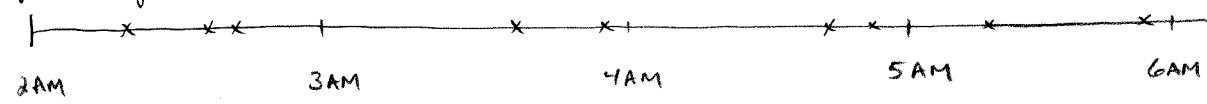


Poisson process

Situation

You answer telephones for a mail-order company.
 You work the 2 AM - 6 AM shift; there are very few calls, and no time tends to be busier than other times. 3 per hour, on average.
 You pay attention to the times at which the phone rings, and you have lots of time to think!

A typical night:



Modeling

Think about the number of calls received over each hour:

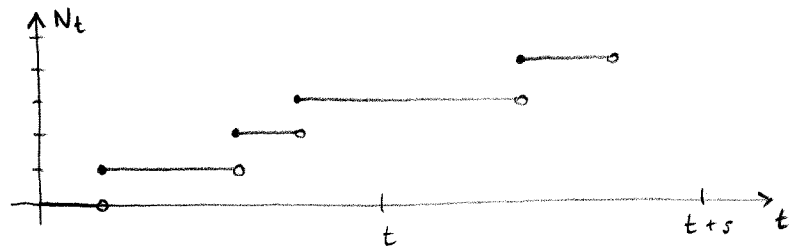


Are these random variables independent?
 identically distributed?

Both seem reasonable.

Number of calls over each half hour, B₁, ... B₇? Different distribution, but still iid.

Call 2 AM time 0, imagine your shift going on forever.
 Let N_t be the number of calls that arrive in (0, t].

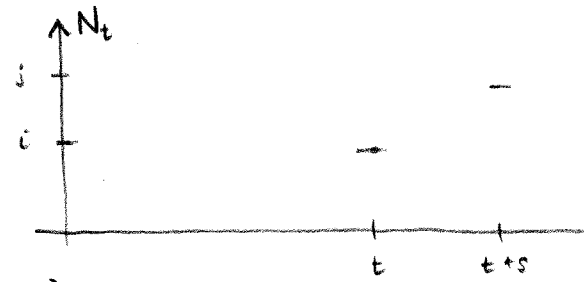


Then N_{t+s} - N_t is the number of calls that arrive in (t, t+s].

N is a stochastic process with index set [0, ∞) and state space {0, 1, 2, ...}.

Is N a Markov process?

$$\begin{aligned}
& \mathbb{P}(N_{t+s} = j \mid N_t = i, N_r, r \leq t) \\
&= \mathbb{P}(N_{t+s} - N_t = j - i \mid N_t = i, N_r, r \leq t) \\
&= \mathbb{P}(j-i \text{ calls arrive in } (t, t+s] \mid N_t = i, N_r, r \leq t) \\
&= \mathbb{P}(j-i \text{ calls arrive in } (t, t+s]) \quad * \text{ reasonable modeling assumption} \\
&= \mathbb{P}(j-i \text{ calls arrive in } (t, t+s] \mid N_t = i) \\
&= \mathbb{P}(N_{t+s} = j \mid N_t = i), \quad \text{so yes, it is a Markov process} \\
&= \mathbb{P}(j-i \text{ calls arrive in } (0, s]) \quad * \text{ reasonable modeling assumption} \\
&= \mathbb{P}(N_s = j \mid N_0 = i) \quad \text{time homogeneous}
\end{aligned}$$



So far, so good.

By the Decomposition Theorem, the successive states visited by N form a Markov chain Y_0, Y_1, Y_2, \dots

The value of Y always increases.

Can it increase by more than one?

Modeling assumption: no, cannot have two calls arrive at exactly the same time.
 Only if people conspired to call at the same time.

