

Example : Suppose independent tests are conducted on mice to develop a vaccine. If the prob of success is $\frac{1}{3}$ in each trial, what is the prob that at least 5 trials are needed to get the first success?

$X \rightarrow$ no of trials needed to get the first success.

$$P(X = k) = \left(\frac{2}{3}\right)^{k-1} \cdot \frac{1}{3}, \quad k=1, 2, \dots$$

$$P(X \geq 5) = \sum_{k=5}^{\infty} \left(\frac{2}{3}\right)^{k-1} \cdot \frac{1}{3}$$

$$= \left(\frac{2}{3}\right)^4 \cdot \frac{1}{3} \left[1 + \frac{2}{3} + \left(\frac{2}{3}\right)^2 + \dots \right]$$

$$= \left(\frac{2}{3}\right)^4 \cdot \frac{1}{3} \cdot \frac{1}{1 - \frac{2}{3}} = \left(\frac{2}{3}\right)^4 = \frac{16}{81}$$

Memoryless Property of Geometric

$$\underline{\text{Dist}^n} : P(X > m) = \sum_{k=m+1}^{\infty} q^{k-1} p$$

$$= q^m p (1 + q + q^2 + \dots)$$

$$= q^m \cdot p \cdot \frac{1}{(1-q)} = q^m$$

So $P(\text{success has not been achieved in } m \text{ trials}) = q^m.$

$$P(\underbrace{X > m+n}_E \mid \underbrace{X > n}_F) = \frac{P(E \cap F)}{P(F)}$$

$$= \frac{P(E)}{P(F)} = \frac{P(X > m+n)}{P(X > n)}$$

$$= q^{m+n} / q^n = q^m = P(X > m)$$

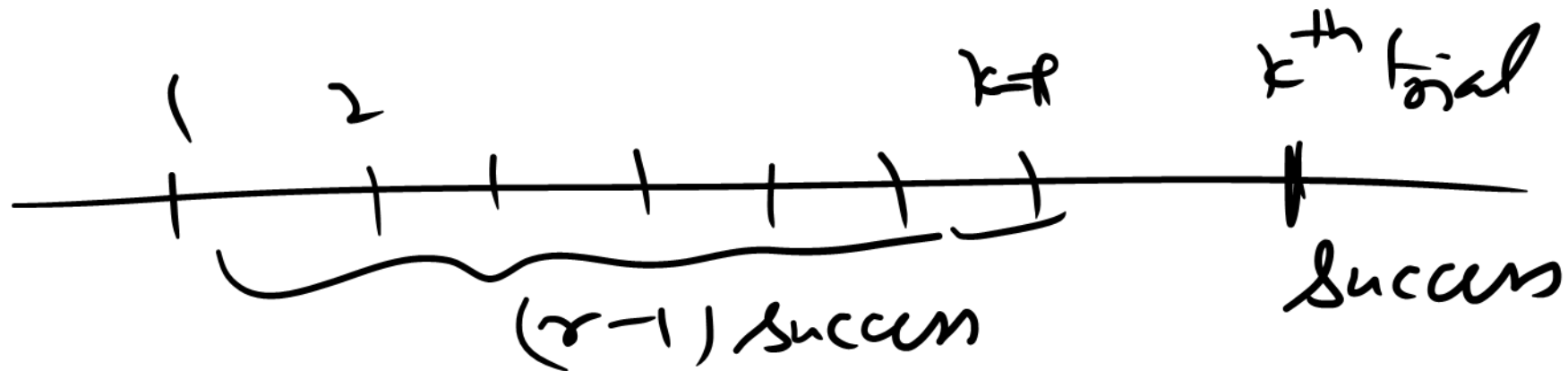
Suppose independent Bernoulli trials
are performed under identical conditions

(with prob of success p in each trial)
until r^{th} success is achieved.

$X \rightarrow$ the no. of trials required

$X \rightarrow r, r+1, r+2, \dots$

$$p_X(k) = P(X=k) = \binom{k-1}{r-1} p^{r-1} q^{k-r} \cdot p$$



$$= \binom{k-1}{r-1} q^{k-r} p^r, \quad k=r, r+1, \dots$$

Negative Binomial / Inverse Bin.
distⁿ.

