

$a(\varepsilon, \delta)$

RENEWAL THEORY

⇒ Stochastic process $\{M(t)\}_{t \geq 0}$

→ Counting process, $M(0) = 0$
 $M(t) \in \{0, 1, 2, \dots\}$

$$M(t) \leq M(t+\tau) \quad \forall \tau \geq 0$$

→ each jump indicates an \uparrow in count.

⇒ x_1, x_2, \dots are called the lifetimes

(eg. - turning on bulbs until they die & replacing them later.)

⇒ Life Times are INDEPENDENT

⇒ $\{x_2, x_3, \dots\}$ are iid variables
 x_1 can be different because the time at which you start observing is almost unknown?

$$x_1 \sim A, x_2 \sim F$$

some distribution f^h

Define, Z_k ,

time of
the k^{th} -increment

$$Z_k = \sum_{i=1}^k X_i$$

\Rightarrow Probability law for $\{M(t)\}_{t \geq 0} \Rightarrow$

$$P(M(t) \geq n) = P(Z_n \leq t)$$

i.e., say, $P(M(t) \geq 1) = P(Z_1 \leq t) = P(X_1 < t) = A(t)$

(if at a time t , more than one bulb's are on,
it means that the 1^{st} transition must've occurred
before t ' itself.)

$$P(M(t) \geq 2) = P(Z_2 \leq t) = P(X_1 + X_2 \leq t)$$

$$= \int_0^t P(X_1 + X_2 \leq t | X_1 = u) \cdot (dA(u))$$

$$= \int_0^t P(X_2 \leq t-u | X_1 = u) \cdot (dA(u))$$

$$= \int_0^t P(X_2 \leq t-u) \cdot (dA(u))$$

(since X_1, X_2 are p.i.d. RV's)

$$P(X_1 + X_2 \leq t) = \int_0^t F(t-u) d(A(u)) = F^* A(t)$$

$$F^* A(t) = \int_0^t F(t-u) \cdot d(A(u))$$

$$\Rightarrow P(M(t) \geq 3) = P(X_1 + X_2 + X_3 \leq t)$$

$$= \int_0^t (A^* F)(t-u) \cdot d(F(u))$$

⇒ since x_1, x_2, x_3, \dots have some probability distribution, at every jump the process renews itself.

⇒ If we start observing from a jump point, we simply renew the process.

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$$= \int_0^t P(X_2 + X_3 \leq t-u) \cdot d(A(u)).$$

$$= \int_0^t \frac{P(X_3 \leq t-u)}{F^*(A(t-u))} \cdot d(F(u)).$$

$$= \int_0^t A^* F(t-u) \cdot d(F(u)) \Rightarrow P(M(t) \geq 3) = (A^* F^{(2)})(t).$$

so,

$$P(M(t) \geq n) = (A^* F^{(n-1)})(t)$$

Renewal equation :- let

↓ renewal fnc

$$m(t) = E[M(t)]$$

$m: [0, \infty) \rightarrow [0, \infty)$
(deterministic; renewal fnc)

→ At renewal points / epochs, you can start looking at the process beyond that point and it will look the same!

→ $M(t)$ in itself is random, but has a fixed expectation.

$$m(t) = \sum_{R=0}^{\infty} R \cdot P(M(t) = R)$$

$$= \sum_{k=0}^{\infty} P(M(t) \geq k)$$

(alternate defⁿ of expectation value)

[discrete, non-negative RV's]

$$m(t) = \sum_{R=1}^{\infty} (A^* F^{(R-1)})(t)$$

(all x_j have the same distribution for $j \geq 2$)

→ Here, we need to try to separate out x_1 and the others.

If $E[X_j] = 0$ ($\forall j \geq 2$) $\Rightarrow P(X_j=0) = 1$

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all events after x_j happen at same time
at x_1 , and the process doesn't proceed further

since, $E[X_j] > 0 \Rightarrow$ Then, $\exists (\delta > 0)$ s.t. $P(X_j > \delta) > 0 \forall j \geq 2$

From the Borel-Cantelli Lemma,

$$P(X_j \leq \delta \text{ i.o.}) = 1$$

$$B_j = \{X_j \leq \delta\}$$

$$P(B_j) > 0$$

$$B_j^c = \{X_j > \delta\}$$

$$P(B_j^c) > 0$$

PROOF :- For any $n \geq 1$ and $0 \leq m \leq n$, we have,

$$F^{(n)}(t) \leq F^{(n-m)}(t) \cdot F^{(m)}(t)$$

$$P(X_1 + X_2 + \dots + X_{n+1} \leq t)$$

(n-m) RV's

(m) RV's \Rightarrow iid sequence

$P((n-m) \text{ Variables} \leq t)$ and $P(m \text{ variables} \leq t)$

has more soln than $P(n \text{ variables} \leq t)$.

Mathematical Proof :-

$$F^{(n)}(t) = \int_0^t F^{(n-m)}(t-u) dF^{(m)}(u)$$

$$\leq \int_0^t F^{(n-m)}(t) dF^{(m)}(u)$$

$$\leq [F^{(n-m)}(t) \cdot F^{(m)}(t)]$$

Hence proved

Note:- $F^{(nr+k)}(t) \leq F^{((r-1)r+k)}(t) \cdot F^{(r)}(t)$

applying again and again,

$$F^{(nr+k)}(t) \leq (F^{(r)}(t))^n \cdot F^{(r)}(t)$$

$$\leq \sum_{n=0}^{\infty} \sum_{k=0}^{\infty} F^{(k)}(t) \cdot (F^{(n)}(t))^n \leq \sum_{n=0}^{\infty} \sum_{k=0}^{\infty} 1 \cdot (F^{(k)}(t))^n$$

$$\leq \sum_{n=0}^{\infty} r (F^{(n)}(t))^n$$

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$$m(t) = \sum_{n=1}^{\infty} (A^* F^{(n)}) (t)$$

$$\leq \sum_{n=0}^{\infty} F^{(n)}(t)$$

$$= \sum_{n=0}^{\infty} \sum_{R=0}^{nr} F^{(nr+R)}(t)$$

$$A^* F^{(n+1)}(t) \leq A(t) F^{(n+1)}(t)$$

$$\leq F^{(n+1)}(t) + t$$

Sum converges
if $r < 1$

$$\leq r F^{(nr)}(t) \quad (\text{true if } r > 1)$$

$$F^{(r)}(t) < 1 \quad \leq \sum_{n=0}^{\infty} r (F^{(r)}(t))^n = r \sum_{n=0}^{\infty} (F^{(r)}(t))^n \quad m(t) < \infty \quad t \in [0, \infty)$$

So, For any given t , $\exists r > 1$ s.t. $F^{(r)}(t) < 1$
 If $\exists a, P(X_2 + \dots + X_{r+1} \leq t) = 1 + r \geq 1$ we should
 $P(X_2 \leq t) = 1 \leftarrow$ have a (i.e. $P(X_2 + \dots + X_{r+1} \leq t) < 1$)
 $P(X_2 + X_3 \leq t) = 1 \rightarrow P(X_3 > \delta) > 0 \rightarrow P(X_2 + X_3 + \dots + X_{r+1} > t) > 0$ (For some δ)
 and for this r , $P(X_2 + \dots + X_{r+1} \leq t) < 1$

Elementary Renewal Theory :-

⇒ Also called the Strong Law of Large Numbers for Renewal Processes

Theorem :- Consider the following :-

① $R \geq 1$, $P(X_k < \infty) = 1$ (all life times are proper RV's)

② $0 \leq E[X_1] < \infty$

③ $R \geq 2$, $E[X_R] < \infty$, let $E[X_2] = \frac{1}{\mu}$

Under these conditions, we have

a) $\lim_{t \rightarrow \infty} \frac{m(t)}{t} = \mu$

b) $\lim_{t \rightarrow \infty} \frac{M(t)}{t} = \mu$

almost surely (or w.p. 1)

For every w , it is a fnc of t $\left\{ \begin{array}{l} M(t) \\ t \end{array} \right\}_{t \geq 0} \rightarrow$ (deterministic process for) some fixed w

$$\lim_{t \rightarrow \infty} M(t) = \mu \text{ w.p. 1}$$

there can be some w for which the limit may not exist \Rightarrow w.p. 1 (AS)

$$P\left(\{w : \lim_{t \rightarrow \infty} \frac{M(t, w)}{t} = \mu\}\right) = 1$$

$$P\left(\{w : \lim_{t \rightarrow \infty} \frac{M(t, w)}{t} = \mu\}\right) = 1$$

$$\lim_{t \rightarrow \infty} \frac{m(t)}{t} = \frac{1}{E[X_2]} = \frac{1}{\mu}$$

avg. period after avg. life time \Rightarrow which a renewal would occur

$$m(t)/t$$

$$E[X_2] = \lambda u$$

rate of renewal = actual rate at which renewals actually happen in the system.

as $t \rightarrow \infty$, the

rate of renewal equals the expected renewal per unit time

Strong LLN $\Rightarrow Z_n = \sum_{k=1}^n X_k$, $\lim_{n \rightarrow \infty} \frac{Z_n}{n} = \frac{1}{\mu}$ w.p. 1

strong law of large numbers

Can we say that

$$\lim_{t \rightarrow \infty} \frac{Z_{M(t)}}{M(t)} = \frac{1}{\mu}$$
 w.p. 1 ?

