# 11 Lecture 11: Feb 12

#### Last time

- Multiple correlation
- Confidence intervals and hypothesis tests
- R practice with questions

# Today

- R practice with questions
- HW1 deadline extends to 11:59 pm Sunday, Feb 14
- Probability review

#### Reference:

- Statistical Inference, 2nd Edition, by George Casella & Roger L. Berger
- Review of Probability Theory by Arian Maleki and Tom Do

# Probability theory review

A few basic elements to define a probability on a set:

- Sample space S is the set that contains all possible outcomes of a particular experiment.
- An **event** is any collection of possible outcomes of an experiment, that is , any subset of S (including S itself).
- Event operations
  - 1. Union: The union of A and B, written  $A \cup B$ , is the set of elements that belong to either A or B or both:

$$A \cup B = \{x : x \in A \text{ or } x \in B\}$$

2. Intersection: The intersection of A and B, written  $A \cap B$ , is the set of elements that belong to both A and B:

$$A \cap B = \{x : x \in A \text{ and } x \in B\}$$

3. Complementation: The complement of A, written as  $A^c$ , is the set of all elements that are not in A:

$$A^c = \{x : x \notin A\}.$$

- Sigma algebra (or Borel field): A collection of subsets of S is called a sigma algebra (or Borel field), denoted by  $\mathcal{B}$ , if it satisfies the following three properties:
  - 1.  $\emptyset \in \mathcal{B}$  (the empty set is an element of  $\mathcal{B}$ )
  - 2. If  $A \in \mathcal{B}$ , then  $A^c \in \mathcal{B}$  ( $\mathcal{B}$  is closed under complementation).
  - 3. If  $A_1, A_2, \dots \in \mathcal{B}$ , then  $\bigcup_{i=1}^{\infty} A_i \in \mathcal{B}$  ( $\mathcal{B}$  is closed under countable unions).
- Axioms of probability: Given a sample space S and an associated sigma algebra  $\mathcal{B}$ , a probability function is a function Pr() with domain  $\mathcal{B}$  that satisfies
  - 1.  $Pr(A) \ge 0$  for all  $A \in \mathcal{B}$
  - 2. Pr(S) = 1.
  - 3. If  $A_1, A_2, \dots \in \mathcal{B}$  are pairwise disjoint, then  $\Pr(\bigcup_{i=1}^{\infty} A_i) = \sum_{i=1}^{\infty} \Pr(A_i)$ .

## Properties:

If Pr() is a probability function and A and B are any sets in  $\mathcal{B}$ , then

- $Pr(\emptyset) = 0$ , where  $\emptyset$  is the empty set *Proof:*
- $Pr(A) \leq 1$ Proof:
- $Pr(A^c) = 1 Pr(A)$ Proof:
- $\Pr(B \cap A^c) = \Pr(B) \Pr(A \cap B)$ Proof:
- $Pr(A \cup B) = Pr(A) + Pr(B) Pr(A \cap B)$ Proof:
- $Pr(A \cup B) = Pr(A) + Pr(B \cap A^c) = Pr(A) + Pr(B) Pr(A \cap B)$
- If  $A \subset B$ , then  $Pr(A) \leq Pr(B)$ . Proof:

# Conditional probability

Definition: If A and B are events in S, and Pr(B) > 0, then the conditional probability of A given B, written Pr(A|B), is

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

Note that what happens in the conditional probability calculation is that B becomes the sample space:  $\Pr(B|B) = 1$ , in other words,  $\Pr(A|B)$  is the probability measure of the event A after observing the occurrence of event B.

Definition: Two events A and B are statistically independent if  $Pr(A \cap B) = Pr(A) Pr(B)$ . When A and B are independent events, then Pr(A|B) = Pr(A) and the following pairs are also independent

- A and  $B^c$  proof:
- $A^c$  and B
- $A^c$  and  $B^c$

### Random variables

Definition: A random variable is a function from a sample space S into the real numbers.

| Experiment                 | Random variable                  |
|----------------------------|----------------------------------|
| Toss two dice              | X = sum of the numbers           |
| Toss a coin 25 times       | X = number of heads in 25 tosses |
| Apply different amounts of |                                  |
| fertilizer to corn plants  | X = yield/acre                   |

Suppose we have a sample space

$$S = \{s_1, \dots, s_n\}$$

with a probability function Pr and we define a random variable X with range  $\mathcal{X} = \{x_1, \ldots, x_m\}$ . We can define a probability function  $\Pr_X$  on  $\mathcal{X}$  in the following way. We will observe  $X = x_i$  if and only if the outcome of the random experiment is an  $s_j \in S$  such that  $X(s_j) = x_i$ . Thus,

$$\Pr_X(X = x_i) = \Pr(\{s_j \in S : X(s_j) = x_j\}).$$

We will simply write  $Pr(X = x_i)$  rather than  $Pr_X(X = x_i)$ .

A note on notation: Randon variables are often denoted with uppercase letters and the realized values of the variables (or its range) are denoted by corresponding lowercase letters.

#### Distribution functions

Definition: The <u>cumulative distribution function</u> or <u>cdf</u> of a random variable (r.v.) X, denoted by  $F_X(x)$  is defined by

$$F_X(x) = \Pr(X \leq x)$$
, for all  $x$ .

The function F(x) is a cdf if and only if the following three conditions hold:

- 1.  $\lim_{x\to\infty} F(x) = 1.$
- 2. F(x) is a nondecreasing function of x.
- 3. F(x) is right-continuous; that is, for every number  $x_0$ ,  $\lim_{x\downarrow x_0} = F(x_0)$ .

Definition: A random variable X is <u>continuous</u> if F(x) is a continuous function of x. A random variable X is discrete if F(x) is a step function of x.

The following two statements are equivalent:

- 1. The random variables X and Y are identically distributed.
- 2.  $F_X(x) = F_Y(x)$  for every x.

#### Density and mass functions

Definition: The probability mass function (pmf) of a discrete random variable X is given by

$$f_X(x) = \Pr(X = x)$$
 for all  $x$ .

Example (Geometric probabilities) For the geometric distribution, we have the pmf

$$f_X(x) = \Pr(X = x) = \begin{cases} p(1-p)^{x-1} & \text{for } x = 1, 2, \dots \\ 0 & \text{otherwise.} \end{cases}$$

Definition: The probability density function or  $\underline{pdf}$ ,  $f_X(x)$ , of a continuous random variable X is the function that satisfies

$$F_X(x) = \int_{-\infty}^x f_X(t)dt$$
 for all  $x$ .

A note on notation: The expression "X has a distribution given by  $F_X(x)$ " is abbreviated symbolically by " $X \sim F_X(x)$ ", where we read the symbol " $\sim$ " as " is distributed as".

Example (Logistic distribution) For the logistic distribution, we have

$$F_X(x) = \frac{1}{1 + e^{-x}}$$

and, hence,

$$f_X(x) = \frac{d}{dx} F_X(x) = \frac{e^{-x}}{(1 + e^{-x})^2}.$$

A function  $f_X(x)$  is a pdf (or pmf) of a random variable X if and only if

- 1.  $f_X(x) \ge 0$  for all x
- 2.  $\sum_{x} f_X(x) = 1 \ (pmf)$  or  $\int_{-\infty}^{\infty} f_X(x) dx = 1 \ (pdf)$ .