

2. Independence and Bernoulli Trials

Independence: Events A and B are independent if

$$P(AB) = P(A)P(B). \quad (2-1)$$

- It is easy to show that A, B independent implies \bar{A}, B ; A, \bar{B} ; \bar{A}, \bar{B} are all independent pairs. For example, $B = (A \cup \bar{A})B = AB \cup \bar{A}B$ and $AB \cap \bar{A}B = \phi$, so that $P(B) = P(AB \cup \bar{A}B) = P(AB) + P(\bar{A}B) = P(A)P(B) + P(\bar{A}B)$ or

$$P(\bar{A}B) = P(B) - P(A)P(B) = (1 - P(A))P(B) = P(\bar{A})P(B),$$

i.e., \bar{A} and B are independent events.

As an application, let A_p and A_q represent the events

$A_p =$ "the prime p divides the number N "

and

$A_q =$ "the prime q divides the number N ".

Then from (1-4)

$$P\{A_p\} = \frac{1}{p}, \quad P\{A_q\} = \frac{1}{q}$$

Also

$$P\{A_p \cap A_q\} = P\{\text{"} pq \text{ divides } N \text{"}\} = \frac{1}{pq} = P\{A_p\} P\{A_q\} \quad (2-2)$$

Hence it follows that A_p and A_q are independent events!

- If $P(A) = 0$, then since the event $AB \subset A$ always, we have

$$P(AB) \leq P(A) = 0 \Rightarrow P(AB) = 0,$$

and (2-1) is always satisfied. Thus the event of zero probability is independent of every other event!

- Independent events obviously cannot be mutually exclusive, since $P(A) > 0$, $P(B) > 0$ and A, B independent implies $P(AB) > 0$. Thus if A and B are independent, the event AB cannot be the null set.
- More generally, a family of events $\{A_i\}$ are said to be independent, if for every finite sub collection

$A_{i_1}, A_{i_2}, \dots, A_{i_n}$, we have

$$P\left(\bigcap_{k=1}^n A_{i_k}\right) = \prod_{k=1}^n P(A_{i_k}). \quad (2-3)$$

- Let

$$A = A_1 \cup A_2 \cup A_3 \cup \cdots \cup A_n, \quad (2-4)$$

a union of n independent events. Then by De-Morgan's law

$$\bar{A} = \bar{A}_1 \bar{A}_2 \cdots \bar{A}_n$$

and using their independence

$$P(\bar{A}) = P(\bar{A}_1 \bar{A}_2 \cdots \bar{A}_n) = \prod_{i=1}^n P(\bar{A}_i) = \prod_{i=1}^n (1 - P(A_i)). \quad (2-5)$$

Thus for any A as in (2-4)

$$P(A) = 1 - P(\bar{A}) = 1 - \prod_{i=1}^n (1 - P(A_i)), \quad (2-6)$$

a useful result.

Example 2.1: Three switches connected in parallel operate independently. Each switch remains closed with probability p . (a) Find the probability of receiving an input signal at the output. (b) Find the probability that switch S_1 is open given that an input signal is received at the output.

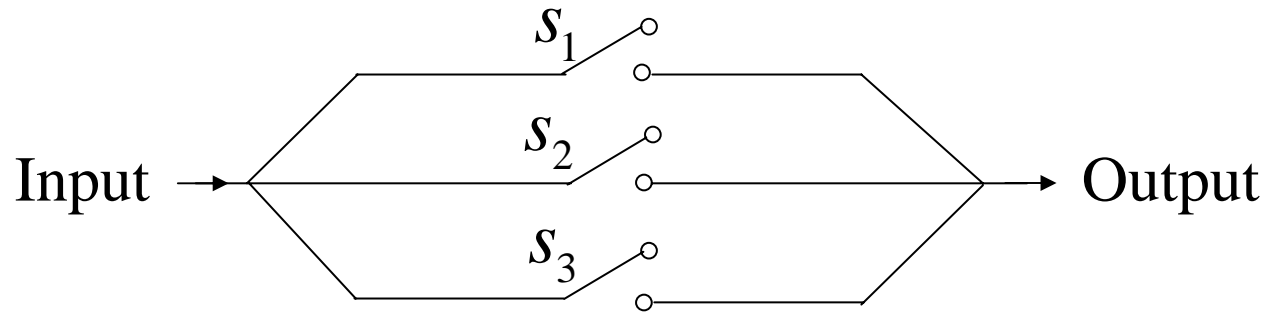


Fig.2.1

Solution: a. Let $A_i =$ “Switch S_i is closed”. Then $P(A_i) = p$, $i = 1 \rightarrow 3$. Since switches operate independently, we have

