Lecture 12: Sept 19

Last time

- Random variables
- Distribution Functions
- Types of Random Variables

Today

- Counting Techniques
- Transformations of Random Variables

Definition The number of r-tuples we can make $r \leq n$, using n different symbols (each only once), is called the number of permutations of n things r at a time and is denoted by ${}^{n}P_{r}$ which is calculated as

$${}^{n}P_{r}=n(n-1)\cdots(n-r+1).$$

Example Fifteen cars enter a race. In how many different ways could trophies for first, second, and third place be awarded? *Solutions:*

Example How many of the 3-tuples just counted have car number 15 in the first position? *Solutions:*

Definition The number of distinct subsets, each of size r, that can be constructed from a set with n elements is called the number of *combinations* of n things r at a time: this number is represented by $\binom{n}{r}$ which reads n choose r.

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

Example How many distinct 5-card hands can be dealt from a standard 52-card deck?

$$\binom{5}{2}5 = \frac{52!}{5!47!} = 2,598,960.$$

Theorem If x and y are any two real numbers and n is a positive integer, then

$$(x+y)^n = \sum_{i=1}^n \binom{n}{i} x^i y^{n-i}, \quad \text{where } \binom{n}{i} = \frac{n!}{(n-i)!i!}.$$

Transformations of Random Variables

Theorem If X is a r.v. with sample space $\mathcal{X} \subset \mathbb{R}$ and cdf $F_X(x)$, then any function of X, say Y = g(X) is also a random variable. The new random variable Y has a new sample space $\mathcal{Y} = g(X) \subset \mathbb{R}$. The objective is to find the cdf $F_Y(y)$ of Y.

Probability mapping: For any set $A \subset \mathcal{Y}$:

$$Pr(Y \in A) = Pr(g(X) \in A)$$
$$= Pr(\{x \in \mathcal{X} : g(x) \in A\})$$
$$= Pr(X \in g^{-1}(A)),$$

where we have defined

$$g^{-1}(A) = \{x \in \mathcal{X} : g(x) \in A\}.$$

Notice that $g^{-1}(A)$ is well defined even if $g(\cdot)$ is not necessarily bijective.

Example (Binomial transformation) A discrete random variable X has a binomial distribution if its pmf is of the form

$$f_X(x) = \Pr(X = x) = \binom{n}{x} p^x (1-p)^{n-x}, \quad x = 0, 1, \dots, n,$$

where n is a positive integer and $0 \le p \le 1$. Values such as n and p that can be set to different values, producing different probability distributions, are called *parameters*. Consider a random variable Y = g(X), where g(x) = n - x; that is, Y = n - X. Here $\mathcal{X} = \{0, 1, ..., n\}$ and $\mathcal{Y} = \{y : y = g(x), x \in \mathcal{X}\} = \{0, 1, ..., n\}$. For any $y \in \mathcal{Y}$, n - x = g(x) = y if and only if x = n - y. Therefore, $g^{-1}(y) = n - y$ and

$$f_Y(y) = \sum_{x \in g^{-1}(y)} f_X(x)$$

$$= f_X(n - y)$$

$$= \binom{n}{n} - yp^{n-y}(1 - p)^{n - (n - y)}$$

$$= \binom{n}{y} (1 - p)^y p^{n - y}.$$

Therefore, Y also has a binomial distribution, but with parameters n and 1-p.

Example (exercise 2.3) Suppose X has the geometric pmf $f_X(x) = \frac{1}{3}(\frac{2}{3})^x$, $x = 0, 1, 2, \ldots$. Determine the probability distribution of Y = X/(X+1). Note that here both X and Y are discrete random variables. To specify the probability distribution of Y, specify its pmf. Solution:

Theorem Suppose a continuous random variable X has cdf $F_X(x)$, let Y = g(X), and let \mathcal{X} and \mathcal{Y} be defined as

$$\mathcal{X} = \{x : f(x) > 0\}$$
 and $\mathcal{Y} = \{y : y = g(x) \text{ for some } x \in \mathcal{X}\}.$

Then,

- 1. If g is an increasing function on \mathcal{X} , $F_Y(y) = F_X(g^{-1}(y))$ for $y \in \mathcal{Y}$.
- 2. If g is a decreasing function on \mathcal{X} , $F_Y(y) = 1 F_X(g^{-1}(y))$ for $y \in \mathcal{Y}$.

Proof: We start with

$$F_Y(y) = \Pr(Y \le y)$$

= $\Pr(g(X) \le y)$