Then Q is very easy!

$$Q(i, i+1) = 1, \text{ all others } 0.$$

$$Q = \begin{matrix} & \begin{matrix} 0 & 1 & 2 & 3 & \dots \end{matrix} \\ \begin{matrix} 0 \\ 1 \\ 2 \\ 3 \\ \vdots \end{matrix} & \begin{bmatrix} 0 & 1 & 0 & 0 & \dots \\ 0 & 0 & 1 & 0 & \dots \\ 0 & 0 & 0 & 1 & \dots \\ \vdots & \vdots & \vdots & \vdots & \ddots \end{bmatrix} \end{matrix}$$

From the theory we've done, we know that the length of time spent in state i satisfies

$$P(W_t > u \mid N_t = i) = e^{-\lambda(i)u}, \quad u \geq 0,$$

where the parameter λ may depend on i .

But look closer:

$$\begin{aligned} P(W_t > u \mid N_t = i) &= P(\text{no arrivals in } [t, t+u] \mid N_t = i) \\ &= P(\text{no arrivals in } [t, t+u]), \end{aligned}$$

irrespective of i , the number of arrivals up until time t .

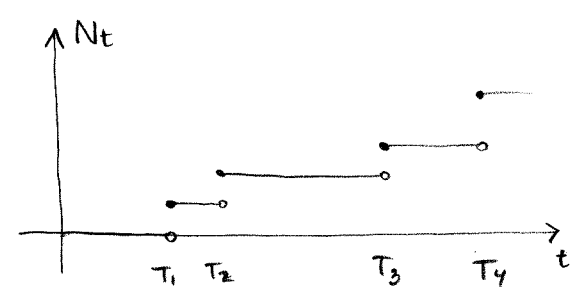
Thus, $\lambda(i)$ is the same for all i ; we'll just write λ .

Poisson process

- assumptions
- $N_t, t \geq 0$ takes values in $\{0, 1, 2, \dots\}$
 - N increases by jumps of size one only.
 - $N_{t+s} - N_t$ is independent of $N_r, r \leq t$.
 - The distribution of $N_{t+s} - N_t$ does not depend on t

Then there is a number $\lambda \geq 0$, called the arrival rate for which the length of time N spends in each state is Exponential with mean $\frac{1}{\lambda}$.

Successive times between jumps are independent and identically distributed.



Example $\lambda = 3$ calls per hour.

Suppose you arrive $\frac{1}{2}$ hour late for your shift. What is the probability that you have not missed a call?

$$P(T_1 > \frac{1}{2}) = e^{-\lambda \cdot \frac{1}{2}} = e^{-3 \cdot \frac{1}{2}} = 0.2231.$$

Pretty low. Better not be that late!

$$5 \text{ minutes late? } P(T_1 > \frac{1}{12}) = e^{-3 \cdot \frac{1}{12}} = 0.7788.$$

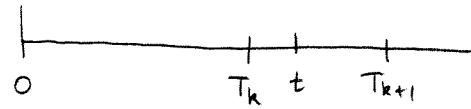
Distribution of N_t

First, relate N_t to the times of arrivals:

$$\{N_t = k\} = \{T_k \leq t \text{ and } T_{k+1} > t\},$$

because arrivals occur singly.
So

$$\begin{aligned} P(N_t = k) &= P(T_k \leq t \text{ and } T_{k+1} > t) \\ &= P(T_k \leq t \text{ and not } T_{k+1} \leq t) \\ &= P(T_k \leq t) - P(T_{k+1} \leq t) \end{aligned}$$



$$P(T_k \leq t) = P(T_{k+1} \leq t) + P(T_k \leq t \text{ and not } T_{k+1} \leq t)$$

What is $P(T_k \leq t)$?

This is the cdf of T_k , which is the sum of k independent Exponential random variables.