$$P(A_i A_j) = P(A_i)P(A_j); \quad P(A_1 A_2 A_3) = P(A_1)P(A_2)P(A_3).$$

Let $R =$ “input signal is received at the output”. For the event R to occur either switch 1 or switch 2 or switch 3 must remain closed, i.e.,

$$R = A_1 \cup A_2 \cup A_3. \quad (2-7)$$

Using (2-3) - (2-6),

$$P(R) = P(A_1 \cup A_2 \cup A_3) = 1 - (1 - p)^3 = 3p - 3p^2 + p^3. \quad (2-8)$$

We can also derive (2-8) in a different manner. Since any event and its complement form a trivial partition, we can always write

$$P(R) = P(R | A_1)P(A_1) + P(R | \bar{A}_1)P(\bar{A}_1). \quad (2-9)$$

But $P(R | A_1) = 1$, and $P(R | \bar{A}_1) = P(A_2 \cup A_3) = 2p - p^2$

and using these in (2-9) we obtain

$$P(R) = p + (2p - p^2)(1 - p) = 3p - 3p^2 + p^3, \quad (2-10)$$

which agrees with (2-8).

Note that the events A_1, A_2, A_3 do not form a partition, since they are not mutually exclusive. Obviously any two or all three switches can be closed (or open) simultaneously.

Moreover, $P(A_1) + P(A_2) + P(A_3) \neq 1$.

b. We need $P(\bar{A}_1 | R)$. From Bayes' theorem

$$P(\bar{A}_1 | R) = \frac{P(R | \bar{A}_1)P(\bar{A}_1)}{P(R)} = \frac{(2p - p^2)(1 - p)}{3p - 3p^2 + p^3} = \frac{2 - 2p + p^2}{3p - 3p^2 + p^3}. \quad (2-11)$$

Because of the symmetry of the switches, we also have

$$P(\bar{A}_1 | R) = P(\bar{A}_2 | R) = P(\bar{A}_3 | R).$$

Repeated Trials

Consider two independent experiments with associated probability models (Ω_1, F_1, P_1) and (Ω_2, F_2, P_2) . Let $\xi \in \Omega_1$, $\eta \in \Omega_2$ represent elementary events. A joint performance of the two experiments produces an elementary events $\omega = (\xi, \eta)$. How to characterize an appropriate probability to this “combined event” ?

Towards this, consider the Cartesian product space $\Omega = \Omega_1 \times \Omega_2$ generated from Ω_1 and Ω_2 such that if $\xi \in \Omega_1$ and $\eta \in \Omega_2$, then every ω in Ω is an ordered pair of the form $\omega = (\xi, \eta)$. To arrive at a probability model we need to define the combined trio (Ω, F, P) .

Suppose $A \in F_1$ and $B \in F_2$. Then $A \times B$ is the set of all pairs (ξ, η) , where $\xi \in A$ and $\eta \in B$. Any such subset of Ω appears to be a legitimate event for the combined experiment. Let F denote the field composed of all such subsets $A \times B$ together with their unions and compliments. In this combined experiment, the probabilities of the events $A \times \Omega_2$ and $\Omega_1 \times B$ are such that

$$P(A \times \Omega_2) = P_1(A), \quad P(\Omega_1 \times B) = P_2(B). \quad (2-12)$$

Moreover, the events $A \times \Omega_2$ and $\Omega_1 \times B$ are independent for any $A \in F_1$ and $B \in F_2$. Since

$$(A \times \Omega_2) \cap (\Omega_1 \times B) = A \times B, \quad (2-13)$$

we conclude using (2-12) that

$$P(A \times B) = P(A \times \Omega_2) \cdot P(\Omega_1 \times B) = P_1(A)P_2(B) \quad (2-14)$$

for all $A \in F_1$ and $B \in F_2$. The assignment in (2-14) extends to a unique probability measure $P(\equiv P_1 \times P_2)$ on the sets in F and defines the combined trio (Ω, F, P) .

