

Definition 1.1

The **mean** of a sample of n measured responses $y_1 y_2 \dots y_n$ is given by

$$\bar{y} = \frac{1}{n} \sum_{i=1}^n (y_i)$$

The corresponding population mean is denoted μ .

Definition 1.2

The **variance** of a sample of measurements $y_1 y_2 \dots y_n$ is the sum of the square of the differences between the measurements and their mean, divided by $n - 1$. . Symbolically, the sample variance is

$$s^2 = \frac{1}{n-1} \sum_{i=1}^n (y_i - \bar{y})^2$$

The corresponding population variance is denoted by the symbol σ^2 .

Definition 1.3

The **standard deviation** of a sample of measurements is the positive square root of the variance; that is,

$$s = \sqrt{s^2}$$

The corresponding population standard deviation is denoted by $\sigma = \sqrt{\sigma^2}$

Definition 2.7

An ordered arrangement of r distinct objects is called a **permutation**. The number of ways of ordering n distinct objects taken r at a time will be designated by the symbol P_n^r

$$P_n^r = n (n - 1) (n - 2) \dots (n - r + 1) = \frac{n!}{(n-r)!}$$

Definition 2.8

The number of combinations of n objects taken r at a time is the number of subsets, each of size r , that can be formed from the n objects. This number will be denoted by C_r^n or $\binom{n}{r}$.

$$P_n^r = n(n-1)(n-2)\cdots(n-r+1) = \frac{n!}{(n-r)!}$$

Definition 2.9

The **conditional probability** of an event A, given that an event B has occurred, is equal to

$$P(A|B) = \frac{P(A \cap B)}{P(B)}, \text{ provided } P(B) > 0. \text{ [The symbol } P(A|B) \text{ is read "probability of A given B."}]$$

Definition 3.7

A random variable Y is said to have a **binomial distribution** based on n trials with success probability p if and only if $P(y) = \binom{n}{y} p^y q^{n-y}$, $y = 0, 1, 2, \dots, n$ and $0 \leq p \leq 1$.

Definition 3.8

A random variable Y is said to have a **geometric probability distribution** if and only if

$$P(y) = q^{y-1} p, \quad y = 1, 2, 3, \dots, \quad 0 \leq p \leq 1.$$

Definition 3.10

A random variable Y is said to have a **hypergeometric probability distribution** if and only if

$$P(y) = \frac{\binom{r}{y} \binom{N-r}{n-y}}{\binom{N}{n}}$$

where y is an integer 0, 1, 2, ... n, subject to the restrictions $y \leq r$ and $n - y \leq N - r$.

Definition 3.11

A random variable Y is said to have a **Poisson probability distribution** if and only if $P(y) = \frac{\lambda^y}{y!} e^{-\lambda}$, $y = 0, 1, 2, \dots$ $\lambda > 0$.

Definition 3.14

Tchebysheff's Theorem Let Y be a random variable with mean μ and finite variance σ^2 . Then, for any constant $K > 0$, $P(|Y - \mu| < K\sigma) > 1 - \frac{1}{K^2}$ or $P(|Y - \mu| \geq K\sigma) \leq \frac{1}{K^2}$

Definition 4.1

Let Y denote any random variable. **The distribution function** of Y , denoted by $F(y)$. Is such that $F(y) = P(Y < y)$ for $-\infty < y < \infty$.

Definition 4.2

A random variable Y with distribution function $F(y)$ is said to be **continuous** if $F(y)$ is continuous, for $-\infty < y < \infty$.

Definition 4.3

Let $F(y)$ be the distribution for a continuous random variable Y . Then $f(y)$, given by

$$f(y) = \frac{dF(y)}{dy} = F'(y)$$

Wherever the derivative exists, is called the **probability function** for the random variable y .

Theorem 4.2

Property of Density Function If $f(y)$ is a density function for a **continuous random variable**, then

1. $f(y) \geq 0$ for all y , $-\infty < y < \infty$.

2. $\int_{-\infty}^{\infty} f(y) dy = 1$.

Definition 4.6

If $y_1 < y_2$ a random variable Y is said to have a continuous **uniform probability distribution** on the interval (y_1, y_2) if and only if density function Y is

$$f(y) = \begin{cases} \frac{1}{y_2 - y_1}, & y_1 < y < y_2 \\ 0, & \text{elsewhere} \end{cases}$$

Theorem 4.6 (Uniform Distribution)

If $x_1 < x_2$ and Y is a random variable uniformly distributed on the interval (x_1, x_2) , then

$$\mu = E(Y) = \frac{x_1 + x_2}{2} \text{ and } \sigma^2 = V(Y) = \frac{(x_2 - x_1)^2}{12}$$