These are $\text{Gamma}(1, \frac{1}{\lambda})$ variables, and the sum of k iid ones is a $\text{Gamma}(k, \frac{1}{\lambda})$ random variable,

so the density of T_k is $f_k(t) = \frac{\lambda (\lambda t)^{k-1}}{(k-1)!} e^{-\lambda t}$, $t \geq 0$.

The cdf of T_k is $F_k(t) = 1 - e^{-\lambda t} \sum_{n=0}^{k-1} \frac{(\lambda t)^n}{n!}$,

which you can check by differentiating ($F_k' = f_k$).

Finally,

$$\begin{aligned} P(N_t = k) &= F_k(t) - F_{k+1}(t) \\ &= \left(1 - e^{-\lambda t} \sum_{n=0}^{k-1} \frac{(\lambda t)^n}{n!}\right) - \left(1 - e^{-\lambda t} \sum_{n=0}^k \frac{(\lambda t)^n}{n!}\right) \\ &= e^{-\lambda t} \frac{(\lambda t)^k}{k!}, \end{aligned}$$

so that $N_t \sim \text{Poisson}(\lambda t)$.

Example

Telephone center for mail-order catalog.

What we know:

- qualitative: calls arrive according to a Poisson process with some parameter λ (same for all states)
- quantitative: on average, there are 3 calls per hour.

Now we have done enough theory to know that $N_t \sim \text{Poisson}(\lambda t)$,

$$\text{so } \mathbb{E}N_t = \lambda t = 3t,$$

so in fact the parameter λ is equal to the average number of calls per hour:

$$\lambda = 3 \frac{\text{calls}}{\text{hour}} \quad \underline{\text{Arrival rate.}}$$

Now we can do some specific numerical calculations.

- Expected time of first phone call (after time 0):

$$\mathbb{E}T_1 = \frac{1}{\lambda} = \frac{1}{3 \frac{1}{\text{hour}}} = \frac{1}{3} \text{ hour} = 20 \text{ minutes.}$$

That makes sense.

- Expected time of 5th phone call:

$$\begin{aligned} \mathbb{E}T_5 &= \mathbb{E}T_5 - T_4 + T_4 - T_3 + \dots + T_1 \\ &= 5 \cdot \frac{1}{\lambda} \\ &= 5 \cdot \frac{1}{3} \text{ hour} \\ &= 100 \text{ minutes} \end{aligned}$$

- Average number of calls per 4-hour shift:

$$\mathbb{E}N_4 = \lambda \cdot 4 = 12.$$

- Probability of more than 15 calls in a shift

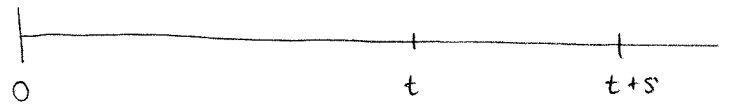
$$\begin{aligned} \mathbb{P}(N_4 > 15) &= 1 - \mathbb{P}(N_4 \leq 15) \\ &= 1 - \text{poisscdf}(15, 12) \\ &= 0.1556 \end{aligned}$$

Matlab

- $P(N_1 = 4, N_2 = 8, N_3 = 12)$
= $P(N_3 - N_2 = 4, N_2 - N_1 = 4, N_1 = 4)$
= $P(N_3 - N_2 = 4) P(N_2 - N_1 = 4) P(N_1 = 4)$
= $P(N_1 = 4)^3$
= $\left(\frac{e^{-\lambda} \lambda^4}{4!}\right)^3$
= 0.0047442874

445/515
Theoretical results

• Poisson distribution



$$N_{t+s} - N_t \sim \text{Poisson}(\lambda s)$$

N_t and $N_{t+s} - N_t$ are $\text{Poisson}(\lambda t)$ and $\text{Poisson}(\lambda s)$ and indep.

Their sum, N_{t+s} is $\text{Poisson}(\lambda(t+s))$.

One adds the parameters.

