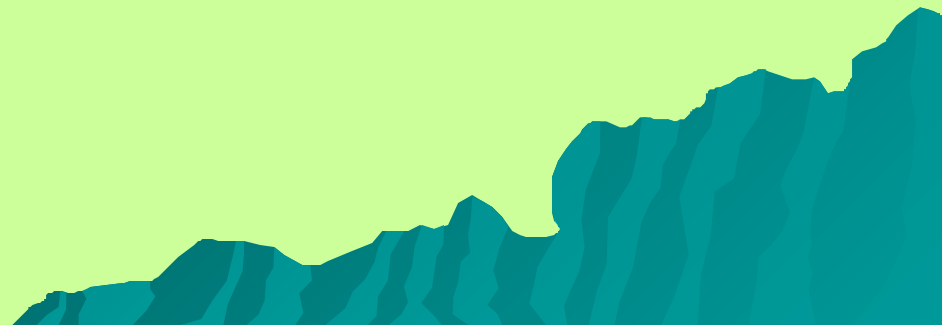


Lecture # 3

Elements of probability



Definition 1

A simple event is an outcome of an experiment that can not be decomposed into a simpler outcome

Example 1. Tossing 1 fair coin

$\{\omega_1, \omega_2\}$ $\omega_1 \rightarrow \text{head (H)}$
 $\omega_2 \rightarrow \text{tail (T)}$

Example 2. Tossing 3 fair coins at once

$\{\omega_1, \omega_2, \omega_3, \omega_4, \omega_5, \omega_6, \omega_7, \omega_8\}$

$\omega_1 \rightarrow \text{HHH}$ $\omega_5 \rightarrow \text{TTT}$

$\omega_2 \rightarrow \text{TTH}$ $\omega_6 \rightarrow \text{HHT}$

$\omega_3 \rightarrow \text{THT}$ $\omega_7 \rightarrow \text{HTH}$

$\omega_4 \rightarrow \text{HTT}$ $\omega_8 \rightarrow \text{TTH}$

A random variable $X(\omega)$ is a function of ω , e.g., $X(\omega)$ is the number of heads in one trial, hence,

$$\begin{aligned} X(\omega_1) &= 3, & X(\omega_6) &= X(\omega_7) = X(\omega_8) = 2, \\ X(\omega_2) &= X(\omega_3) = X(\omega_4) = 1, & X(\omega_5) &= 0 \end{aligned}$$

Definition 2

An event is a collection of one or more simple events

Example 3.

A – 2 Heads in one toss

$$A = \{\omega_6, \omega_7, \omega_8\}$$

B – More than one H in a trial

$$B = \{\omega_1, \omega_6, \omega_7, \omega_8\}$$

C – Exactly 2 tails (T)

$$C = \{\omega_2, \omega_3, \omega_4\}$$

D – At least one H or one T

$$D = \{\omega_1, \omega_2, \omega_3, \omega_4, \omega_5, \omega_6, \omega_7, \omega_8\}$$