$$\sum_{k=r}^{\infty} \binom{k-1}{r-1} q^{k-r} p^r = p^r \cdot \frac{1}{(1-q)^r}$$

$$E(X) = \sum_{k=r}^{\infty} k \cdot \binom{k-1}{r-1} q^{k-r} p^r = 1$$

$$= \frac{r}{p} \left[\sum_{k=r+1}^{\infty} \underbrace{\frac{k!}{r! (k-r)!}}_1 q^{k+1-\overline{r+1}} p^{r+1} \right]$$

$$= \frac{r}{p}$$

$$E(X^2) = E(\underbrace{X(X-1)}_{\text{}}) + E(X) \quad (\otimes)$$

$$V(X) = \frac{rq}{p^2}$$

$$M_X(t) = E(e^{tX})$$

$$= \sum_{k=r}^{\infty} e^{tk} \binom{k-1}{r-1} q^{k-r} p^r$$

$$= p^r e^{tr} \sum_{k=r}^{\infty} \binom{k-1}{r-1} (q e^t)^{k-r}$$

$$= (p e^t)^r / (1 - q e^t)^r$$

$$= \left(\frac{pe^t}{1 - qe^t} \right)^r, \quad qe^t < 1$$

or $t < -\log_e q$

Example: Suppose a heavy machine has several independent components each with failure probability p . The machine fails if 4 components fail. In how many epochs the machine will stop

functioning $\rightarrow X$

$X \rightarrow 4, 5, \dots$

$$P(X = k) = \binom{k-1}{3} q^{k-4} p^4, \\ k = 4, 5, \dots$$

Hypergeometric Distribution

Suppose a population has N items

Type I	Type II
M	N-M

Suppose we randomly select n items without replacement.

$X \rightarrow$ the no. of items of Type I in the sample

$X \rightarrow 0, 1, \dots, n$

$$p_x(k) = \frac{\binom{M}{k} \binom{N-M}{n-k}}{\binom{N}{n}}, \quad k=0,1,\dots,n$$

$$\text{s.t.} \quad k \leq M, \quad n-k \leq N-M$$

$$\sum_{k=0}^n \binom{M}{k} \binom{N-M}{n-k} = \binom{N}{n} \quad \textcircled{\times}$$

Consider the coefficient of x^n in

the expansion of $(1+x)^N = (1+x)^M (1+x)^{N-M}$

$$E(X) = \sum_{k=0}^n k \cdot \frac{\binom{M}{k} \binom{N-M}{n-k}}{\binom{N}{n}} = \frac{Mn}{N}$$

$$E[X(X-1)] = \frac{M(M-1)n(n-1)}{N(N-1)}$$

$$\begin{aligned} E(X^2) &= E\{X(X-1)\} + E(X) \\ &= \frac{nM(nM - n - M + N)}{N(N-1)} \end{aligned}$$

$$\begin{aligned}
 V(X) &= E(X^2) - (E(X))^2 \\
 &= \frac{nM(N-n)(N-M)}{N^2(N-1)} \quad \textcircled{5}
 \end{aligned}$$

$$= \left(\frac{N-n}{N-1} \right) \frac{nM}{N} \left(1 - \frac{M}{N} \right)$$

Theorem: Let X have hypergeometric distⁿ with parameters (N, M, n) .

As $N \rightarrow \infty$, $M \rightarrow \infty \Rightarrow \frac{M}{N} \rightarrow p$,
the $P(X=x) \rightarrow \binom{n}{x} p^x q^{n-x}$

ie hypergeometric can be
approximated by a binomial distⁿ
for large populations.

Proof (*)

Hypergeometric distⁿ has wide application

in estimating items of one category /
estimating population size etc. /

Suppose there is need for survey to
estimate amount of a costly mineral
in a mine area. Here

$$E(\bar{X}) = \frac{nM}{N} \quad \Rightarrow \quad M \approx \frac{N\bar{X}}{n}$$

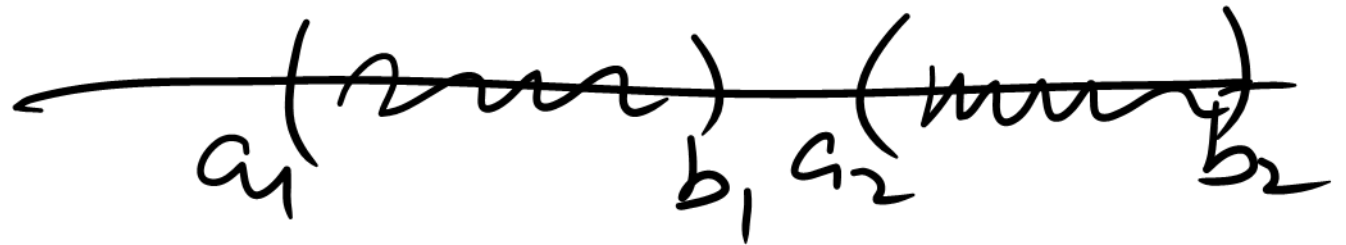
We can also estimate total population size if type I items are known

$$N \approx \frac{nM}{X}$$

Poisson Process : Events/
occurrences / happenings / observations
over time / area / space are said
to follow a Poisson process if they

satisfy the following assumptions:

1. The number of occurrences in disjoint time intervals are indept

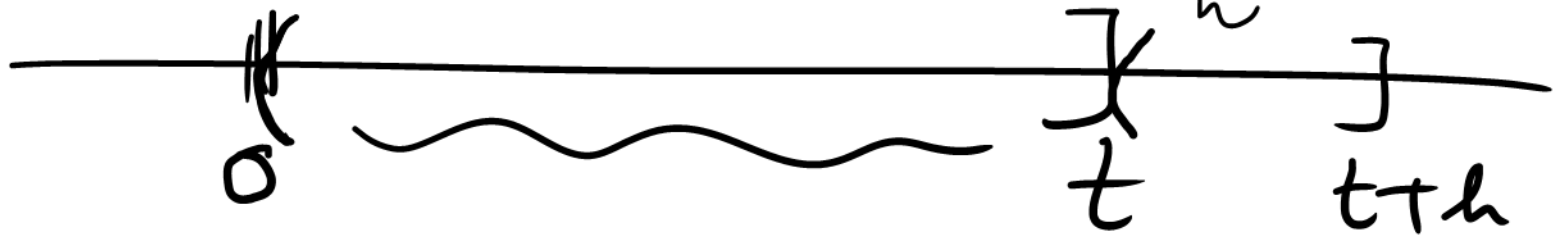


2. The probability of a single occurrence in a small time interval is proportional to the length of interval.

$$P_1(h) = \lambda h$$

3. The probability of more than one occurrence in a small time interval is negligible. $P_2(h) + P_3(h) + \dots = o(h)$
 $\Rightarrow 1 - P_0(h) - P_1(h) = o(h)$

$$\frac{o(h)}{h} \rightarrow 0$$



$X(t) \rightarrow$ number of occurrence
in $[0, t]$
(in an interval of length t)

$$\begin{aligned}
 P(X(t) = n) &= P_n(t) \\
 &= P(n \text{ occurrences in an interval} \\
 &\quad \text{of length } t)
 \end{aligned}$$

Under assumptions (1) - (3), the
distⁿ of $X(t)$ is given by

$$P_n(t) = \frac{e^{-\lambda t} (\lambda t)^n}{n!}, \quad n=0, 1, 2, \dots \quad \dots (1)$$

Proof: First we write assumptions
(2) & (3) in a mathematical form

$$P_1(h) = \lambda h + \underbrace{O(h)} \quad \dots (2)$$

$$1 - P_0(h) - P_1(h) = O(h)$$

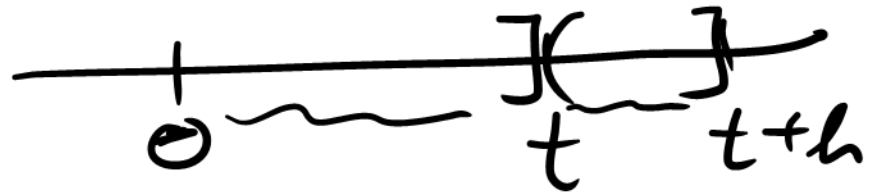
$$\Rightarrow P_0(h) = 1 - \lambda h + O(h) \quad \dots (3)$$

In order to prove (1), we use

the principle of induction.

$n=0$

Consider



$$P_0(t+h) = P(\text{no occurrence in } (0, t+h])$$

$$= P(\underbrace{\text{no occurrence in } (0, t]} \cap \underbrace{\text{no occurrence in } (t, t+h]})$$

$$= P(\text{no occurrence in } (0, t]) \cdot P(\text{no occurrence in } (t, t+h])$$

as the no. of occurrences in disjoint intervals are independent

$$= P_0(t) P_0(h)$$

$$= P_0(t) (1 - \lambda h + o(h))$$

$$\Rightarrow P_0(t+h) - P_0(t) = -\lambda h P_0(t) + o(h) P_0(t)$$

Divide by h on both sides and take

$$h \rightarrow 0,$$

$$P_0'(t) = -\lambda P_0(t)$$

This is first order ODE (variable separable form) with solution

$$P_0(t) = c e^{-\lambda t}$$

We can use initial condition

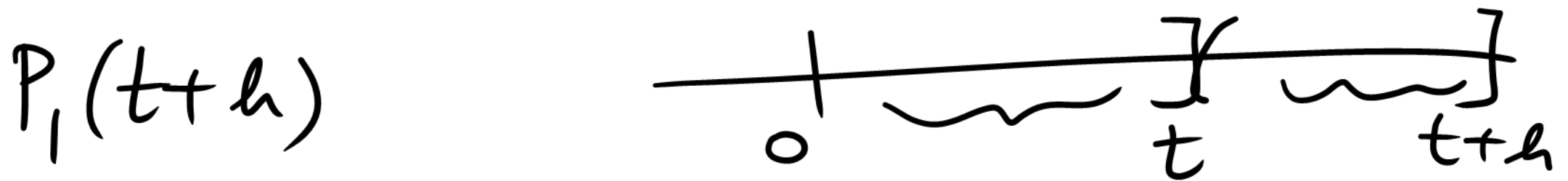
$$P_0(0) = 1, \text{ then we get } c = 1$$