$Z_{M(t)}$ → Time of the last renewal before or after 't'

$Z_{M(t)+1}$ → Time of the first renewal after t. We need $M(t) \rightarrow \infty$ as $t \rightarrow \infty$.

when we say \Rightarrow

$$\lim_{t \rightarrow \infty} g(t) = \infty$$

+ seq. $\{t_n\}_{n \geq 1}$ s.t. $\lim_{n \rightarrow \infty} t_n = \infty$

$$\lim_{n \rightarrow \infty} g(t_n) = a$$

$$\lim_{n \rightarrow \infty} \frac{Z_{M(t_n)}}{m(t_n)}$$

$Z_{M(t)}$: the time instance of the last renewal before/after t



$$m(t_1) = 1$$

$$m(t_2) = 2$$

$$m(t_3) = 4$$

$$Z_{M(t_1)} = Z_1$$

$$Z_{M(t_3)} = Z_4$$

$Z_{M(t)+1}$ → time instance of the first renewal after t.

if $\{t_n\}$ is a subset of

$$\left\{ \frac{Z_{M(t_n)}}{m(t_n)} \right\}_{n \geq 1}$$

(it is an w-wise convergence)

$$\left\{ \frac{Z_n}{n} \right\}_{n \geq 1}$$

$$\lim_{t \rightarrow \infty} M(t) = \infty \text{ almost surely}$$

→ CLAIM

as $\left\{ \frac{Z_n}{n} \right\}$ converges, so does

$$\left\{ \frac{Z_{M(t)}}{M(t)} \right\}$$

Suppose NOT,

$$\text{let } P\left(\lim_{t \rightarrow \infty} m(t) = R\right) \Leftrightarrow P(X_1 < \infty, \dots, X_K < \infty, X_{K+1} = \infty)$$

↓ IID

$$P(X_{K+1} = \infty) = 0$$

(as $P(X_n < \infty) = 1 \forall n \geq 1$)

$$\begin{aligned}
 P\left(\lim_{t \rightarrow \infty} M(t) < \infty\right) &= P\left(\exists k < \infty \text{ s.t. } \lim_{t \rightarrow \infty} M(t) = k\right) \\
 &= P\left(\bigcup_{k=1}^{\infty} \left\{ \lim_{t \rightarrow \infty} M(t) = k \right\}\right) \\
 &\leq \sum_{k=1}^{\infty} P\left(\lim_{t \rightarrow \infty} M(t) = k\right) = 0
 \end{aligned}$$

(union-bound)

(each case is
mutually disjoint).
also

as $P\left(\lim_{t \rightarrow \infty} M(t) < \infty\right) = 0$

and $\lim_{t \rightarrow \infty} M(t) = \infty$ w.p. 1 (almost surely)

$$\lim_{t \rightarrow \infty} \frac{Z_{M(t)}}{M(t)} = \frac{1}{u} \quad \text{w.p. 1}$$

(as $\lim_{n \rightarrow \infty} \frac{z_n}{n} = \frac{1}{u}$, $\lim_{t \rightarrow \infty} n = \infty$)

Let $A = \left\{ \omega : \frac{z_n(\omega)}{n} \rightarrow \frac{1}{u} \right\}$ ($P(A) = 1$).

$B = \left\{ \omega : \lim_{t \rightarrow \infty} M(t) = \infty \right\}$ ($P(B) = 1$)

Also, if $w \in A \cap B$, then,

$$\frac{z_{M(t)}(\omega)}{M(t)(\omega)} \rightarrow \frac{1}{u} \Rightarrow \left\{ \omega : \frac{z_{M(t)}}{M(t)} \rightarrow \frac{1}{u} \right\}$$

\Rightarrow is a superset of $A \cap B$

also, $\Rightarrow P\left\{ \omega : \frac{z_{M(t)}}{M(t)} \rightarrow \frac{1}{u} \right\} \geq P(A \cap B) = 1$

$$\frac{z_{M(t)}}{M(t)} \leq \frac{t}{M(t)} \leq \frac{z_{M(t)+1}}{(M(t)+1)} \cdot \frac{M(t)+1}{M(t)}$$

(using sandwich theorem)

$$\frac{1}{u} \rightarrow \frac{1}{u}$$

$$z_{M(t)} \leq t \leq z_{M(t)+1}$$

so, $\frac{M(t)}{t} \rightarrow u$ w.p. 1 as $t \rightarrow \infty$

We also must have $\lim_{n \rightarrow \infty} \frac{x_n}{n} \rightarrow 0$ w.p. 1

$$Z_{M(t)} \leq t$$

$$E[Z_{M(t)}] \leq t$$

$$E\left[\sum_{k=1}^{M(t)} X_k\right] \leq t$$

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$$Z_n = X_1 + \sum_{k=2}^n X_k$$

$$\frac{Z_n}{n} = \frac{X_1}{n} + \frac{1}{n} \sum_{k=2}^n X_k$$

$$\frac{Z_n}{n} \rightarrow 0$$

$$\frac{Z_n}{n} = \frac{X_1}{n} + \frac{(n-1)}{(n)} \cdot \frac{1}{(n-1)} \sum_{k=2}^n X_k$$

$$\frac{1}{n} \leftarrow \text{now} \rightarrow 0 \rightarrow Y_u$$

Now,

$$E\left[\lim_{t \rightarrow \infty} \frac{M(t)}{t}\right] = \lim_{t \rightarrow \infty} E\left(\frac{M(t)}{t}\right)$$

$$\text{For iid r.v's } E\left[\sum_{k=1}^n X_k\right] = E[X_k] \times n$$

$$\text{and } E\left[\sum_{k=1}^N X_k\right] = E[N] E[X_k]$$

$$E[Z_{M(t)}] = E\left[\sum_{k=1}^{M(t)} X_k\right]$$

$$E[Z_{M(t)}] = E[M(t)] \cdot E[X_k]$$

$$E[Z_{M(t)}] = m(t) E[X_k]$$

Note:- $t \geq Z_{M(t)}$

$$\frac{m(t)}{t} \leq u + t$$

$$E[Z_{M(t)+1}] = \frac{(m(t)+1)}{u} + t < \frac{m(t)+1}{u} + t \Rightarrow u < \frac{m(t)+1}{t}$$

$$\frac{m(t)}{t} \leq u \leq \frac{m(t)+1}{t}$$

(The result now follows from SANDWICH THM.)

$$\begin{aligned}
 \Rightarrow \text{Now, we have } E\left[\sum_{k=1}^N X_k\right] &= \sum_{n=1}^{\infty} E\left[\sum_{k=1}^n X_k \mid N=n\right] \cdot P(N=n) \\
 &= \sum_{n=1}^{\infty} E\left[\sum_{k=1}^n X_k\right] P(N=n). \quad [E[X_k] \text{ is the same for all } k] \\
 &= \sum_{n=1}^{\infty} n E[X_1] P(N=n) \\
 &= E[X_1] E[N]
 \end{aligned}$$

- This can be done only if X_k 's and N are independent.
- ⇒ But, in the previous proof, $M(t)$ and X_k are NOT independent.

WALD'S LEMMA :-

Let N be a stopping time for process $\{X_n\}_{n \geq 1}$ if

- i) $E[N] < \infty$
- ii) $E[|X_n|] \leq B$ (constant) $< \infty$
- iii) $E[X_n] = E[X_1] + n \geq 2$

Then, again, $E\left[\sum_{k=1}^N X_k\right] = E[N] E[X_1]$

Proof:- $E\left[\sum_{k=1}^N X_k\right] = E\left[\sum_{k=1}^{\infty} I_{\{K \leq N\}} \cdot X_k\right]$

$\uparrow 1 \text{ if } K \leq N, 0 \text{ o/w}$

as N is a stopping time, $I_{\{K \leq N\}}$ is a f.c. of x_1, x_2, \dots, x_{k-1}

$$(I_{\{K \leq N\}} = I_{\{k \leq N\}} = g(x_1, \dots, x_{k-1}))$$

assuming that we can switch E and \sum ,

$$\begin{aligned}
 \sum_{R=1}^{\infty} E[X_k] E[I_{\{N \geq R\}}] &= E[X_1] \sum_{R=1}^{\infty} P(N \geq R) \\
 &= E[X_1] E[N]
 \end{aligned}$$

→ Proof for the swapping

$$\text{Let } Y_n = \sum_{k=1}^n X_k I_{\{k \leq n\}}$$

$\{Y_n\}$ are monotone if X_n is positive,

However Wald's lemma doesn't say X_n should be positive.

DOMINATED CONVERGENCE THEOREM :-

$\{Y_n\}$ s.t. $|Y_n| \leq Y$ for some Y with $E[Y] < \infty$, then,

$$\lim E[Y_n] = E[\lim Y_n]$$

Now, let's define, $Y = \sum_{k=1}^{\infty} |X_k| I_{\{N \geq k\}}$

$$E[Y] = \sum_{k=1}^{\infty} E(|X_k|) P(N \geq k)$$

monotones

By BCT, $\leq E[N] \cdot B < \infty \Rightarrow E$ and E can be swapped

$$|Y_n| = \left| \sum_{k=1}^n X_k I_{\{N \geq k\}} \right| \leq \sum_{k=1}^n |X_k| I_{\{N \geq k\}}$$

(Trivial Δ -inequality)

$|Y_n| \leq Y$ and $E[Y] < \infty \Rightarrow E$ and \sum can be

swapped for Y_n even if

limit of \sum is ∞ .

→ IS $M(t)$ A STOPPING TIME?

To use Wald's lemma on $\sum X_k$, it must be a stopping time

$$\{M(t) > n\} = \{M(t) \geq n+1\} = \{Z_{n+1} \leq t\}$$

$= g(x_1, x_2, \dots, x_{n+1})$ on Z_{n+1}
depends on them.

⇒ However, if $M(t)$ is a stopping time, it should depend only on $\{x_1, \dots, x_n\}$, and not x_{n+1} .

