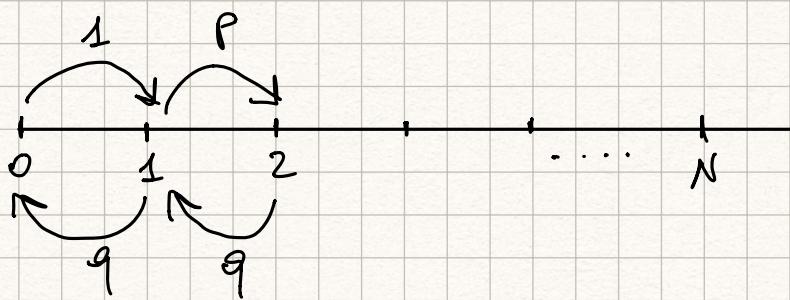


Q 3: consider a random walk over the non negative integers with the following transition probability: $P_{0i} = 1$, $P_{i,i+1} = p$, $P_{i,i-1} = q$ $i > 0$
 $p + q = 1$. Study its behaviour and in particular characterize its recurrence or transience and derive the steady state distribution.



in the trivial case in which $p = 0$ the random walk is recurrent in the state 0 and 1 with period 2.

In the non-trivial case in which $0 < p < 1$
if $p > \frac{1}{2} \rightarrow$ RW goes to $+\infty$
if $p < \frac{1}{2} \rightarrow$ RW goes to 0

Q 5: Prove that for a renewal process
 $E[S_N(t) + 1] = E[X](R(t) + 1)$

We start by conditioning the time of the first renewal on $X_1 = x$

$$E[S_N(t) + 1 | X_1 = x] = \begin{cases} x, & \text{if } x > t \\ x + A(t - x), & \text{if } x \leq t \end{cases}$$

Invoking the law of total probability
we obtain:

$$A(t) = E[S_N(t) + 1] = E[X_1] + \int_0^t A(t-x) dF(x)$$

This is a renewal equation in the
case of $E[X_1] = \omega(t)$

So we can rewrite $E[X_1] + \int_0^t A(t-x) dF(x)$

as

$$\begin{aligned} A(t) &= \omega(t) + \int_0^t \omega(t-x) dM(x) \\ &= E[X_1] \times [1 + \bar{M}(t)] \end{aligned}$$

Q9 prove that for a Markov chain the
n-step transition probabilities $P_{ij}^{(n)}$ satisfy
the relationship $P_{ij}^{(n)} = \sum P_{im}^{(k)} P_{mj}^{(n-k)} \quad k=0,1-n$

$$\begin{aligned} P_{ij}^{(n)} &= \sum_{m=0}^n P[X_n=j, X_k=m \mid X_0=i] \\ &= P_{mj}^{(n-k)} \xrightarrow{\text{forget for } k \text{ prop}} P_{im}^{(k)} \\ &= \sum_m P[X_n=j \mid \cancel{X_0=i}, X_k=m] \cdot P[X_k=m \mid X_0=i] \\ &= \sum_m P_{im}^{(k)} \cdot P_{mj}^{(n-k)} \end{aligned}$$

Q7: Prove that a MC with finite number of states cannot have a null recurrent state.

null recurrent state means that

m_i for that state is equal to $+\infty$

and m_i being $E[R_i | X_0=i]$

so the expected time to return to stat i conditioned on the fact that we started from i.

so if $m_i = \infty$ and the state is recurrent we need infinite number of recurrent states in his class to satisfy this property \rightarrow but we have finite states for hypothesis go contradiction.

Q11: prove that in a Markov chain with a finite number of states:

a) there must be a recurrent state

b) it must not have a NULL recurrent state.

b) is answered in Q7.

2) A recurrent state is defined as a state with $f_{ii} = 1$ this means that is certain that starting from him we will return to him.

A transient state on the contrary has $f_{ii} < 1$ this means that even in an ∞ number of step we are not guaranteed to return to him.

In a finite number of state if we don't have a recurrent state is trivial to guess that all transient states will have $f_{ii} = 1$ so becoming recurrent.

Thus we need at least a recurrent state to keep transient states transient.