So the solution is $P_0(t) = e^{-\lambda t}$

So the statement (1) is proved for

$$n=0.$$

Now we take $n=1$. Consider



$$= P(\text{one occurrence in } (0, t+h])$$

$$= P(\{\text{one occurrence in } \underline{(0, t)}\} \cap$$

$\{ \text{no occurrence in } \underline{(t, t+h]} \}$

$+ P(\{ \text{no occurrence in } (0, t] \} \cap$
 $\{ \text{one occurrence in } (t, t+h] \})$

$= \frac{P(\text{one occurrence in } (0, t])}{P(\text{no occurrence in } (t, t+h])}$

$+ P(\text{no occurrence in } (0, t])$

$\frac{P(\text{one occurrence in } (t, t+h])}{P(\text{no occurrence in } (t, t+h])}$

$$= P_1(t) P_0(h) + P_0(t) P_1(h)$$

$$= P_1(t) (1 - \lambda h + o(h))$$

$$+ e^{-\lambda t} (\lambda h + o(h))$$

$$\Rightarrow \frac{P_1(t+h) - P_1(t)}{h} = \frac{-\lambda P_1(t)}{h} + \lambda e^{-\lambda t} + \frac{o(h)}{h} (P_1(t) + e^{-\lambda t})$$

Taking limit as $t \rightarrow 0$, we get

$$P_1'(t) = -\lambda P_1(t) + \lambda e^{-\lambda t}.$$

This is again a first order linear ODE and has solution

$$P_1(t) = \lambda t e^{-\lambda t} + C_1,$$

Using initial condition $P_1(0) = 0$,

we get $C_1 = 0$. So

$$P_1(t) = \lambda t e^{-\lambda t}$$

So the statement (1) holds for

$$n = 1.$$

Next we assume that (1) holds for $n \leq k$. Now consider

$$n = k + 1. \quad \begin{array}{c} | \qquad \qquad \qquad | \qquad | \\ \hline 0 \qquad \qquad \qquad t \qquad t+h \end{array}$$

$$P_{k+1}(t+h) = P((k+1) \text{ occurrence in } (0, t+h])$$

$$= P(\{(k+1) \text{ occurrence in } (0, t]\} \cap \{ \text{no occurrence in } (t, t+h]\})$$

$$+ P(\{k \text{ occurs in } (0, t]\} \cap \{ \text{one occur in } (t, t+h]\})$$

$$+ \sum_{j=1}^k P(\{(k-j) \text{ occur in } (0, t]\} \cap \{(j+1) \text{ occur in } (t, t+h]\})$$

$$\begin{aligned}
&= P_{k+1}(t) P_0(h) + P_k(t) P_1(h) \\
&\quad + \sum_{j=1}^k P_{k-j}(t) P_{j+1}(h)
\end{aligned}$$

$$\begin{aligned}
&= P_{k+1}(t) (1 - \lambda h + o(h)) \\
&\quad + \frac{e^{-\lambda t} (\lambda t)^k}{k!} (\lambda h + o(h))
\end{aligned}$$

$$+ o(h) \left(\sum_{j=1}^k P_{kj}(t) \right)$$

$$\Rightarrow \frac{P_{k+1}(t+h) - P_{k+1}(t)}{h}$$

$$= -\lambda P_{k+1}(t) + \lambda \frac{e^{-\lambda t} (\lambda t)^k}{k!}$$

$$+ \frac{O(h)}{h} \left[\sum_{j=1}^{k+1} P_j(t) \right]$$

Taking limit as $h \rightarrow 0$, we get

$$P'_{k+1}(t) = -\lambda P_{k+1}(t) + \frac{\lambda t^k}{k!} e^{-\lambda t}$$

This is a linear first order ODE.

The general solution is

$$P_{k+1}(t) = \frac{e^{-\lambda t} (\lambda t)^{k+1}}{(k+1)!} + C_2$$

Taking initial condition $P_{k+1}(0) = 0$,
we get $C_2 = 0$. So the solⁿ is

$$P_{k+1}(t) = \frac{e^{-\lambda t} (\lambda t)^{k+1}}{(k+1)!}$$

So statement (1) holds for $n=k+1$.

So by the Principle of Mathematical
Induction $P_n(t)$ holds for all
 $n=0, 1, 2, \dots$.