⇒ Let's see $\{M(t)+1 > n\} = \{M(t) \geq n\}$

$$= \{z_n \leq t\}$$

$$= g(x_1, \dots, x_n)$$

$M(t)+1$ is a stopping time!

Then, $E[Z_{M(t)+1}] = E\left[\sum_{k=1}^{M(t)+1} X_k\right] = E[M(t)+1] E[X_k]$

$$= (m(t)+1) \cdot \frac{1}{\mu}$$

as $t < Z_{M(t)+1}$

$$t < E[Z_{M(t)+1}]$$

$$t < (m(t)+1) \cdot \frac{1}{\mu}$$

$$\mu < \frac{m(t)}{t} + \frac{1}{t} + t$$

(LIMIT MIGHT NOT HOLD somewhere)

Consider the sequence $\{a_n\}$, $\limsup_{n \rightarrow \infty} a_n$ exists.

$$\Rightarrow b_n = \sup_{k \geq n} q_k = \sup \{a_n, a_{n+1}, \dots\}$$

$$\lim_{n \rightarrow \infty} \sup a_n = \lim_{n \rightarrow \infty} b_n$$

(exists because b_n decreases monotonically)

$$\Rightarrow \text{similarly, let } c_n = \inf_{n \geq k} q_k$$

$$\text{and, } \lim_{n \rightarrow \infty} \inf a_n = \lim_{n \rightarrow \infty} c_n \rightarrow c_n \text{ is increasing}$$

Now, if $\liminf_{n \rightarrow \infty} a_n = \limsup_{n \rightarrow \infty} a_n = a$, then $\lim_{n \rightarrow \infty} a_n = a$

$$\text{we have } u \leq \frac{m(t)}{t} + \frac{1}{t} \quad \forall t$$

$$u \leq \liminf_{t \rightarrow \infty} \frac{m(t)}{t} + \liminf_{t \rightarrow \infty} \frac{1}{t}$$

$$u \leq \liminf_{t \rightarrow \infty} \frac{m(t)}{t}$$

\Rightarrow For other side of the inequality

$$u \geq Z_M(t)$$

$t > E[Z_M(t)]$ \rightarrow $Z_M(t)$ is not a stopping time

\hookrightarrow Wald's Lemma (\times)

\rightarrow constructing another renewal process with lifetimes $\{X_n\}_{n \geq 1}$

$$t < Z_{N(t)+1}$$

Suppose $\exists c$ s.t. $Z_{N(t)+1} \leq t+c$

(A new random time)

$$P(X_1 = 1) = P(X_1 = 2) = \dots = P(X_1 = n) = \dots$$

$$P(X_1 = 1) = P(X_1 = 2) = \dots = P(X_1 = n) = \dots$$

(construction of a new renewal function $\psi(n)$)

(no stationary)

$$P(X_1 = 1) = P(X_1 = 2) = \dots = P(X_1 = n) = \dots$$

problem 2: $\psi(n) \leq \min_{1 \leq i \leq n} \psi(i)$ than

Review :-

$\{M(t)\}_{t \geq 0}$ is a counting process, with $M(0) = 0$

Z_k denote the k^{th} -jump epoch of $M(t)$, then,

$$Z_k = \sum_{j=1}^k X_j, \quad \text{where } \{X_j\}_{j \geq 1} \text{ are called lifetimes.}$$

$P(X_j > 0) = 1$, X_j 's are independent or $\{X_j\}_{j \geq 2}$ are i.i.d.

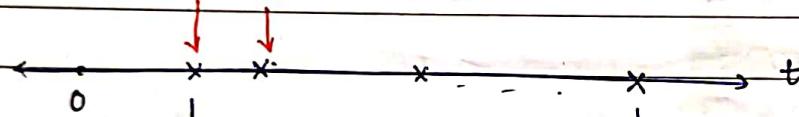
\Rightarrow Elementary Renewal Theory : $\lim_{t \rightarrow \infty} \frac{M(t)}{t} = u$ w.p. 1 and

$$\lim_{t \rightarrow \infty} \frac{M(t)}{t} = u \quad \text{where } E[X_2] = \frac{1}{u}$$

Renewal Reward Process

fare at end of
1st trip

(R₁) (R₂)



eg:- Uber-driver

1st trip
ends here

nth
trip ends
here



lifetimes

Frequency of trips = ? $\rightarrow E[X_2]$

Model for the fare

pre-paid

part-paid

keep getting reward per unit time

Process :- $\{X_j, R_j\}_{j \geq 1}$

lifetimes

rewards

\Rightarrow More the lifetime, more the rewards.

$\rightarrow R_j$ may depend on $X_j \equiv R_j = f(X_j)$

(Joint distributions also)
must be considered

(not necessary linear)

NOT monotone
relⁿ necessary

Reward Process :- $\{R_j\}_{j \geq 1}$

$\Rightarrow R_j$ depends on X_j but not on $X_i \neq j$ (typically)
or,

$$R_j \text{ iff } X_i \neq j$$

(as $X_i \perp\!\!\!\perp X_j \neq i$) and if RV's are independent, their functions are also independent.

$\Rightarrow \{R_j\}_{j \geq 1}$ is also a seq' of independent RV's

Also $\{X_j\}_{j \geq 2}$ are iid $\Rightarrow \{R_j\}_{j \geq 2}$ are also i.i.d. RV's

\Rightarrow The joint distribution MUST be the same (R_j, X_j) can be a reward/ a penalty
(No restriction to be non-negative)

\Rightarrow Let $C(t)$ denote the total reward earned until time t

$$C(t) = \sum_{j=1}^{M(t)} R_j : \text{POST-PAID}$$

$$C(t) = \sum_{j=1}^{M(t)+1} R_j : \text{PRE-PAID}$$

$$C(t) = \sum_{j=1}^{M(t)} R_j + R(t) : \text{UNIT-TIME reward model.}$$

↓
partial reward at t

RENEWAL REWARD THEOREM

If $\{(X_j, R_j)\}_{j \geq 1}$ is a renewal reward process, with $E[R_1] < \infty$ and $0 < E[X_1] < \infty$ and $E[X_2] > 0$ ($< \infty$), then,

a) $\lim_{t \rightarrow \infty} \frac{C(t)}{t} = \frac{E[R_1]}{E[X_1]}$ w.p. 1

b) $\lim_{t \rightarrow \infty} \frac{c(t)}{t} = \frac{E[R_1]}{E[X_1]}$ w.p. 1 where, $c(t) = E[C(t)]$

say $R_j = 1$ w.p. 1 $\Rightarrow C(t) = M(t) \rightarrow \lim_{t \rightarrow \infty} \frac{M(t)}{t} = \frac{1}{E[X_1]} \neq E\left[\frac{1}{X_1}\right]$

PROOF : $C(t) = \sum_{k=1}^{M(t)} R_k$

$$\frac{C(t)}{t} = \frac{1}{t} \sum_{k=1}^{M(t)} R_k = M(t) \cdot \frac{1}{t} \sum_{k=1}^{M(t)} R_k$$

and as $t \rightarrow \infty$ $\frac{1}{t} \rightarrow 0$

$$\frac{1}{t} \sum_{k=1}^{M(t)} R_k \xrightarrow{\text{w.p. 1}} E[R_2]$$

by ERT

strong Law of Large No.
(Strong LLN)

$$P(X_1 < \infty) = 1$$

$$\infty > E[X_2] > 0$$

$$P(X_2 = 0) < 1$$

$$\Rightarrow \exists A \subseteq \Omega, A \in \mathcal{F} \text{ and } P(A) = 1$$

$$\lim_{t \rightarrow \infty} \frac{M(t, \omega)}{t} = \frac{1}{E[X_2]}$$

and, $\exists B \subseteq \Omega, B \in \mathcal{F}, P(B) = 1$ and

$$\lim_{t \rightarrow \infty} \frac{\sum_{k=1}^{M(t, \omega)} R_k(\omega)}{M(t, \omega)} = E[R_2]$$

$\forall \omega \in (A \cap B)$.

since the probabilities are exactly 1, it does NOT matter.

(set of $P=0$ is independent of everything else
set of $P=1$ i.e. $\omega \in A \cap B$)

EXAMPLE OF Renewal Reward Process

Define a residual time process

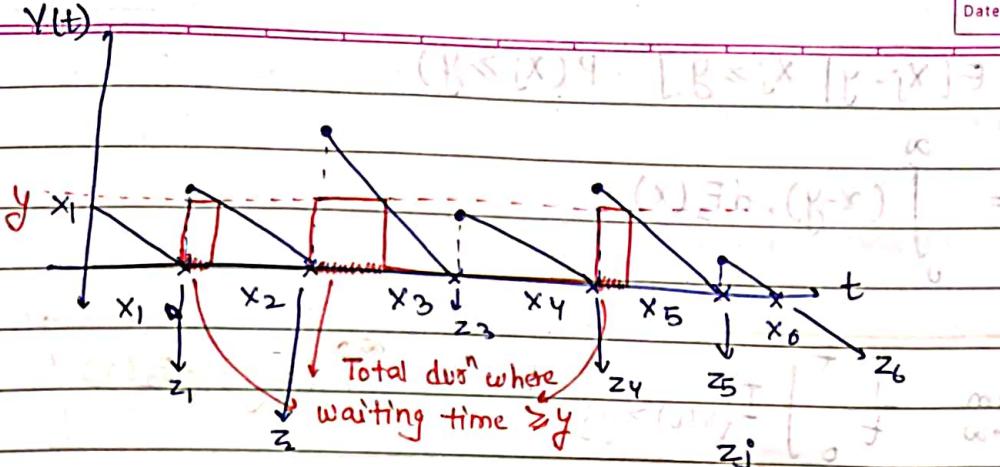
$$\text{eg. } \{Y(t)\}_{t \geq 0} \text{ as } Y(t) = Z_{M(t)+1} - t$$

