(Y) = V[E(Y|X)] + E[V(Y|X)]$

Example 5: Define the random variables  $X$  and  $Y$  so that  $X$  indicates the urn selected (1, 2 or 3), and  $Y$  equals the number drawn from that urn. It is straightforward to show that  $Y$  is distributed as

$y$	4	5	6	7	8	9
$P(Y = y)$	1/12	4/12	3/12	1/12	2/12	1/12

and compute

$$E(Y) = 6.17 \quad \text{and} \quad V(Y) = 2.14$$

The conditional distributions  $P(Y = y|X = x)$  for  $x = 1, 2, 3$  are given in the following table, from which are easily computed the conditional means and variances  $E(Y|X = x)$  and  $V(Y|X = x)$ :

$x$	4	5	6	7	8	9	$E$	$V$
1	1/4	0	1/4	1/4	0	1/4	6.50	3.25
2	0	0	1/2	0	1/2	0	7.00	1.00
3	0	1	0	0	0	0	1.00	0.00

Then we have

$$\begin{aligned}
 E[E(Y|X)] &= \sum_{x=1}^3 E(Y|X = x)P(X = x) \\
 &= \frac{1}{3}(6.5 + 7.0 + 5.0) = 6.17
 \end{aligned}$$

confirming Property 1 above.



Also

$$\begin{aligned} E[V(Y|X)] &= \sum_{x=1}^3 V(Y|X = x)P(X = x) \\ &= \frac{1}{3}(3.25 + 1.00 + 0.00) = 1.4167 \end{aligned}$$

and

$$\begin{aligned} V[E(Y|X)] &= \sum_{x=1}^3 [E(Y|X = x)]^2 P(X = x) - (EY)^2 \\ &= \frac{1}{3}(6.5^2 + 7.0^2 + 5.0^2) - 6.17^2 = 0.7222 \end{aligned}$$

and thus

$$V[E(Y|X)] + E[V(Y|X)] = 2.14$$

confirming Property 2.