Definition 3

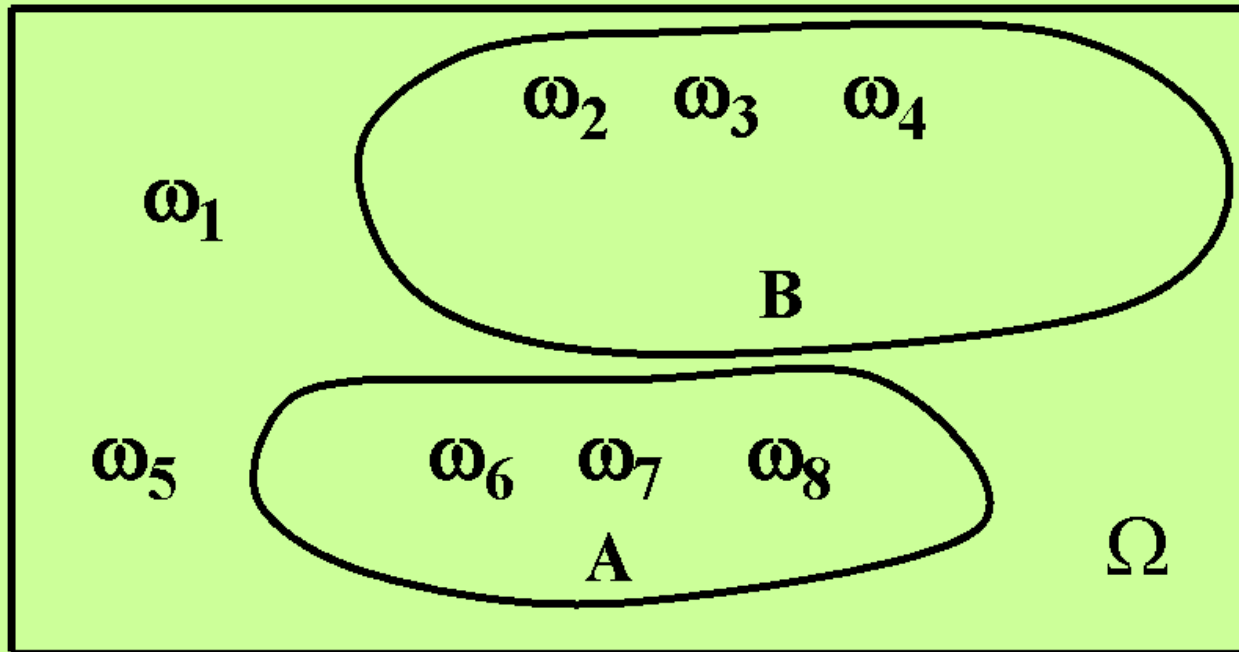
The sample space of an experiment is the collection of all possible simple events

Denote it by Ω

For example 1: $\Omega = \{\omega_1, \omega_2\}$

For example 2 : $\Omega = \{\omega_1, \omega_2, \dots, \omega_8\}$

Venn diagram



Definition 4.

The probability p of a simple event is a number that measures the likelihood that the event will occur when the experiment is performed.

I. Classical definition of probabilities. If one has n equiprobable simple events $\omega_1, \dots, \omega_n$, then the probability $p(A)$ of $A = \{\omega_{i_1}, \omega_{i_2}, \dots, \omega_{i_m}\}$ containing m simple events is

$$P(A) = \frac{m}{n}$$

Evidently, that $P(\Omega) = 1$.

II. We may take p being equal to the relative frequency f_i/n of a simple i th event if n is very large.

III. We may select p based on a priori knowledge of a situation under study.

The case III is more frequently occurred. We guess or formulate a hypothesis H concerning p and, using statistics, test this hypothesis based on realizations (x_1, \dots, x_n) of a sample $X_n = \{X_1, \dots, X_n\}$.

Definition 5

The probability $P(A)$ of an event A is calculated by summing the probabilities of the simple events belonging to A .

Steps for calculating probabilities

- ◆ 1. List the simple events
- ◆ 2. Assign probabilities to simple events
- ◆ 3. Determine the number of simple events containing in the event
- ◆ 4. Sum the simple event probabilities to obtain the event probability

For example 2

$$p(\omega_i) = 1/8, \quad i = \overline{1,8}$$

$$p(A) = p(2H) = p\{\omega_6, \omega_7, \omega_8\} = \frac{1}{8} + \frac{1}{8} + \frac{1}{8} = \frac{3}{8}$$