Q1 State and prove (the elementary Renewal Theorem)

A: let $\{N(t) : t \geq 0\}$ be a Renewal (counting) process and $M(t) = E[N(t)]$
Then

$$\lim_{t \rightarrow \infty} \frac{M(t)}{t} = \lim_{t \rightarrow \infty} \frac{E[N(t)]}{t} = \frac{1}{\mu}, \text{ where } \mu \text{ is the mean interoccurrence time}$$

PROOF: (Note: does not follow directly from strong LN for renewal counting process, since almost sure convergence does not imply convergence of means)

We start by showing that $N(t)$ is finite $\forall t > 0$
let $s_n = x_1 + x_2 + \dots + x_n = \sum_{i=1}^n x_i, n \geq 1$; waiting time until the n -th event, where x_i is the elapsed time from the $(i-1)$ -th until the i -th event

$$\text{let } F_n(t) = P[S_n \leq t] = P[N(t) \geq n]$$

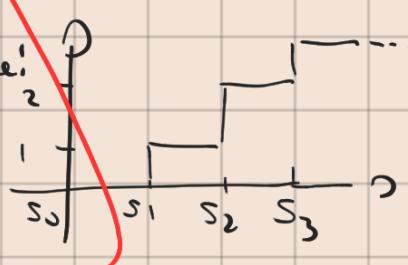
Using the convolution formula we have:

$$F_n(t) = \int_0^t F_{n-m} (t-\varepsilon) dF_m(\varepsilon)$$

To prove?

and

$$\int F_{n-m} (t-\varepsilon) dF_m(\varepsilon) \leq F_{n-m} (t) \cdot F_m(t), 1 \leq m \leq n-1$$



$$\Rightarrow \pi(t) = E[N(t)] = \sum_{n=1}^{+\infty} P[S_n \leq t] = \sum_{n=1}^{+\infty} F_n(t)$$

$$\begin{aligned} \Rightarrow \pi(t) &= \sum_{n=1}^{+\infty} F_n(t) = \sum_{m=0}^{+\infty} \sum_{k=1}^{\infty} F_{m+k}(t) \leq \sum_{m=0}^{+\infty} \sum_{k=1}^{\infty} (F_m(t))^k \\ &= \left(\sum_{m=0}^{+\infty} (F_m(t))^k \right) \cdot \underbrace{\left(\sum_{k=1}^{\infty} F_m(t) \right)}_{\text{geometrically}} \\ &\leq R \text{ since } F_m(t) \end{aligned}$$

$\Rightarrow \pi(t)$ converges if $F_m(t) < 1$

There is a theorem which states: $\forall t, \exists n$ s.t. $F_n(t) < 1$

Proof: we can always find $t_0 > 0$ s.t. $F(t) < 1, \forall t < t_0$

\Rightarrow We have:

$$\begin{aligned} 1 - F_n(t) &= P[S_n > t] \geq (P[X_i > \frac{t}{n}])^n \\ &= (1 - F(\frac{t}{n}))^n \end{aligned}$$

which is > 0 if $\frac{t}{n} < t_0$ (by the ... $F(t) < 1 \forall t < t_0$)
 $\Rightarrow t/n < t_0$

So, $\pi(t)$ converges geometrically

Show $\frac{\pi(t)}{t} \rightarrow \frac{1}{\mu}$ ($t \rightarrow +\infty$)

Q?

Prove that in a Markov Chain the period is a class property

Proof:

Period is a class property means: $i \leftrightarrow j \Rightarrow d(i) = d(j)$

By hp ($i \leftrightarrow j \Rightarrow i$ and j communicate) $\exists m, n \in \mathbb{N} > 0$
s.t.

$$P_{ij}^{(m)} > 0 \text{ and } P_{ji}^{(n)} > 0$$

$$\Rightarrow P_{jj}^{(m+n)} \geq P_{ji}^{(n)} \cdot P_{ij}^{(m)} > 0$$

Assume that $P_{ii}^{(s)} > 0$, $s \in \{m \geq 1 : P_{ii}^{(s)} > 0\}$, then

$$P_{jj}^{(m+s+n)} \geq P_{ji}^{(m)} \cdot P_{ii}^{(s)} \cdot P_{ij}^{(n)} > 0$$

$$\text{So } m+s+n \in \{m \geq 1 : P_{jj}^{(m)} > 0\},$$

Let $d(j)$ be gcd between $\{m \geq 1 : P_{jj}^{(m)} > 0\} \Rightarrow d(j) | m+n+s$