In fact, if $X \sim \text{Poisson}(\mu)$, $Y \sim \text{Poisson}(\nu)$, independent, then

$$X + Y \sim \text{Poisson}(\mu + \nu).$$

Proof ① Let $\lambda t = \mu$, $\lambda s = \nu$, use Poisson processes!

$$\textcircled{2} \mathbb{P}(X+Y=k) = \sum_{\ell=0}^k \mathbb{P}(X=\ell, X+Y=k) \dots$$

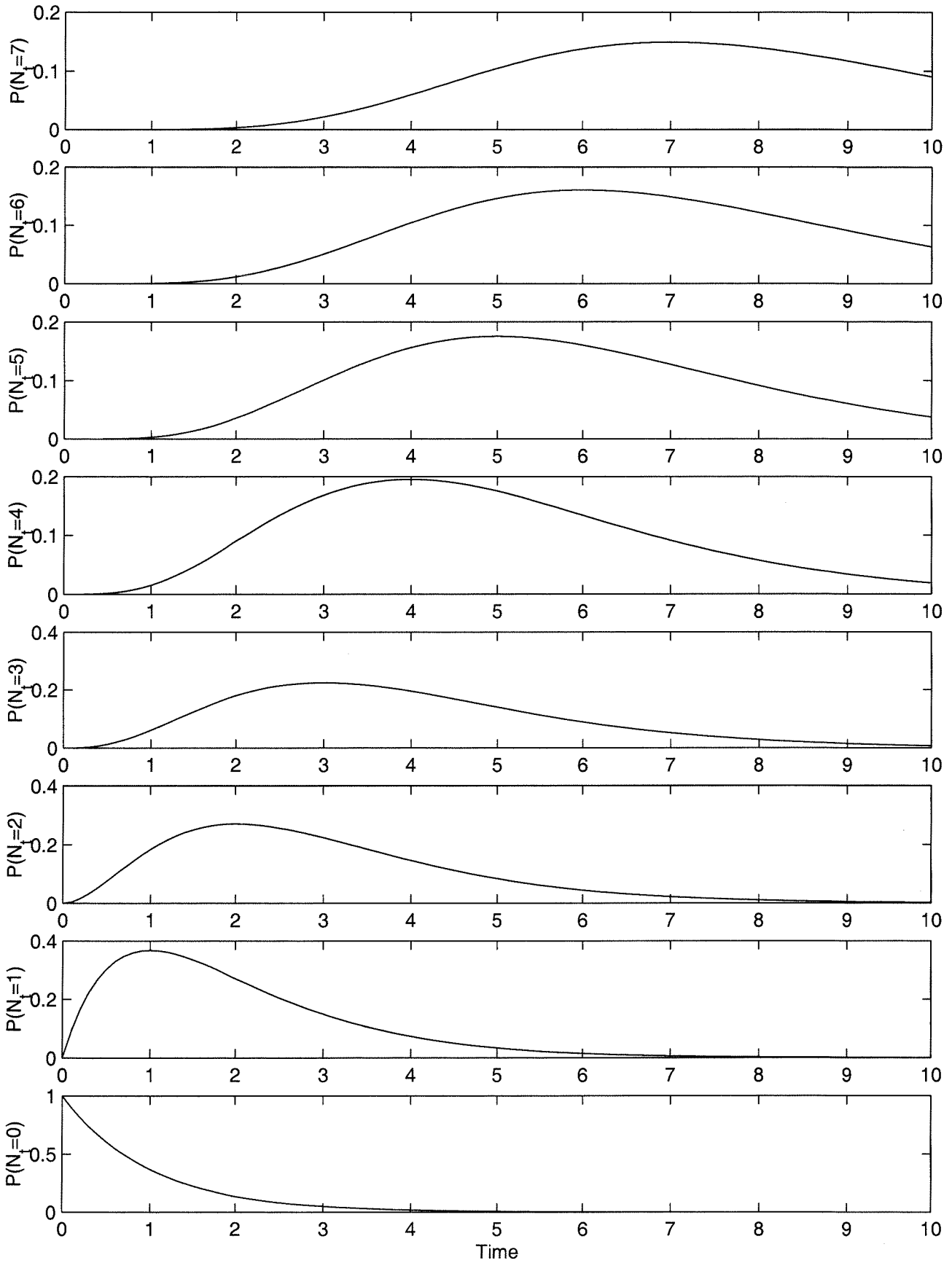
• Transition matrix

$$\begin{aligned} P_{ij}(s) &= \mathbb{P}(N_{t+s} = j \mid N_t = i) \\ &= \mathbb{P}(N_{t+s} - N_t = j - i) \\ &= \begin{cases} \frac{e^{-\lambda s} (\lambda s)^{j-i}}{(j-i)!} & j \geq i \\ 0 & \text{otherwise} \end{cases} \end{aligned}$$

$$P(s) = \begin{matrix} 0 \\ 1 \\ 2 \end{matrix} \begin{bmatrix} e^{-\lambda s} & e^{-\lambda s} \lambda s & \frac{e^{-\lambda s} (\lambda s)^2}{2} & \frac{e^{-\lambda s} (\lambda s)^3}{3!} & \dots \\ 0 & e^{-\lambda s} & e^{-\lambda s} \lambda s & & \\ 0 & 0 & e^{-\lambda s} & & \end{bmatrix}$$