Generalization: Given n experiments $\Omega_1, \Omega_2, \dots, \Omega_n$, and their associated F_i and P_i , $i = 1 \rightarrow n$, let

$$\Omega = \Omega_1 \times \Omega_2 \times \dots \times \Omega_n \quad (2-15)$$

represent their Cartesian product whose elementary events are the ordered n -tuples $\xi_1, \xi_2, \dots, \xi_n$, where $\xi_i \in \Omega_i$. Events in this combined space are of the form

$$A_1 \times A_2 \times \dots \times A_n \quad (2-16)$$

where $A_i \in F_i$, and their unions and intersections.

If all these n experiments are independent, and $P_i(A_i)$ is the probability of the event A_i in F_i then as before

$$P(A_1 \times A_2 \times \cdots \times A_n) = P_1(A_1)P_2(A_2) \cdots P_n(A_n). \quad (2-17)$$

Example 2.2: An event A has probability p of occurring in a single trial. Find the probability that A occurs exactly k times, $k \leq n$ in n trials.

Solution: Let (Ω, F, P) be the probability model for a single trial. The outcome of n experiments is an n -tuple

$$\omega = \{ \xi_1, \xi_2, \cdots, \xi_n \} \in \Omega_0, \quad (2-18)$$

where every $\xi_i \in \Omega$ and $\Omega_0 = \Omega \times \Omega \times \cdots \times \Omega$ as in (2-15). The event A occurs at trial # i , if $\xi_i \in A$. Suppose A occurs exactly k times in ω .

Then k of the ξ_i belong to A , say $\xi_{i_1}, \xi_{i_2}, \dots, \xi_{i_k}$, and the remaining $n - k$ are contained in its complement in \bar{A} .

Using (2-17), the probability of occurrence of such an ω is given by

$$\begin{aligned}
 P_0(\omega) &= P(\{\xi_{i_1}, \xi_{i_2}, \dots, \xi_{i_k}, \dots, \xi_{i_n}\}) = P(\{\xi_{i_1}\})P(\{\xi_{i_2}\}) \cdots P(\{\xi_{i_k}\}) \cdots P(\{\xi_{i_n}\}) \\
 &= \underbrace{P(A)P(A) \cdots P(A)}_k \underbrace{P(\bar{A})P(\bar{A}) \cdots P(\bar{A})}_{n-k} = p^k q^{n-k}. \quad (2-19)
 \end{aligned}$$

However the k occurrences of A can occur in any particular location inside ω . Let $\omega_1, \omega_2, \dots, \omega_N$ represent all such events in which A occurs exactly k times. Then

$$\text{"A occurs exactly } k \text{ times in } n \text{ trials"} = \omega_1 \cup \omega_2 \cup \dots \cup \omega_N. \quad (2-20)$$

But, all these ω_i s are mutually exclusive, and equiprobable.

Thus

$$\begin{aligned} P(\text{" } A \text{ occurs exactly } k \text{ times in } n \text{ trials" }) \\ = \sum_{i=1}^N P_0(\omega_i) = NP_0(\omega) = Np^k q^{n-k}, \end{aligned} \quad (2-21)$$

where we have used (2-19). Recall that, starting with n possible choices, the first object can be chosen n different ways, and for every such choice the second one in $(n-1)$ ways, ... and the k th one $(n-k+1)$ ways, and this gives the total choices for k objects out of n to be $n(n-1)\cdots(n-k+1)$. But, this includes the $k!$ choices among the k objects that are indistinguishable for identical objects. As a result

$$N = \frac{n(n-1)\cdots(n-k+1)}{k!} = \frac{n!}{(n-k)!k!} \triangleq \binom{n}{k} \quad (2-22)$$

represents the number of combinations, or choices of n identical objects taken k at a time. Using (2-22) in (2-21), we get

$$\begin{aligned} P_n(k) &= P(\text{" } A \text{ occurs exactly } k \text{ times in } n \text{ trials"}) \\ &= \binom{n}{k} p^k q^{n-k}, \quad k = 0, 1, 2, \dots, n, \end{aligned} \quad (2-23)$$

a formula, due to Bernoulli.

Independent repeated experiments of this nature, where the outcome is either a “success” ($= A$) or a “failure” ($= \bar{A}$) are characterized as Bernoulli trials, and the probability of k successes in n trials is given by (2-23), where p represents the probability of “success” in any one trial.

Example 2.3: Toss a coin n times. Obtain the probability of getting k heads in n trials ?

Solution: We may identify “head” with “success” (A) and let $p = P(H)$. In that case (2-23) gives the desired probability.