(Time until the next renewal)

$$\lim_{t \rightarrow \infty} \frac{1}{t} \int_0^t I_{\{Y(u) \geq y\}} du$$

time to wait for the next bus

fracⁿ of time for which waiting time $\geq y$ [Total duration for which residual time $\geq y$ in $[0, t]$]



We can define rewards as $R_j = \int_{z_{j-1}}^{z_j} I_{\{Y(u) > y\}} du$

(during the x_j -th lifetime) (For how much time was the process Y , bigger than y)

$$M(t) = \sum_{j=1}^{\infty} R_j + \int_{z_{M(t)}}^t I_{\{Y(u) > y\}} du$$

\Rightarrow To show that $\{x_j, R_j\}_{j \geq 1}$ is a R.R.P., we know,

$$\int_{z_{j-1}}^{z_j} I_{\{Y(u) > y\}} du \stackrel{Y \equiv Y}{=} \int_0^{x_j} I_{\{Y(u-z_{j-1}) > y\}} du$$

$\therefore E[R_j] = \int_0^{x_j} I_{\{Y(u) > y\}} du \Rightarrow$ depends on x_j only

$$\Rightarrow \text{Here, } R_j = \begin{cases} 0 & \text{if } x_j \leq y \\ x_j - y & \text{o/w} \end{cases} \Rightarrow (R_j \text{ depends ONLY on } x_j)$$

$$\text{So, } E[R_j] = E[R_2] = E[R_2 | x_j \leq y] P(x_j \leq y) + E[R_2 | x_j > y] P(x_j > y)$$

so,

$$E[R_2] = E[x_j - y] P(x_j > y)$$

$$E[R_2] = \left(\frac{1-y}{u} \right) F^c(y)$$

$$\text{So, } \lim_{t \rightarrow \infty} \frac{1}{t} \int_0^t I_{\{Y(u) > y\}} du = F^c(y) - u y F^c(y)$$

$$= (1-u y) F^c(y) \quad \text{w.p. 1}$$

$$E[x_j - y | x_j \geq y] \cdot P(x_j \geq y)$$

$$= \int_0^\infty (x-y) \cdot dF_2(x)$$

∴ $\lim_{t \rightarrow \infty} \frac{1}{t} \int_0^t I_{\{Y(u) \geq y\}} du = \int_0^\infty (x-y) dF_2(x)$

eg. 2) An observer visits a renewal process with lifetimes $\{x_j\}_{j \geq 1}$ at a random time in interval $[0, t]$, find the probability that the residual time he/she observes is $\leq y$, for some $y \geq 0$ and let T be the time of visit.

i.e.

$$\int_0^t P(Y(u) \leq y) \cdot \frac{1}{t} du$$

⇒ Let T : the time of visit of the observer.

→ we need to find $P(Y(T) \leq y)$:

$$P(Y(T) \leq y | T=u) \cdot f_T(u) du$$

uniform R.V.
in $[0, t]$

$$\int_0^t P(Y(u) \leq y) \cdot \frac{1}{t} du$$

$$= \int_0^t P(Y(u) \leq y) \cdot \frac{1}{t} du$$

So, we need to evaluate,

$$\lim_{t \rightarrow \infty} \int_0^t P(Y(u) \leq y) \cdot \frac{1}{t} du$$

* Renewal Reward Theorem

$$E[I_A] = P(A)$$

$$P(Y(u) \leq y) = E[I_{\{Y(u) \leq y\}}]$$

Consider for any time t ,

$$\frac{1}{t} \int_0^t P(Y(u) \leq y) du = \frac{1}{t} \int_0^t E[I_{\{Y(u) \leq y\}}} du \\ = E\left[\frac{1}{t} \int_0^t I_{\{Y(u) \leq y\}} du\right]$$

Now,

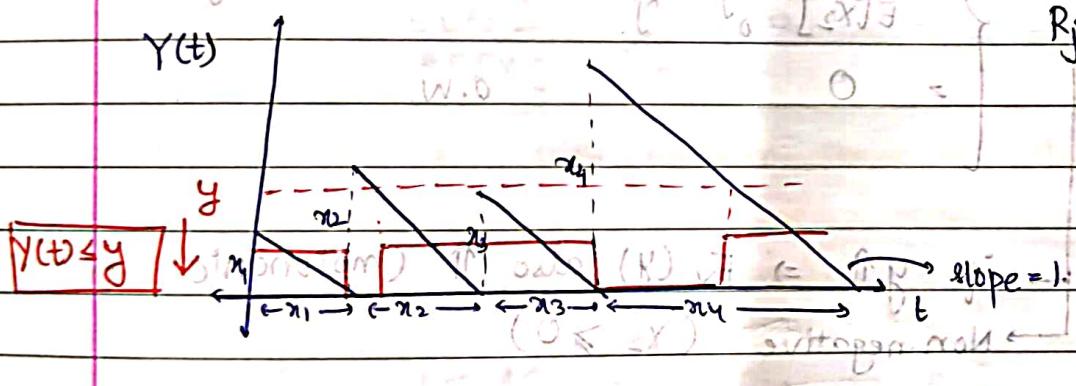
$$\lim_{t \rightarrow \infty} \frac{c(t)}{t} = \frac{E[R_2]}{E[X_2]}$$

i.e. $\lim_{t \rightarrow \infty} \frac{E[c(t)]}{t} = \frac{E[X_2]}{E[R_2]}$

Let, $c(t) = \int_0^t I_{\{Y(u) \leq y\}} du$

$$R_j = \int_{z_{j-1}(u)}^{z_j(u)} I_{\{Y(u) \leq y\}} du$$

$$R_j = \begin{cases} X_j & \text{if } X_j \leq y \\ y & \text{if } X_j > y \end{cases}$$



$$R_j = \min_{u \in [x_{j-1}, x_j]} \{X_j, y\}$$

$$E[R_j] = \int_0^t P(R_j > u) du$$

$$P(R_j > u) = P(\min\{X_j, y\} > u)$$

$$= P(X_j > u, y > u)$$

→ intersection of two events

certain (y, u are two constants)

null

$$= \begin{cases} 0 & \text{if } y \leq u \\ P(X_j > u) & \text{if } y > u \end{cases}$$

$f_{X_j}(u)$

$$E[R_j] = \int_0^y P(R_j > u) du + \int_y^\infty P(R_j > u) du.$$

y ∞

$u > y$

$E[R_j] = \int_0^y F_{x_j}^C(u) \cdot du$

Using the Renewal Reward Theorem,

$$\lim_{t \rightarrow \infty} \frac{1}{t} \int_0^t P(Y(u) \leq y) \cdot du = \frac{c(t)}{t} = \frac{E[R_2]}{E[X_2]}$$

i.e. $\boxed{\lim_{t \rightarrow \infty} \frac{1}{t} \int_0^t P(Y(u) \leq y) \cdot du = \frac{1}{E[X_2]} \int_0^y F_{x_2}^C(u) \cdot du}$

Let $F_Y(y) = \frac{1}{E[X_2]} \int_0^y F_{x_2}^C(u) \cdot du$ for $y \geq 0$

$= 0$ o.w.

if $y \uparrow \Rightarrow F_Y(y)$ also \uparrow (monotonic)
 Non-negative ($X_2 \geq 0$)

Also,

$$\lim_{y \rightarrow \infty} \frac{1}{E[X_2]} \int_0^y F_{x_2}^C(u) \cdot du = E[X_2] \text{ as } y \rightarrow \infty$$

Note :- if $P(X_2 > 0) = 1$, then,

$$E[X_2] = \int_0^\infty P(X_2 > u) \cdot du.$$

$$\lim_{y \rightarrow \infty} F_Y(y) = 1$$

eg. 3) Average waiting time for the observer, i.e.

$$\lim_{t \rightarrow \infty} \frac{1}{t} \int_0^t Y(u) \cdot du \quad (\text{expected: } E[X_2] / 2)$$

$\int_0^t Y(u) \cdot du$ → random variable ⇒ can be defined as $c(t)$ directly

$$c(t) = \int_0^t Y(u) \cdot du$$

$$R_j = \int_{z_{j-1}}^{z_j} Y(u) \cdot du$$

(Y at a time 'u')

(integral in a lifetime)

$$R_j = \frac{x_j \cdot 2}{2}$$

$$E[R_j] = \frac{1}{2} E[X_j^2]$$

$$\text{So, } \lim_{t \rightarrow \infty} \frac{1}{t} \int_0^t Y(u) \cdot du = \frac{E[X_2]}{2 E[X_2]} \quad \text{w.p. 1}$$

$$E[X_2^2] > E[X_2]^2$$

avg. waiting time, with random times is always perceived to be higher than one with certain times.