& $d(j) | m+n+s \Rightarrow d(j)$ divides $(m+n+s) - (m+n) = s$

$\Rightarrow d(j)$ divides $d(i)$

With similar computation we can prove that $d(i) | d(j)$

$$\text{So } d(i) = d(j)$$

\Rightarrow i and j have the same period

Q4 - For a Poisson process of rate λ , prove that the interarrival times are iid exp. with mean $\frac{1}{\lambda}$

A4: interarrival times = sojourn time

Let's prove that $S_0 \sim \text{exp}(\lambda)$

$$\mathbb{P}[S_0 > t] = \mathbb{P}[X(t) = 0] = \frac{(\lambda t)^0 e^{-\lambda t}}{0!} = e^{-\lambda t} \Rightarrow S_0 \sim \text{exp}(\lambda)$$

Since an exponent r.v. has mean $\frac{1}{\lambda}$, so is $E(S_0)$

$\mathbb{P}[w_1 \leq s, w_2 > t]$ where w_i is waiting time
and $0 \leq s < t$

$$\begin{aligned} \mathbb{P}[X(s) = 1, X(t) = 1] &= \mathbb{P}\{X(s) = 1, X(t) - X(s) = 0\} = \\ &= \mathbb{P}[X(s) = 1] \cdot \mathbb{P}[X(t) - X(s) = 0] \\ &\stackrel{\text{independent}}{=} \frac{(\lambda s)^1 e^{-\lambda s}}{1!} \cdot \frac{\lambda(t-s)}{0!} e^{-\lambda(t-s)} = \lambda s e^{-\lambda t} \end{aligned}$$

$$\text{Instead, } \mathbb{P}[w_1 \leq s, w_2 \leq t] = \mathbb{P}[w_1 \leq s] - \mathbb{P}[w_1 \leq s, w_2 > t] = 1 - e^{-\lambda s} - \lambda s e^{-\lambda t}$$

$= F_{w_1, w_2}(s, t) \Rightarrow$ density of r.v. (w_1, w_2) is:

$$\begin{aligned} f_{w_1, w_2}(s, t) &= \frac{\partial}{\partial s} \frac{\partial}{\partial t} F_{w_1, w_2}(s, t) = \frac{\partial}{\partial s} \frac{\partial}{\partial t} (1 - e^{-\lambda s} - \lambda s e^{-\lambda t}) \\ &= \frac{\partial}{\partial s} \frac{\partial}{\partial t} (-\lambda s e^{-\lambda t}) \\ &= \lambda^2 e^{-\lambda t}, \quad 0 \leq s < t \end{aligned}$$

Let us make this change of variables: $s_0 = w_1$, $s_1 = w_2 - w_1$,
 $\Rightarrow w_1 = s_0$, $w_2 = s_1 + s_0 \Rightarrow (s_0, s_1) \in (0, +\infty)^2$

and

$$f_{s_0, s_1}(v, n) = |\det \bar{J}| \cdot f_{w_1, w_2}(v, v+n), \text{ where } \bar{J} = \begin{bmatrix} 1 & 0 \\ 1 & 2 \end{bmatrix}, |\det \bar{J}| = 1$$

$$f_{s_0, s_1}(v, n) = \lambda^2 e^{-\lambda(v+n)} = \lambda e^{-\lambda v} \cdot \lambda e^{-\lambda n} = f_{s_0}(v) \cdot f_{s_1}(n)$$

$\Rightarrow s_0 \perp s_1$ & $s_0 \sim \exp(\lambda)$, $s_1 \sim \exp(\lambda)$

$$\Rightarrow \mathbb{E}[s_0] = \mathbb{E}[s_1] = \frac{1}{\lambda}$$

The general case for s_0, \dots, s_m is similar

Q6: prove that if i and j of a TC communicate and i is recurrent, then j is also recurrent

A6: by h_p , $i \leftrightarrow j \Rightarrow \exists m, m > 1$ s.t.

$$P_{ij}^{(m)} > 0; P_{ji}^{(m)} > 0$$

Let $\lambda > 0$:

$$\sum_{k=0}^{+\infty} P_{jj}^{(n+k+2)} \geq \sum_{k=0}^{+\infty} P_{ji}^{(n)} \cdot P_{ii}^{(2)} \cdot P_{ij}^{(n)}$$

$$= P_{ji}^{(n)} \cdot P_{ii}^{(2)} \sum_{k=0}^{+\infty} P_{ii}^{(2)}$$

Since i is recurrent by hypothesis, then

$$\sum_{j=0}^{+\infty} P_{ii}^{(j)} = +\infty \Rightarrow \sum_{j=0}^{+\infty} P_{jj}^{(j)} = +\infty$$