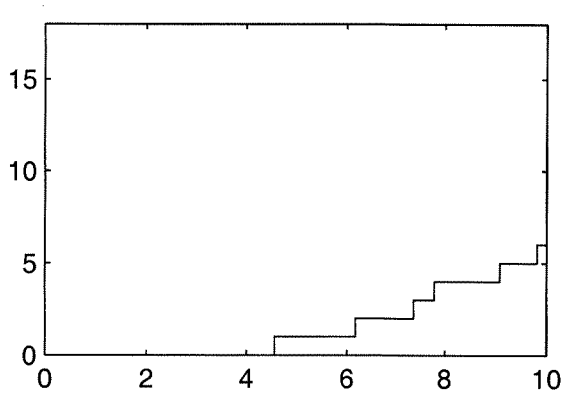
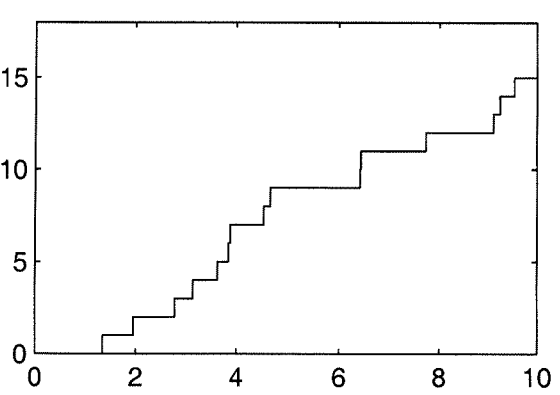
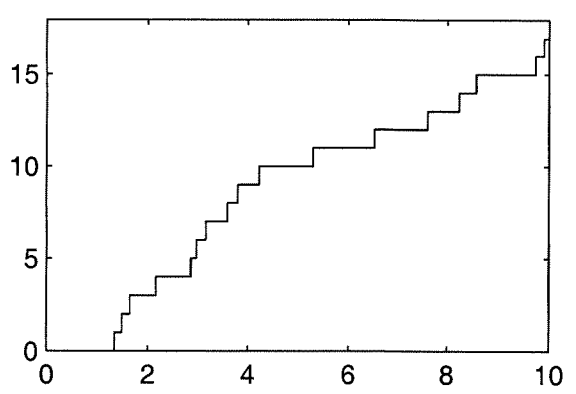
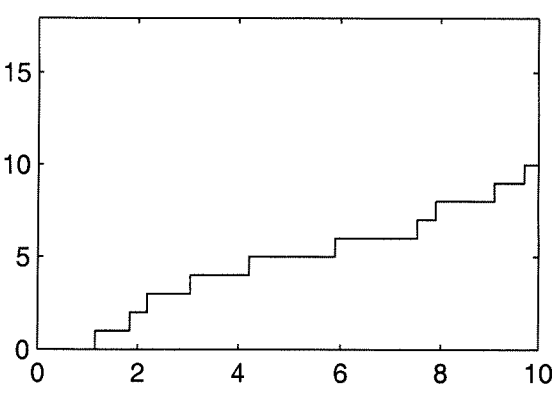
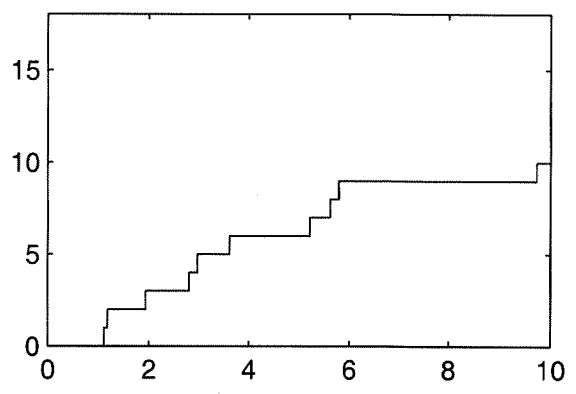
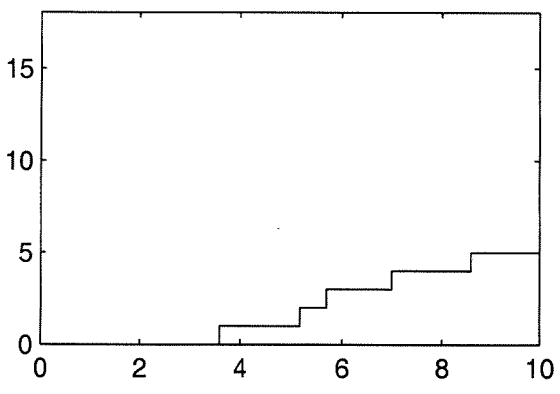