→ **BATCH BIASING**

DTMC as a Renewal Process

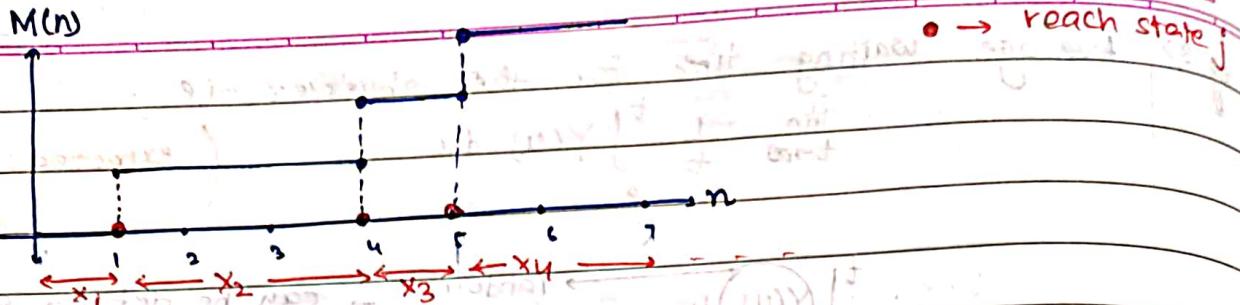
⇒ $\{P_n\}_{n \geq 0}$ defined by $\bar{\alpha}$ and P_{0j} 's

⇒ Take any state $j \in S$, define

CLAIM :- $\{M(n)\}_{n \geq 1}$ is a renewal process

$$M(n) = \sum_{k=1}^n \mathbb{I}\{X_k = j\}$$

(no. of visits to state j until n-steps)



$M(n)$ is a counting process

1) All the lifetimes (x_j 's) are **INDEPENDENT** because of the **STRONG MARKOV PROPERTY**

2) So, x_j 's, $j \geq 2$ i.e. $\{x_j\}_{j \geq 2}$ are identically distributed.

\Rightarrow (first visit to state j , time, T is a stopping time)

Distribution of x_2

~~Recurrence~~

$$P(X_n = k) = f_{ij}^{(R)}$$

First Passage Time Distribution

$$P(X_n = k) = P(D_1 \neq j, D_2 \neq j, \dots, D_k = j \mid D_0 = j) ; n \geq 2$$

and

$$P(X_1 = k) = \sum_{i \in S} \alpha_i f_{ij}^{(R)}$$

probability of visiting the j^{th} -state for the first time

$\rightarrow X_2$: the steps we take from the 1^{st} visit to the 2^{nd} visit

SUPPOSITIONS OF ERT

$$\textcircled{1} \quad P(X_1 < \infty) = 1$$

$$\textcircled{2} \quad E[X_2] \in [0, \infty]$$

Propositions of ERT

$$\textcircled{1} \quad \lim_{n \rightarrow \infty} \frac{M(n)}{n} = \frac{1}{E[X_2]} \text{ w.p. 1}$$

$$\textcircled{2} \quad \lim_{n \rightarrow \infty} \frac{m(n)}{n} = \frac{1}{E[X_2]}$$

$$P(X_1 < \infty) = \sum_{i \in S} \alpha_i f_{ij}^{(R)}$$

Suppose DTMC is **irreducible** and **recurrent**

irreducible, C.C. has recurrent states $\Rightarrow f_{ij} = f_{ji} = 1$

and so, $P(X_1 < \infty) = 1$

$X_2 \geq 1$ i.e. $E[X_2] \geq 1$ and also the other cond holds
 $\Rightarrow E[X_2] \in [1, \infty]$

so, irreducible & recurrent DTMC's

Thm:- If $\{D_n\}_{n \geq 0}$ is an irreducible, positive recurrent DTMC with a stationary measure π . then,

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=1}^n p_{ij}^{(k)} = \pi_j = \frac{1}{v_{jj}} \quad \forall j \in S$$

PROOF:- $p_{ij}^{(k)} = P(D_k=j | D_0=i)$

Define $\sum \{D_k=j | D_0=i\}$

$$M(n) = \sum_{k=1}^n \sum \{D_k=j | D_0=i\}$$

$$E[M(n)] = \sum P(D_k=j | D_0=i) \\ = \sum p_{ij}^{(k)}$$

visits to state-j in first n-steps starting from state-i.

To show $\{M(n)\}_{n \geq 1}$ is a renewal process

$$P(X_1 < \infty) = f_{ij} = 1$$

$$E[X_2] \in [0, \infty]$$

so,

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=1}^n \sum \{D_k=j | D_0=i\} = \frac{1}{v_{jj}} \quad \text{w.p. 1}$$

Also,

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=1}^n p_{ij}^{(k)} = \frac{1}{v_{jj}} \quad \left(\lim_{n \rightarrow \infty} \frac{M(n)}{n} = E[M(n)] \right)$$

Thm:- If $\{D_n\}_{n \geq 0}$ is an irreducible, positive recurrent DTMC with a stationary measure π . Then, find expected no. of visits to state i b/w two consecutive visits to state j. $E[V_{ji}]$

Consider you earn a reward 1 every time state i is visited.

Total reward collected in the n^{th} lifetime

$$\sum_{k=1}^{Z_n} I_{\{D_k=i\}} = R_n$$

$c(n) \Rightarrow \# \text{ of times state } i \text{ is visited}$

$$\lim_{n \rightarrow \infty} \frac{c(n)}{n} = \frac{E[V_{ji}]}{E[X_2]} \text{ w.p. 1}$$

total no. of visits to state i until ' n '.

LHS is same

\Rightarrow consider the R.R.P. $\{M_i(n)\}$

$$\lim_{n \rightarrow \infty} \frac{M_i(n)}{n} = \frac{1}{\pi_{ii}} \text{ w.p. 1}$$

Finally,

$$E[V_{ji}] = \frac{\pi_i}{\pi_j}$$

11/03

RECAP :-

Renewal Process

(ERT)

RRT

Renewal Reward Process (RRP)

Time-avg. distribⁿ of residual time

Time avg. of residual - batch biasing

Applications to DTMC :-

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{j=1}^n p_{ij} c(j) = \pi_j + i \in S$$

and

$$E[V_{ji}] = \frac{\pi_i}{\pi_j}$$

POISSON'S PROCESS

A poisson's process, $\{N(t)\}_{t \geq 0}$ is a COUNTING PROCESS with the following properties :-

① All jumps of $\{N(t)\}$ are of unit size w.p. 1.

$\Rightarrow N(t+s) - N(t)$ \Rightarrow # of jumps in interval $(t, t+s]$
increments of $\{N(t)\}_{t \geq 0}$ in $(t, t+s]$

(no. of jumps - 1)
 $N(t+s) - 1$

(2) $\{N(t)\}_{t \geq 0}$ has independent increments in NON-OVERLAPPING time intervals.

e.g. consider two time intervals $(t_1, t_1+s_1]$ and $(t_2, t_2+s_2]$, then,
 $(N(t_2+s_2) - N(t_2)) \perp\!\!\!\perp (N(t_1+s_1) - N(t_1))$ as long as
 $(t_1, t_1+s_1] \cap (t_2, t_2+s_2] = \emptyset$ (are disjoint)

(3) $\{N(t)\}_{t \geq 0}$ has stationary increments

Distribution of increments in I is same as no. of increments in interval $I+t \forall t \geq 0$.

$$I = [t_1, t_2]$$

$$I+t = [t_1+t, t_2+t]$$

distrib' of no. of intervals remains the same w.r.t. time

⇒ Using the above 3, we can derive other properties of the Poisson's process :-

Lemma 1:- If $0 \leq t_1 < t_2 < t_3 \dots < t_n < \infty$ and $0 \leq i_1 \leq i_2 \leq \dots \leq i_n$ (whole numbers)

$$\left(N(t_1) = i_1, N(t_2) = i_2, \dots, N(t_n) = i_n \right) = \prod_{k=1}^n P(N(t_k) - N(t_{k-1}))$$

PROOF:- LHS $\Rightarrow (N(t_1) = i_1, N(t_2) - N(t_1) = i_2 - i_1, \dots, N(t_n) - N(t_{n-1}) = i_n - i_{n-1})$

By the property of independent increments,

$$\Rightarrow \prod_{k=1}^n P(N(t_k) - N(t_{k-1})) = i_k - i_{k-1}$$

Lemma 2:- $\exists \lambda > 0$ s.t. $\forall t \geq 0, P(N(t) = 0) = e^{-\lambda}$

PROOF:- Define $G(t) = P(N(t) = 0)$

$$G(t+s) = P(N(t+s) = 0)$$

$$\Rightarrow P(N(t) = 0, N(t+s) - N(t) = 0) = \prod_{j=t}^{t+s-1} P(N(t+j) = 0)$$

By independent increment property.

$$G(t+s) = P(N(t)=0) \cdot P(N(s)=0)$$

↳ stationary increment property

so, $G(t+s) = G(t) \cdot G(s)$

(so, $G(t) = e^{\alpha t}$)

(α can be >0 or <0)

since it is a probability

as $G(t) \in [0,1] \Rightarrow \alpha \in [-\infty, 0]$

→ $G(t) = e^{-\lambda t}$ for some $\lambda \geq 0$.