; i is recurrent

prob. because

$$\sum_{j=0}^{+\infty} P_{jj}^{(n+m+2)} \geq \sum_{j=0}^{+\infty} P_{ii}^{(j)} = +\infty$$

Q 8: prove that for a PP $X(t)$ the statistic of $X(s)$ conditioned on $X(t), s < t, k \leq m$, is binomial and provide the expression of $P[X(s)=k | X(t)=m]$

$$\begin{aligned}
 P[X(s)=k | X(t)=m] &= \frac{P[X(s)=k, X(t)=m]}{P[X(t)=m]} \\
 &= \frac{P[X(s)=k, X(t)-X(s)=m-k]}{P[X(t)=m]} = \\
 &= P[X(s)=k] \cdot P[X(t)-X(s)=m-k] \cdot \frac{1}{P[X(t)=m]} \\
 &= \frac{(\lambda s)^k e^{-\lambda s}}{k!} \cdot \frac{(\lambda(t-s))^{m-k} e^{-\lambda(t-s)}}{(m-k)!} \cdot \frac{\frac{m!}{(\lambda t)^m e^{\lambda t}}}{(\lambda t)^m} = \binom{m}{k} \cdot \frac{(\lambda s)^k (\lambda(t-s))^{m-k}}{(\lambda t)^m} \\
 &= \binom{m}{k} \cdot \frac{s^k (t-s)^{m-k}}{t^m - \boxed{t^m \cdot t^{m-k}}} = \boxed{\binom{m}{k} \left(\frac{s}{t}\right)^k \cdot \left(1 - \frac{s}{t}\right)^{m-k}} \\
 &\Rightarrow P[X(s)=k | X(t)=m] = \text{Bin}(m, s/t)
 \end{aligned}$$

Q10: For a Renewal process state precisely and prove what is the value of $\lim_{t \rightarrow +\infty} \frac{N(t)}{t}$

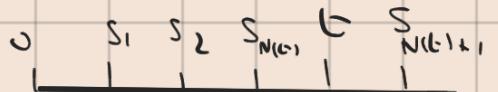
Note: $N(\infty) = \infty$ with prob. 1:

$$\begin{aligned} P[N(\infty) < \infty] &= P[x_n = \infty \text{ for some } n] = \\ &= P\left[\bigcup_{n=1}^{+\infty} \{x_n = +\infty\}\right] \leq \sum_{n=1}^{+\infty} P[x_n = +\infty] = 0 \end{aligned}$$

↓
cannot have
finite renewals in
an infinite time

$N(t) \rightarrow +\infty$ as $t \rightarrow +\infty$

$$\lim_{t \rightarrow +\infty} \frac{N(t)}{t} = \frac{1}{\mu} \quad \text{with prob. 1}$$



Proof: since $s_{N(t)} \leq t < s_{N(t)+1} \rightarrow 0$

$$\Rightarrow \frac{s_{N(t)}}{N(t)} \leq \frac{t}{N(t)} < \frac{s_{N(t)+1}}{N(t)}$$

$\frac{s_{N(t)}}{N(t)}$ is the average of the first $N(t)$ interarrival times

then by the strong LLN: $\frac{s_{N(t)}}{N(t)} \rightarrow \mu$ vs $N(t) \rightarrow +\infty$

but also $\frac{s_{N(t)}}{N(t)} \rightarrow \mu$ vs $t \rightarrow +\infty$

$$\frac{s_{N(t)+1}}{N(t)} = \frac{s_{N(t)+1}}{N(t)+1} \cdot \frac{N(t)+1}{N(t)} \rightarrow \text{then we obtain}$$

$$\frac{S_{N(t)+1}}{N(t)} \rightarrow \mu \text{ as } t \rightarrow \infty$$

By the combination law we get also that

$$\frac{t}{N(t)} \rightarrow \mu \text{ as } t \rightarrow \infty$$

or

$$\frac{N(t)}{t} \rightarrow \frac{1}{\mu} \text{ as } t \rightarrow \infty$$

Q12 : same as Q8

Q14 : Give a definition of recurrent state and prove that a state is recurrent (\Rightarrow

$$\sum_{n=1}^{\infty} P_i^{(n)} = \infty$$

We say that a state i is recurrent

$$\text{if } f_{ii} = 1$$

\hookrightarrow prob. of returning to i

$$\text{if recurrent } \Leftrightarrow \sum_{n=1}^{\infty} P_{ii}^{(n)} = +\infty$$

