For n = 0, (9.1) simplifies since the only possible event is "no particles in t, no particles in  $\Delta t$ ," and (9.4) becomes, for n = 0,

$$\frac{dP_0}{dt} = -\mu P_0.$$

Then, since  $P_0(0) =$  "probability that no particle is emitted during a zero time interval" = 1, integration of (9.5) gives

$$(9.6) P_0 = e^{-\mu t}.$$

Substituting (9.6) into (9.4) with n = 1 gives a differential equation for  $P_1(t)$ ; its solution (Problem 1) is  $P_1(t) = \mu t e^{-\mu t}$ . Solving (9.4) successively (Problem 1) for  $P_2$ ,  $P_3$ ,  $\cdots$ ,  $P_n$ , we obtain

(9.7) 
$$P_n(t) = \frac{(\mu t)^n}{n!} e^{-\mu t}.$$

Putting t = 1, we get for the probability of exactly n counts per unit time

(9.8) 
$$P_{n} = \frac{\mu^{n}}{n!} e^{-\mu}.$$

The meaning of  $\mu$  is important for applications of this formula; a proof is outlined in Problem 2 that  $\mu$  is just  $\bar{n}$ , the average number of counts per unit time. The probability function (9.8) is called the *Poisson distribution*.

The Poisson distribution is useful in a great variety of problems in which the probability of some occurrence is small and constant (see Problems 3 to 9, and Parratt, Chapter 5).

Example 1. The number of particles emitted each minute by a radioactive source is recorded for a period of 10 hours; a total of 1800 counts are registered. During how many 1-minute intervals should we expect to observe no particles; exactly one; etc.?

The average number of counts per minute is  $1800/(10 \cdot 60) = 3$  counts per minute; this is the value of  $\mu$ . Then by (9.8), the probability of n counts per minute is

$$P_n = \frac{3^n}{n!} e^{-3}.$$

A graph of this probability function is shown in Figure 9.1. For n = 0, we find  $P_0 = e^{-3} = 0.05$ ; then we should expect to observe no particles in about 5% of the 600 1-minute intervals, that is, during 30 1-minute intervals. Similarly, we could compute the expected number of 1-minute intervals during which 1, 2,  $\cdots$ , particles would be observed.

In Section 8, we discussed the fact that the binomial distribution can be approximated by the normal distribution for large n and large np. If p is very small so that np is very much less than n (say, for example,  $p = 10^{-3}$ , n = 2000, np = 2), the normal