$P(N(t)=0) = e^{-\lambda t}$ for some $\lambda \geq 0$

Lemma 3 :- $\lim_{t \downarrow 0} \frac{1}{t} P(N(t) \geq 2) = 0$

proved using property - ①

Lemma 4 :- $\lim_{t \downarrow 0} \frac{1}{t} P(N(t)=1) = \lambda$ (probability of $P(N(t+\tau) - N(t) = 1)$ for a small enough τ)

if the interval is very small, the $P(N(t)=1)$ increases at

PROOF :- $P(N(t)=1) = 1 - P(N(t)=0) - P(N(t) \geq 2)$

$$\lim_{t \downarrow 0} \frac{1}{t} (P(N(t)=1)) = \lim_{t \downarrow 0} \left[\frac{1 - P(N(t)=0)}{t} \right] - \lim_{t \downarrow 0} \frac{1}{t} P(N(t) \geq 2)$$

$$= \lim_{t \downarrow 0} \frac{(1 - e^{-\lambda t})}{t} = 0$$

$\lim_{t \downarrow 0} \frac{1}{t} (P(N(t)=1)) = \lambda$

THEOREM :-

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

Poisson PMF with parameter (λt)

PROOF :-

Define $E[\sum_{i=1}^{N(t)} (C_i(t))^k]$

$$= \sum_{R=0}^{\infty} R^k \cdot P(N(t)=R)$$

Note that :- $G_1(t+s) = E[z^{N(t+s)}]$

$$= E[z^{(N(t+s)-N(t)+N(t))}]$$

$$= E[z^{(N(t+s)-N(t))}] \quad (\text{stationary-increment property})$$

$$= E[z^{N(s)}] \cdot E[z^{N(t)}]$$

$$= E[z^{N(s)}] \cdot E[z^{N(t)}]$$

$$= E[z^{N(s)}] \cdot E[z^{N(t)}]$$

$$= G_1(s) G_1(t)$$

So,

$$G_1(t) = e^{g(z)t}$$

$$g(z) = \lim_{t \rightarrow 0} \frac{e^{g(z)t} - 1}{t}$$

i.e.



$$g(z) = \lim_{t \rightarrow 0} \frac{1}{t} [e^{g(z)t} - 1]$$

$$\Rightarrow \lim_{t \rightarrow 0} \frac{1}{t} [G_1(t) - 1]$$

$$= \lim_{t \rightarrow 0} \frac{1}{t} \left[\sum_{k=0}^{\infty} z^k \cdot P(N(t)=k) - 1 \right]$$

$$= \lim_{t \rightarrow 0} \frac{1}{t} \left[P(N(t)=0) + z \cdot P(N(t)=1) + \sum_{k=2}^{\infty} z^k P(N(t)=k) - 1 \right]$$

$$= \lim_{t \rightarrow 0} \left[\frac{e^{-\lambda t} - 1}{t} + z \frac{1}{t} P(N(t)=1) + \frac{1}{t} \sum_{k=2}^{\infty} z^k P(N(t)=k) \right]$$

$$- \lambda$$

$$g(z)$$

$$0$$

$$[g(z) = -\lambda + \lambda z = (1+z)\lambda]$$

So, we get,

$$G_1(t) = e^{(-\lambda + \lambda z)t}$$

$$= (e^{-\lambda t}) \cdot \sum_{k=0}^{\infty} \frac{(\lambda z t)^k}{k!}$$

$$\text{So, } G_1(t) = e^{-\lambda t} \sum_{k=0}^{\infty} z^k P(N(t)=k) = K!$$

comparing with $G_1(t) = \sum_{k=0}^{\infty} z^k P(N(t)=k)$,

$$G_1(t) = \sum_{k=0}^{\infty} z^k P(N(t)=k)$$

we get, $P(N(t)=k) = e^{-\lambda t} \frac{(\lambda t)^k}{k!}$

Hence Proved!!

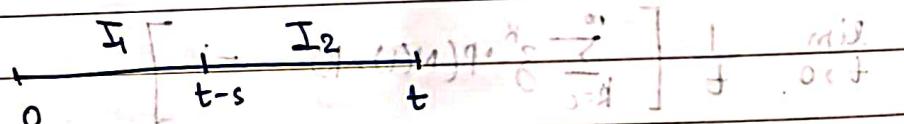
$\{N(t)=0\} \Rightarrow \text{no jump till time } t'$

\Rightarrow let x denote the time for the first jump, then,

$$\{N(t)=0\} = \{x > t\}$$

$$\Rightarrow P(x > t) = e^{-\lambda t}$$

$$\Rightarrow x \sim \exp(\lambda) \text{ i.e. } f_x(x) = \lambda e^{-\lambda x} + n > 0$$



$$N(t-s) = 0 \text{ and } N(t) - N(t-s) = 0$$

\Rightarrow given an info. that say $x > (t-s)$, it gives us NO info. about whatever happens in I_1 , i.e. if I_2 .

Memoryless Property

$$P(x > t+s | x > s) = P(x > t)$$

So, the time until the first jump is exponential

(given the time elapsed, we have no info. about how much later will a jump actually occur)

general :- $P(x > t+s, x > s) = P(x > t) P(x > s)$ (Knowing that I've spent s time in I_1)
 $P(x > t+s) = P(x > t) P(x > s)$. (I spend $(t+s)$ more time)

i.e. $G_1(t+s) = G_1(t) G_1(s) \rightarrow \text{exponential}$

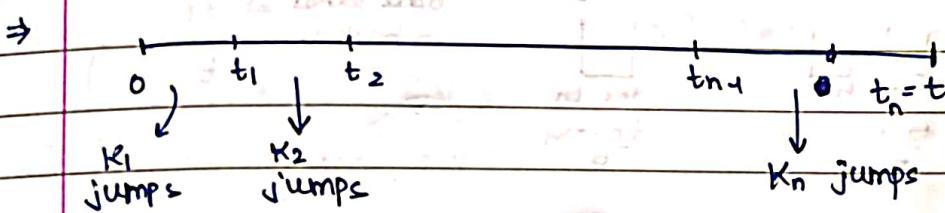
Problem:- Show that :-

Let $N(t)$ be a poisson process with rate $\lambda > 0$, let $0 = t_0 < t_1 < t_2 < \dots < t_n = t$

Then, $\forall K, K_1, \dots, K_n$ s.t. $\sum_{i=1}^n K_i = K$, ~~Pr{N(t) = K}~~

$$P(N(t_1) = K_1, N(t_2) = K_1 + K_2, \dots, N(t_n) = K \mid N(t) = K)$$

$$= \frac{K!}{\prod_{i=1}^n (K_i)!} \prod_{i=1}^n \left(\frac{t_i - t_{i-1}}{t} \right)^{K_i}$$



$$P(N(t_1) = K_1, N(t_2) = K_1 + K_2, \dots, N(t_n) = K \mid N(t) = K)$$

$$\Rightarrow P(N(t_1) = K_1, N(t_2) - N(t_1) = K_2, \dots, N(t_n) - N(t_{n-1}) = K_n \mid N(t) = K)$$

$$\Rightarrow \frac{P(N(t_0) = K_1, \dots, N(t_n) - N(t_{n-1}) = K_n, N(t) = K)}{P(N(t) = K)}$$

if we pick any 'n'

events from here, the $(n+1)^{th}$ HAS TO OCCUR.

$$\Rightarrow \frac{\prod_{i=1}^n P(N(t_i) - N(t_{i-1}) = K_i)}{P(N(t) = K)}$$

$$= \frac{\prod_{i=1}^n P(N(t_i - t_{i-1}) = K_i)}{P(N(t) = K)} = \frac{\prod_{i=1}^n e^{-\lambda(t_i - t_{i-1})} (\lambda(t_i - t_{i-1}))^{K_i}}{K_i!} \cdot \frac{e^{-\lambda t} (\lambda t)^K}{K!}$$

$$= (K!)^{-1} e^{-\lambda((t_1 - t_0) + (t_2 - t_1) + \dots + (t - t_{n-1}))} \cdot \frac{e^{-\lambda t} \lambda^K t^K}{K!} = \frac{(K!)^{-1} e^{-\lambda(t_1 - t_0) - \lambda(t_2 - t_1) - \dots - \lambda(t - t_{n-1})}}{\prod_{i=1}^n (t_i - t_{i-1})^{K_i}}$$

$$= \boxed{\frac{K!}{\prod_{i=1}^n K_i!} \prod_{i=1}^n \left(\frac{t_i - t_{i-1}}{t} \right)^{K_i}} \quad \left(\text{as } \prod_{i=1}^n t^{K_i} = t^K \right)$$

H.P.

→ Drop k points on the interval $[0, t]$. Let X_k denote the position of the k^{th} -point, $\{X_k\}$ are i.i.d. Uniform $([0, t])$

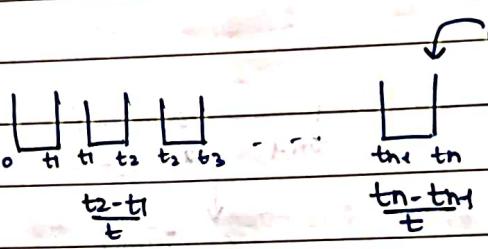
$$f_{X_k}(x) = \frac{1}{t} \quad \forall x \in [0, t]$$

$$= 0 \quad \text{otherwise}$$

→ Let $Z(t, t+s) = \# \text{ of points in interval } (t, t+s]$

⇒ Find $P(Z(0, t_1) = k_1, Z(t_1, t_2) = k_2, \dots, Z(t_n, t_{n+1}) = k_n)$

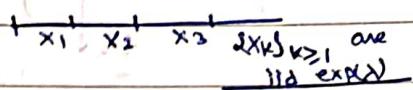
(Analogy :-



EQUIVALENT DEFINITIONS FOR A POISSON PROCESS

Thm:-

A counting process $\{N(t)\}_{t \geq 0}$ is a poisson process with parameter λ , if and only if, successive jump times, $\{T_k\}_{k \geq 1}$ are renewal instances with i.i.d. exp.(λ) life times.



⇒ consider a renewal process, $\{N(t)\}_{t \geq 0}$ with lifetimes $\{X_n\}_{n \geq 1}$.

If X_n 's are iid exp.(λ), then, $\{N(t)\}$ is a poisson process with a rate, λ .