Proof: i recurrent $\Rightarrow f_{ii} = 1$

η counts the total number of visits to state i

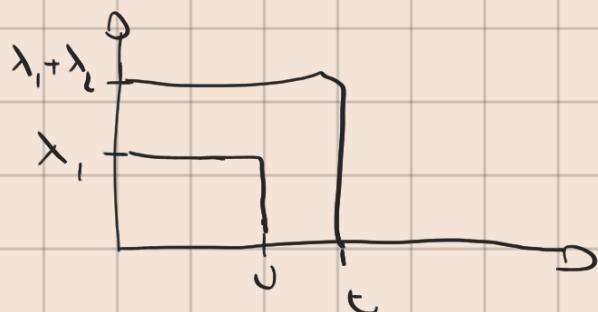
$$M = \sum_{n=1}^{\infty} 1 \mathbb{I}_{\{X_n=i\}} \xrightarrow{\text{D}} 1$$

$$\mathbb{E}[M | X_0=i] = \frac{1}{1 - p} = +\infty$$

so

$$\begin{aligned}
 +\infty &= \mathbb{E}[n | X_0=i] = \mathbb{E}\left[\sum_{n=1}^{\infty} \mathbb{1}_{\{X_n=i\}} | X_0=i\right] \\
 &= \sum_{n=1}^{\infty} \mathbb{E}[\mathbb{1}_{\{X_n=i\}} | X_0=i] \\
 &\quad \mid \mathbb{E}[\mathbb{1}_A] = P[A] \\
 &= \sum_{n=1}^{\infty} P[X_n=i | X_0=i] = \sum_{n=1}^{\infty} P_{ii}^{(n)}
 \end{aligned}$$

Q16: For two PP $X_1(t)$ and $X_2(t)$ with rates λ_1, λ_2 respectively, derive the expression of $P[X_1(u)=k | X_1(t)+X_2(t)=m]$ for $0 \leq u < t$, $0 \leq k \leq m$



$$X(t) = X_1(t) + X_2(t)$$

$$\begin{aligned}
 P[X_1(u)=k | X(t)=m] &= \frac{P[X_1(u)=k, X_1(t)-X_1(u)+X_2(t)=m-k]}{P[X(t)=m]} \\
 &= P[X_1(u)=k] \cdot P[X_1(t)-X_1(u)+X_2(t)=m-k] \\
 &\quad \cdot \overbrace{P[X(t)=m]}^1
 \end{aligned}$$

$$P = \frac{(\lambda_1 u)^k e^{-\lambda_1 u}}{k!} \cdot \frac{(\lambda_1(t-u) + \lambda_2 t)^{n-k} e^{-(\lambda_1(t-u) + \lambda_2 t)}}{(n-k)!}$$

Indep. $\Rightarrow \frac{n!}{(t(\lambda_1 + \lambda_2))^n} e^{-t(\lambda_1 + \lambda_2)} =$

$$= \boxed{\binom{n}{k} \left(\frac{\lambda_1 \cdot u}{t(\lambda_1 + \lambda_2)} \right)^k \cdot \left(\frac{\lambda_1(t-u) + \lambda_2 t}{t(\lambda_1 + \lambda_2)} \right)^{n-k}}$$

Q1S: Prove that for a Renewal process
 $M(t) < \infty \quad \forall t \text{ finite}$

We start by introducing a definition

$$\boxed{F_n(t)} = \int_0^t F_{n-m}(t-\xi) dF(\xi) \leq \boxed{F_{n-m}(t) F_m(t)} \quad 1 \leq m \leq n-1$$

The

$$M(t) = \sum_{j=1}^{+\infty} F_j(t) = \sum_{n=0}^{+\infty} \sum_{k=1}^n F_{n+k}(t)$$

$$\leq \sum_{n=0}^{+\infty} \sum_{k=1}^n (F_n(t))^m F_k(t) =$$

$$= \sum_{n=0}^{+\infty} (F_n(t)^m) \left(\sum_{k=1}^n F_k(t) \right)$$

$\leq n$ D prob. so ≤ 1

Converges geometrically if $F_n(t) < 1$

There is a Th that states that $\forall t, \exists n$ s.t. $F_n(t) < 1$