$$p(B) = P(\text{more than 1H}) =$$

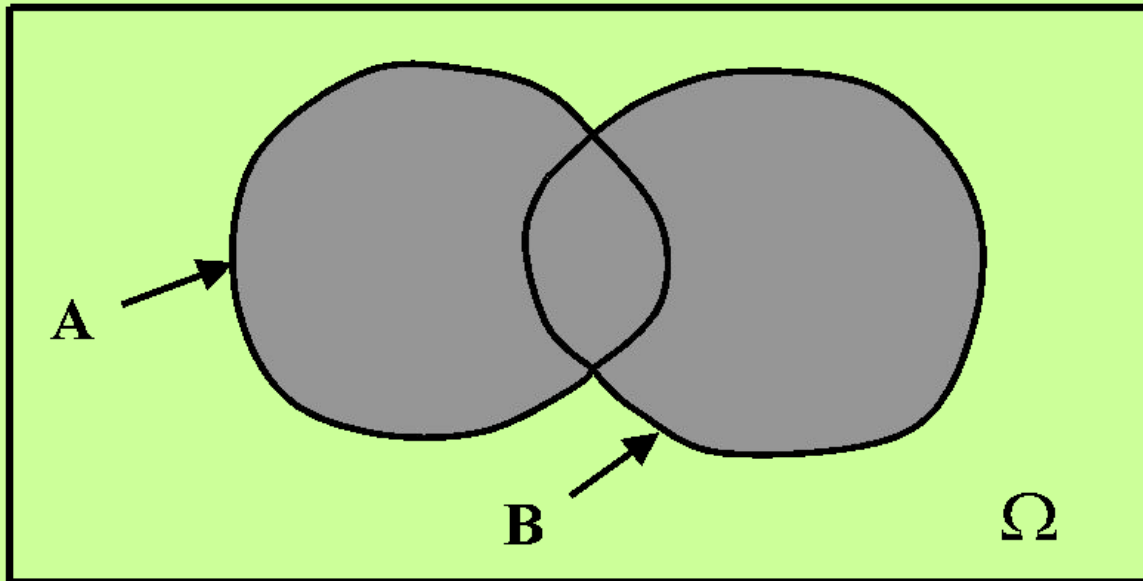
$$= p\{\omega_1, \omega_6, \omega_7, \omega_8\} = \frac{1}{8} + \frac{1}{8} + \frac{1}{8} + \frac{1}{8} = \frac{1}{2}$$

$$p(D) = p(\Omega) = 1 \text{ (certain event)}$$

Compound events

Definition 6

The union of A & B is the event that occurs if either A or B or both occur ($A \cup B$)



$\omega_i, i = \overline{1, n}$, belong A or B or to both of them

Definition 7

The intersection of A & B is the event that occurs if both A and B occur ($A \cap B$)

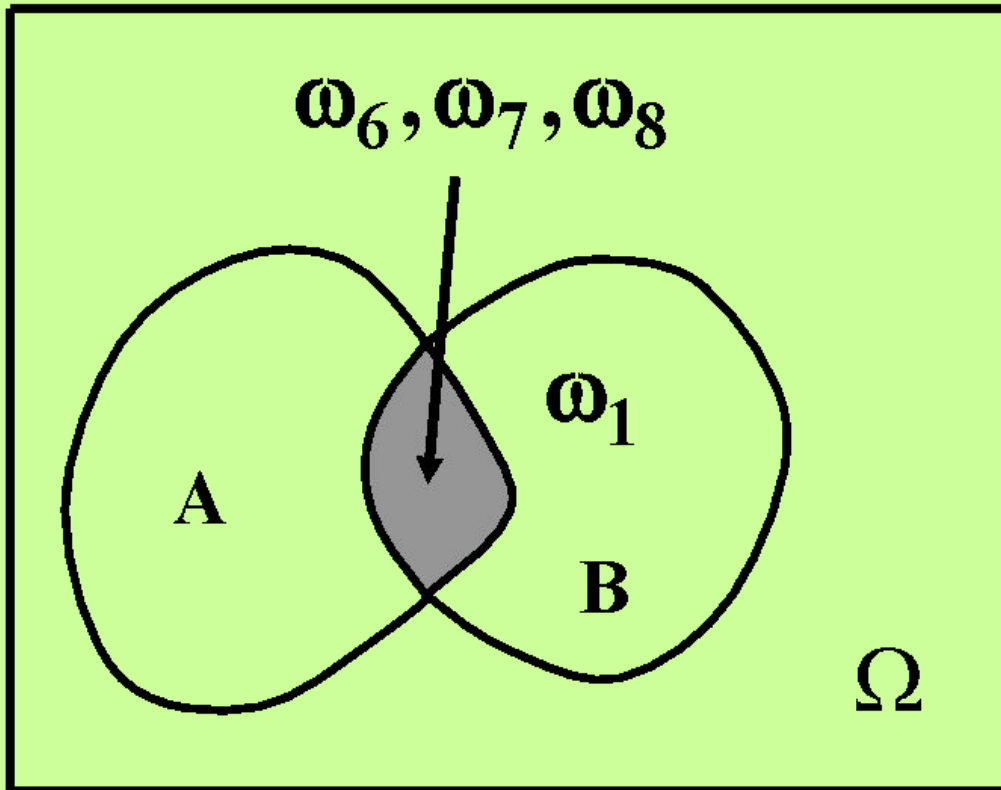
ω_i , belong to both A and B

\bar{A} - the complement event to A

ω_i of \bar{A} do not belong to A!