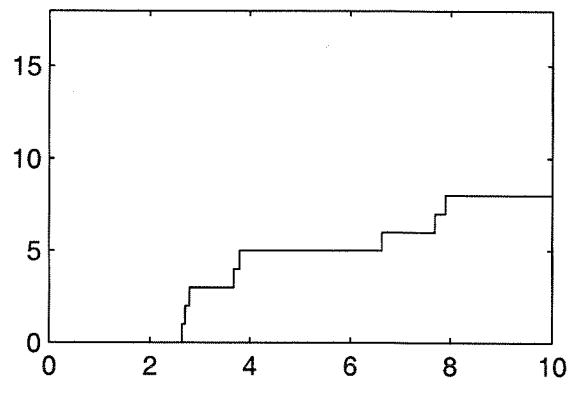
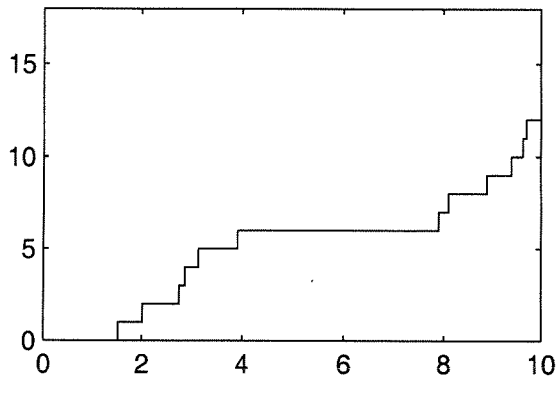
This is one of the only cases in which you can write down $P(s)$ easily/explicitly.

Distribution of N_t



poisson - probs.m

Outcomes of the Poisson process, rate 1



manypoisson.m