⇒ A point process $\{N(t)\}_{t \geq 0}$ is a Poisson process iff

(1) $t_0 < t_1 < \dots < t_n$; $\{N(t_k) - N(t_{k-1})\}_{k=1,2,\dots,n}$ are independent R.V.'s

(2) $\exists \lambda \geq 0$ s.t. $P((N(t) - N(t-s)) = k) = \frac{e^{-\lambda s} (\lambda s)^k}{k!}$ $k=0,1,2,\dots$

(from 17th)

REGENERATIVE PROCESSES

Defn: A process $\{X(t)\}_{t \geq 0}$ is called a regenerative process if \exists a stopping time, T_1 , such that

(1) $\{X(t+T_1)\}_{t \geq 0}$ and $\{X(t)\}_{t \geq 0}$ have same probability law

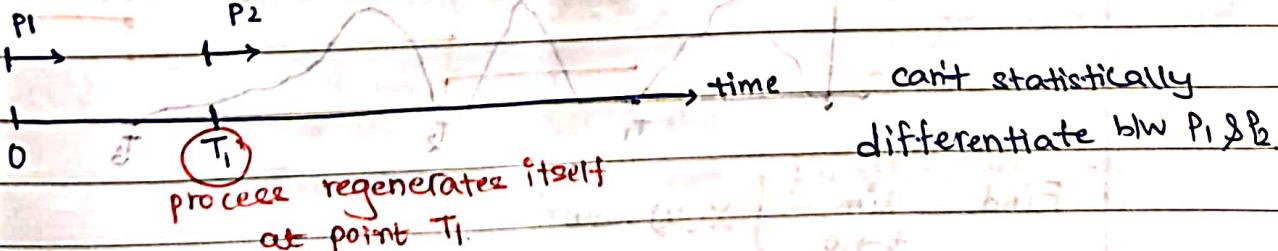
(2) $\{X(t+T_1)\}_{t \geq 0} \perp \{T_1, X(u), u < T_1\}$

→ $\{X(t)\}$ can take any real value.

⇒ T_1 is a stopping time $\Rightarrow \{T_1 \geq t\}$ depends only on $\{X(u)\}_{u < t}$

(continue or not determined by the process before t ONLY.)

Property-1 :-



can't statistically differentiate b/w P1 & P2

Property-2 :- Process after T_1 is independent of T_1 , (whatever value T_1 took)

T_1 won't change how process behaves after T_1

→ also independent of the past.



Poisson process is a \rightarrow process with difference of $(N(t) - N(t_s))$ special type poisson, and all the epoch instant are also regenerative stopping times.

Lemma:- For a regenerative process, $\exists T_2 > T_1$, a stopping time satisfying properties ① & ②.

Corollary:- $\{T_0, T_1, T_2 - T_1, \dots\}$ are i.i.d random variables.

T_2 plays the same role as that of T_1 for $\{X(t+T_1)\}_{t \geq 0}$

Corollary:- Define :-

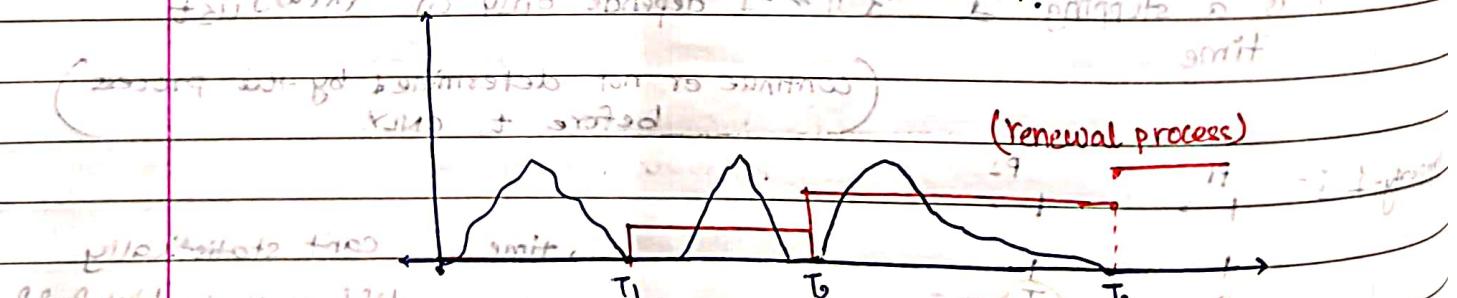
$$X_n = T_n - T_{n-1} \quad (T_0 = 0)$$

$\{X_n\}_{n \geq 0}$ are iid



regenerative process has a renewal embedded

in it with lifetimes as $\{X_n\}_{n \geq 0}$.



Find $\lim_{t \rightarrow \infty} \int_0^t X(u) \cdot du$

(It is an overshoot). If $t \rightarrow \infty$,
Renewal Reward
Theorem

$$R(U) = X(U) \rightarrow \text{reward}$$

longer with a transition rule

using RRT, it is

$$E \left[\int_0^{T_i} X(u) du \right] / E[T_i]$$

Note :-

renewal process \rightarrow i.i.d. $\exp(\lambda)$ lifetimes



$$A = F$$

$$A \neq F$$

Regenerative process :- an embedded ordinary renewal process and the renewal epochs are $\{T_1, T_2, \dots\}$

ERT
RRT \rightarrow limiting-time average behavior

POINT-WISE LIMITS

$$\text{eg. } a_n = \begin{cases} 1 & \text{if } n \text{ is odd} \\ 0 & \text{o/w} \end{cases}$$

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum a_k = \frac{1}{2}$$

$$\text{but point-wise, } \lim_{n \rightarrow \infty} a_n \rightarrow \text{DNE}$$

$\{X_n\} \sim \text{Bernoulli}(p)$ [i.i.d.]

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=1}^n X_k = p \text{ w.p. 1}$$

$\lim_{n \rightarrow \infty} X_n$ (doesn't exist with prob = 1)

(Borel-Cantelli lemma)

\Rightarrow If point-wise limit exists, the time-average one should also exist.

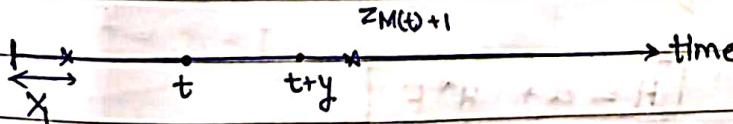
Renewal Equation (Ordinary Renewal Processes)

$P(Y(t) > y)$: If an observer arrives

at time t , what is the probability that it sees a waiting time $>y$.

The time avg. limit,

$$\frac{1}{t} \int_0^t P(Y(u) > y) du \text{ exists}$$



$$Y(t) = ZM(t)+1 - t$$

waiting time

The renewal after t occurs only after $t+y$

⇒ Condition on the first renewal.

$$= P(Y(t) > y \mid X_1 < t) + P(Y(t) > y \mid X_1 > t+y)$$

$$F^c(t+y)$$

(always true)

intersect
with E is
@ C itself.

C is a subset of E $C \Rightarrow E(P(Y(t) > y))$

$t = A$

waiting time is less than y

$$P(Y(t) > y, X_1 < t) = \int_0^t P(Y(t) > y \mid X_1 = u) \cdot dF(u)$$

process after X_1 is independent
of value of X_1 , and so, once

$Y(t-u)$ is being put in, P
won't depend on X_1 anymore

$$= \int_0^t P(Y(t-u) > y) dF(u)$$

$$= \int_0^t G(t-u) \cdot dF(u)$$

$$\text{Let } G(t) = P(Y(t) > y)$$

$$\text{so, } P(Y(t) > y, X_1 < t) = \int_0^t G(t-u) dF(u)$$

$$G * F(t)$$

So, we can write

$$G(t) = G * F(t) + F^c(t)$$

So, any eqn of the form

$$H(t) = a(t) + \int_0^t H(t-x) dF(x)$$

where 'a' is any given func and F is a distrib' fnc is
called a renewal eqn.

consequently :

$$H = a + H * F$$

Thm:- If distribution function, F , has a finite mean and function ' a ' is a bounded $f^n C$, then the unique solⁿ

$$H = a + H * F \quad \text{and} \quad m(t) = (t)H$$

that is bounded on finite intervals, i.e,

$$H(t) = a(t) + \int a(t-u) d m(u), \quad m(t) = \sum_{k=1}^{\infty} F^{(k)}(t)$$

Given \leftarrow finite mean, i.e., $\int_0^\infty n \cdot dF(x) < \infty$ or $\int_0^\infty F^n(x) dx < \infty$

(expectation)

Bounded $\rightarrow \exists / \sup_{t \in \mathbb{R}} |a(t)| < \infty$

Proof:- The ∞ -sum, $m(t) = \sum_{k=1}^{\infty} F^{(k)}(t)$ is finite (proved earlier)

$$(t)H \leq (t)m + (t)$$

as it is an ordinary renewal process.

$$P(M(t) \geq n) = P(Z_k \leq t) = F^{(n)}(t)$$

$$H = a + H * F$$

$$H = a + (a + a * m) * F$$

$$H = a + a * F + a * m * F$$

$$H = a + a * (F + m * F)$$

$$m * F = \sum_{k=2}^{\infty} F^{(k)}$$

$$(B + F + m * F) = B + \sum_{k=1}^{\infty} F^{(k)} = m(t)$$

so, $H = (a + t)H$ is indeed a solution.

$$(B + t)A = (B + t)A =$$

Now, we show that H is bounded for every $T < \infty$

$$H(t) = a(t) + \int_0^t a(t-u) dm(u)$$

$$\begin{aligned} |H(t)| &\leq |a(t)| + \left| \int_0^t a(t-u) dm(u) \right| \\ &\leq |a(t)| + \int_0^t |a(t-u)| dm(u) \end{aligned}$$

$$\sup_{0 \leq t \leq T} |H(t)| \leq \sup_{0 \leq t \leq T} \left[|a(t)| + \int_0^t |a(t-u)| dm(u) \right]$$

$$\leq \sup |a(t)| + \int_0^t \sup |a(t)| b dm(u)$$

(using max start at t) $\sup |a(t-w)| = \sup |a(t)|$ and $\sup(A) + \sup(B) \geq \sup(A+B)$

$$\leq \sup |a(t)| + \sup |a(t)| \int_0^t dm(u)$$

$$= \sup |a(t)| + \sup |a(t)| \cdot m(t)$$

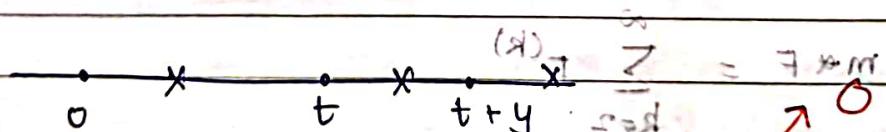
As a is bounded and $m(t)$ is finite,

$\sup |H(t)| < \infty \Rightarrow H(t)$ is bounded.