Example 3

$$A = \{\omega_6, \omega_7, \omega_8\}, \quad B = \{\omega_1, \omega_6, \omega_7, \omega_8\}$$



$$A \cap B = \{\omega_6, \omega_7, \omega_8\}$$

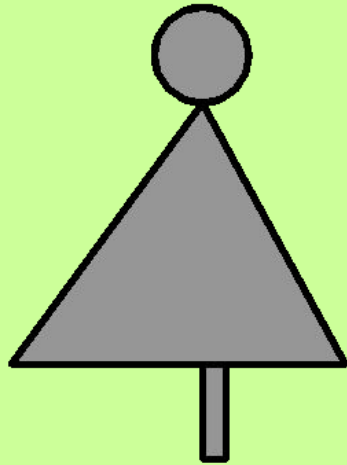
$$A \cup B = \{\omega_1, \omega_6, \omega_7, \omega_8\}$$

Let a set $\Omega = \{\omega_1, \dots, \omega_n\}$ be given, then all possible events (combinations of ω_i 's) $\cup \Omega \cup \emptyset$ (empty set, corresponding to the improbable event) is algebra S of events

If two events A & B are mutually exclusive (incompatible), then

$$P(A \cap B) = P(A) + P(B).$$

Suppose we have a target. A shooter produced a shot.



event

A \Rightarrow to hit the target

B \Rightarrow fail to hit

A&B are incompatible

$$P(A) + P(B) = 1$$

A \cap B is the certain

Axioms of probability

For every Ω (a set of all possible simple events) and S (algebra of events) we postulate

1) Axiom 1. For every $A \in S$ $P(A) \geq 0$

2) Axiom 2. $P(\Omega) = 1$ Ω -is certain event

3) Axiom 3. If $A \in S$ and $B \in S$ are incompatible, then

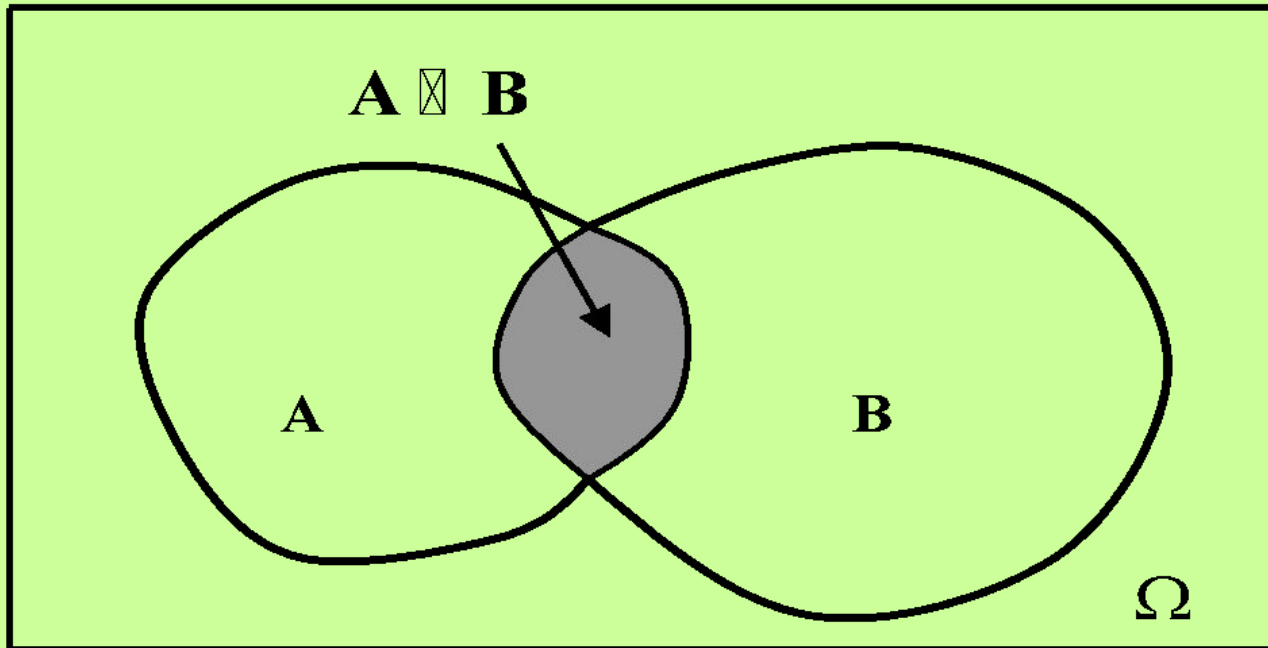
$$P(A \boxplus B) = P(A) + P(B).$$

The following two rules are consequences of axioms:

Additive rule

$$P(A \cup B) = P(A) + P(B) - P(A \cap B)$$

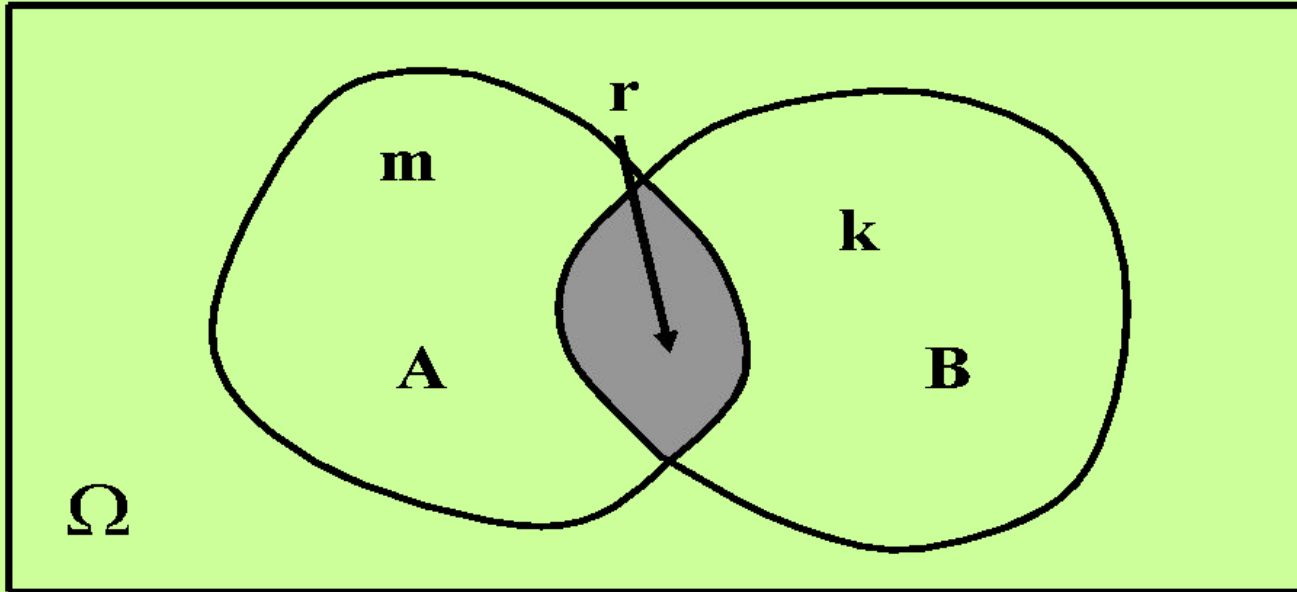
Multiplicative rule



$$P(A \cap B) = P(A | B)P(B)$$

$$P(A \cap B) = P(B | A)P(A)$$

The conditional probability



A contains m elements of Ω

B contains k elements of Ω

$A \cap B$ contains r elements of Ω

$\Omega \in \{\omega_1, \dots, \omega_n\}$, ω_i th are equiprobable

$$P(A \cap B) = \frac{r}{n} \quad P(B) = \frac{k}{n}$$

Let the event B occurred. What is the prob. for A to occur?

$$P(A | B) = \frac{r}{k} = \frac{r/n}{k/n} = \frac{P(A \cap B)}{P(B)}$$

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

Independence

Events A and B are independent if the assumption that B has occurred does not alter the probability that A occurs, or if

$$P(A | B) = P(A)$$

or

$$P(B | A) = P(B)$$

Otherwise, A&B are dependent

Example 2

$A = \{\omega_6, \omega_7, \omega_8\} \Rightarrow$ two H in one tossing

$B = \{\omega_1, \omega_6, \omega_7, \omega_8\} \Rightarrow$ > than 1 H in one tossing

$$P(A) = \frac{3}{8} \quad P(B) = \frac{4}{8} \quad P(\Omega) = \frac{8}{8} = 1$$

$$A \cap B = \{\omega_6, \omega_7, \omega_8\} = A$$

$$P(A \cap B) = P(A) = \frac{3}{8}$$

$$P(A \cap B) = P(A | B)P(B) = \frac{3}{4} \cdot \frac{4}{8} = \frac{3}{8}$$

$$P(A) \cdot P(B) = \frac{3}{8} \cdot \frac{1}{2} = \frac{3}{16} \neq P(A \cap B) = \frac{3}{8}$$