Example 2.4: Consider rolling a fair die eight times. Find the probability that either 3 or 4 shows up five times ?

Solution: In this case we can identify

$$\text{"success"} = A = \{ \text{either 3 or 4} \} = \{f_3\} \cup \{f_4\}.$$

Thus

$$P(A) = P(f_3) + P(f_4) = \frac{1}{6} + \frac{1}{6} = \frac{1}{3},$$

and the desired probability is given by (2-23) with $n=8$, $k=5$ and $p=1/3$. Notice that this is similar to a “biased coin” problem.

Bernoulli trial: consists of repeated independent and identical experiments each of which has only two outcomes A or \bar{A} with $P(A) = p$, and $P(\bar{A}) = q$. The probability of exactly k occurrences of A in n such trials is given by (2-23).

Let

$$X_k = \text{"exactly } k \text{ occurrences in } n \text{ trials"}. \quad (2-24)$$

Since the number of occurrences of A in n trials must be an integer $k = 0, 1, 2, \dots, n$, either X_0 or X_1 or X_2 or \dots or X_n must occur in such an experiment. Thus

$$P(X_0 \cup X_1 \cup \dots \cup X_n) = 1. \quad (2-25)$$

But X_i, X_j are mutually exclusive. Thus

$$P(X_0 \cup X_1 \cup \dots \cup X_n) = \sum_{k=0}^n P(X_k) = \sum_{k=0}^n \binom{n}{k} p^k q^{n-k}. \quad (2-26)$$

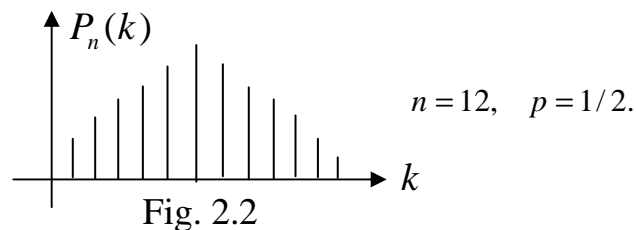
From the relation

$$(a + b)^n = \sum_{k=0}^n \binom{n}{k} a^k b^{n-k}, \quad (2-27)$$

(2-26) equals $(p + q)^n = 1$, and it agrees with (2-25).

For a given n and p what is the most likely value of k ?

From Fig.2.2, the most probable value of k is that number which maximizes $P_n(k)$ in (2-23). To obtain this value, consider the ratio



$$\frac{P_n(k-1)}{P_n(k)} = \frac{n! p^{k-1} q^{n-k+1}}{(n-k+1)!(k-1)!} \frac{(n-k)!k!}{n! p^k q^{n-k}} = \frac{k}{n-k+1} \frac{q}{p}. \quad (2-28)$$

Thus $P_n(k) \geq P_n(k-1)$, if $k(1-p) \leq (n-k+1)p$ or $k \leq (n+1)p$.
 Thus $P_n(k)$ as a function of k increases until

$$k = (n+1)p \quad (2-29)$$

if it is an integer, or the largest integer k_{\max} less than $(n+1)p$, and (2-29) represents the most likely number of successes (or heads) in n trials.

Example 2.5: In a Bernoulli experiment with n trials, find the probability that the number of occurrences of A is between k_1 and k_2 .

Solution: With $X_i, i = 0,1,2,\dots,n$, as defined in (2-24), clearly they are mutually exclusive events. Thus

$$\begin{aligned} &P(\text{"Occurrences of } A \text{ is between } k_1 \text{ and } k_2\text{"}) \\ &= P(X_{k_1} \cup X_{k_1+1} \cup \dots \cup X_{k_2}) = \sum_{k=k_1}^{k_2} P(X_k) = \sum_{k=k_1}^{k_2} \binom{n}{k} p^k q^{n-k}. \quad (2-30) \end{aligned}$$

Example 2.6: Suppose 5,000 components are ordered. The probability that a part is defective equals 0.1. What is the probability that the total number of defective parts does not exceed 400 ?

Solution: Let

$Y_k =$ " k parts are defective among 5,000 components ".

Using (2-30), the desired probability is given by

$$\begin{aligned} P(Y_0 \cup Y_1 \cup \dots \cup Y_{400}) &= \sum_{k=0}^{400} P(Y_k) \\ &= \sum_{k=0}^{400} \binom{5000}{k} (0.1)^k (0.9)^{5000-k}. \end{aligned} \tag{2-31}$$