Hence, A and B are dependent

$$P(A | B) = \frac{P(A \cap B)}{P(B)} = \frac{3/8}{1/2} = \frac{3}{4} \neq P(A) = \frac{3}{8}!!$$

For independent events

$$P(A \cap B) = P(A) \cdot P(B)$$

Random variable $X(\omega)$ (discrete)

$X(\omega)$ is the number of H for example 2. From the symmetry of a coin we may suppose that the probability p to obtain a head in one trial is $p = 1/2$. ($q = 1/2 = 1 - p$). Then

$$P(X = k) = \binom{3}{k} p^k q^{3-k}, \quad k = 0, 1, 2, 3.$$

$$P(A) = P(X = 2) = \binom{3}{2} \left(\frac{1}{2}\right)^2 \left(\frac{1}{2}\right)^{3-2} = 3 \cdot \frac{1}{4} \cdot \frac{1}{2} = \frac{3}{8}$$

$$\begin{aligned} P(B) &= P(X > 1) = P(X = 2) + P(X = 3) = \\ &= \frac{3}{8} + \binom{3}{3} \left(\frac{1}{2}\right)^3 \left(\frac{1}{2}\right)^0 = \frac{3}{8} + \frac{1}{8} = \frac{1}{2} \end{aligned}$$

Events $(X=2)$ & $(X=3)$ are mutually exclusive, hence

$$\mathbf{P ((X=2) \cap (X=3)) = P (X=2)+P (X=3).}$$

We may consider $P(X=k)$ as both known and known within the unknown parameter p .

Population moments (discrete case)

Let $P(X = k)$ be the p. d. of X , $k \in \{\text{set of discrete numbers}\} = K$.

$$\sum_{k \in K} P(X = k) = 1$$

Definition 9

The initial population moment of $P(X = k)$ of order m is

$$\alpha_m = \sum_{k \in K} k^m P(X = k) = E(X^m)$$

EX^m - expectation of X^m or
expected value of X^m

Definition 10

The central population moment of $P(X = k)$ of order m is

$$\beta_m = \sum_{k \in K} (k - EX)^m P(X = k) = E(X - EX)^m$$

Definition 11 (p.210)

The mean or expected value of a discrete r. v. X is

$$\mu = EX = \sum_{k \in K} kP(X = k)$$

Definition 12

The variance of a discrete random variable X is

$$\sigma^2 = E(X - \mu)^2 = \sum_{k \in K} (k - \mu)^2 P(X = k)$$

Example Binomial distribution

$$P = (X = k) = \binom{n}{k} p^k (1-p)^{n-k}, \quad k = \overline{0, n} \Rightarrow \\ \Rightarrow K = \{0, 1, \dots, n\}$$

$$\begin{aligned} \mu = EX &= \sum_{k=0}^n k \binom{n}{k} p^k (1-p)^{n-k} = \\ &= \sum_{k=1}^n \frac{kn!}{k!(n-k)!} p^k (1-p)^{n-k} = \\ &= \sum_{k=1}^n \frac{n(n-1)!}{(k-1)!(n-k)!} p \bullet p^{k-1} q^{n-k} = \\ &= np \sum_{k=1}^n \binom{n-1}{k-1} p^{k-1} q^{n-k} = \\ k-1 \rightarrow r &\Rightarrow k = r+1 \end{aligned}$$

$$= np \sum_{r=0}^{n-1} \binom{n-1}{r} p^r q^{(n-1)-r} = np,$$

$$\sum_{r=0}^{n-1} \binom{n-1}{r} p^r q^{(n-1)-r} = (p + q)^{n-1} = 1$$

Thus,

$$\mu = EX = np$$

By the same lines

$$\sigma^2 = E(X - \mu)^2 = npq$$

Standard deviation

$$\sigma = \sqrt{npq} \quad (1a)$$