Uniqueness can be proved by contradiction.

DELAYED RENEWAL PROCESS

Let $K(t) = P(Y(t) > y)$



$$K(t) = P(Y(t) > y, X_1 \leq t) + P(Y(t) > y, t \leq X_1 \leq t+y)$$

$$+ P(Y(t) > y, X_1 \geq t+y)$$

$$= P(X_1 > t+y) = \boxed{A^c(t+y)}$$

$$P(Y(t) > y, X_1 \leq t) = \int_0^t K_0(t-x) dA(n)$$

(similar to the ordinary case)

where $K_0(t-n) = P(Y_0(t-n) > y)$ where Y_0 is the residual time of the ordinary renewal process.

Now,

$$K(t) = A^c(t+y) + (K_0 * A)(t) = dt$$

renewal eq is of the form

$$H = a + H_0 * A$$

$$H_0 = a_0 + H_0 * F$$

Theorem: The solution of the general renewal eq is $H = a + a_0 * m$, where,

$$m = \sum_{k=1}^{\infty} A * F^{(R-k)}(t)$$

$$H = a + a_0 * m$$

(renewal eq for a delayed renewal process)

Proof: we have shown that the soln for $H_0 = a_0 + H_0 * F$ is $H_0 = a_0 + a_0 * m_0$ (ordinary RP)

$$H = a + H_0 * A$$

$$H = a + (a_0 + a_0 * m_0) * A = a + a_0 * A + a_0 * m_0 * A$$

$$H = a + a_0 * (A + m_0 * A)$$

$$= a + a_0 * (A + m_0 * A)$$

↓

$$\left(\sum F^{(R)}(t) \right) * A$$

(a_0, m_0 : ordinary RP)

$$m_0 = \sum F^{(R)}(t)$$

$$= a + a_0 * \left(A + \sum_{R=1}^{\infty} A * F^{(R)}(t) \right)$$

$$= a + a_0 * \left(\sum_{R=1}^{\infty} A * F^{(R)}(t) \right)$$

$$H = a + a_0 * m_0$$

a after sum $m_0 = x$

and m_0 and x

APPLICATIONS :-

Poisson process :- Can be considered as an ordinary renewal process, with lifetime $\sim \exp(\lambda)$

$$P(Y(t) > y) = F^c(t+y) + \int_0^t F^c(t+y-n) dm(x)$$

$$\text{as } H_0 = a_0 + a_0 + m_0 + a_0(t) = AF^c(t+y)$$

$$F^c(t+y) = e^{-\lambda(t+y)}$$

$$F^c(t+y-n) = e^{-\lambda(t+y-n)}$$

$m \star D + D \cdot M(t)$ is a poisson process with parameter λt , if

$$m(t) = E[M(t)] = \lambda t$$

$$dm(x) = \lambda dx \quad A \sum_{i=1}^t = \lambda t$$

$$\text{Hence, } P(Y(t) > y) = e^{-\lambda(t+y)} + \int_0^t e^{-\lambda(t+y-n)} \cdot \lambda dx$$

$$= e^{-\lambda(t+y)} + \left[e^{-\lambda y} - e^{-\lambda(t+y)} \right]$$

$$(99 \text{ min}) \quad \lambda t \star \lambda D + \lambda D = \lambda t + \lambda D = \lambda t$$

$$= e^{-\lambda y}$$

$$= P(X > y) \quad \text{where } X \text{ is a lifetime.}$$

$$A \star \lambda t \star \lambda D + A \star \lambda D + D = A \star (\lambda t + \lambda D) + D = A$$

Thus, $Y(t)$ has the same distribution as lifetime, i.e., distribution of $Y(t)$ after time t is same as that of lifetime starting from zero.

i.e. from anywhere, the process looks the same and so it is a stationary process.

$$((t) \star A \star + A) \star \lambda D + D =$$

\Rightarrow stationary process. :- A process in which from no matter where we start observing, the process remains the same.

say we start from t , $Y(t) \Rightarrow$ renewal time after t

$x_1 \Rightarrow$ renewal time after 0

They have the same distribution.

Note:- In general, the property of independent increments need not be true for a process to be stationary.

consider

$$\lim_{t \rightarrow \infty} \frac{1}{t} \int_0^t P(Y(u) \leq y) du = \frac{1}{E[X_2]} \int_0^y F^c(x) dx \text{ w.p. 1}$$

Define $F_c(y) = \frac{1}{E[X_2]} \int_0^y F^c(x) dx$

$$P(Y(t+u) \leq y) = \int_0^y \frac{1}{E[X_2]} F^c(x) dx$$

$$(1) - (\text{r.m.b.}) (x-p+s)^2 \int_0^y + (B+s)^2 \int_0^y = (B^2(t)) \int_0^y$$

$$(2) ((\text{r.m.b.})^2 A) \int_0^y = (f(t))^2 A$$

Probability measure

$$((\text{r.m.b.}))^2 A \int_0^y = (r.m.b.)^2 A$$

$$((\text{r.m.b.}))^2 A \int_0^y = (r.m.b.)^2 A$$

$$(r.m.b.)^2 = -1$$

$$x b(r) \int_0^y \frac{1}{E[X_2]} = (f(t)) A$$

$$((r.m.b.)) \frac{1}{E[X_2]^2} - ((f(t)))^2 \frac{1}{E[X_2]^2} = (r.m.b.)^2$$

$$(r.m.b.)^2 = \frac{1}{[x]^{E[2]}} = \frac{(r.m.b.)^2 A}{(r.m.b. - 1)}$$

Examples :-

① Poisson process (λ) :-

$$P(Y(t) > y) = e^{-\lambda y} \quad \forall y \geq 0, \forall t \in \mathbb{R}$$

Alternatively, $Y(t) \sim F_{\exp(\lambda)} \quad \forall t \in \mathbb{R}$

$$\textcircled{2} \quad A(y) = F_e(y) = \frac{1}{E[X_2]} \int_0^y F^c(x) dx \quad \forall y \geq 0$$

$$\hookrightarrow \lim_{t \rightarrow \infty} \frac{1}{t} \int_0^t P(Y(t) \leq y) dy$$

② Delayed renewal process

$$X \sim A = F_e$$

$$P(Y(t) > y) = F_e^c(t+y) + \int_0^t F^c(t+y-x) dm(x) \quad \text{--- (1)}$$

$$m(t) = \sum_{k=1}^{\infty} (A^* F^{(k)}) (t)$$

\downarrow Laplace transform

$$\tilde{m}(s) = \sum_{k=1}^{\infty} \tilde{A}(s) (\tilde{F}(s))^k$$

$$\tilde{m}(s) = \tilde{A}(s) \sum_{k=0}^{\infty} (\tilde{F}(s))^k$$

$\tilde{m}(s) =$	$\frac{\tilde{A}(s)}{1 - \tilde{F}(s)}$
------------------	---

$$A(t) = \frac{1}{E[X_2]} \int_0^t F^{(k)}(x) dx$$

$$\tilde{A}(s) = \frac{1}{E[X_2]} \cdot \frac{1}{s} \ln(1 - F(s)) = \frac{1}{s E[X_2]} (1 - \tilde{F}(s))$$

$$\frac{\tilde{A}(s)}{1 - \tilde{F}(s)} = \frac{1}{s E[X_2]} = \tilde{m}(s)$$

Taking the L.L.T.,

$$m(t) = ut$$

using ①,

$$\begin{aligned}
 P(Y(t) > y) &= F^c(t+y) + \int_0^t F^c(t+y-x) dx \\
 &= F^c(t+y) + \frac{1}{E[X_2]} \int_0^t F^c(t+y-u) du \\
 &= F^c(t+y) + \frac{1}{E[X_2]} \int_y^t F^c(u) du \\
 &= F^c(t+y) + \frac{1}{E[X_2]} \int_0^y F^c(u) du - \frac{1}{E[X_2]} \int_0^y F^c(u) du \\
 &= F^c(t+y) + F^c(y) - F^c(y) \\
 &= F^c(y)
 \end{aligned}$$

For large t ,

from ERT

$$\lim_{t \rightarrow \infty} \frac{m(t)}{t} = u$$

$$m(t+h) - m(t) = [u(t+h) + o(t+h)] - [u(t) + o(t)]$$

$$= uh + [o(t+h) - o(t)]$$

$\rightarrow 0$ assuming $h \ll t$

so, intuitively,

$$\lim_{t \rightarrow \infty} m(t+h) - m(t) = uh$$

the expected rewards are deterministic and ↑ at a rate u .

$dm(x) \rightarrow$ prob. that there is a renewal in $[x, x+dx]$

$$m(x) = \sum_{k=1}^{\infty} (A^* F^{(R-1)})(x)$$

reduces $(x-p)^+$ to $(x-p+k)^+$ $\Rightarrow P(Z_k \leq x)$

$dm(x) \rightarrow$ density of $Z_k \cdot x dx$

① Poisson Process :- $m(x) = \lambda x$

$$dm(x) = \lambda dx = \frac{1}{E[X_2]} dx$$

→ delayed poisson process with $A=F_p$:-

$$\text{again, } dm(x) = \lambda dz = \frac{dx}{E[X_2]}$$

(Time-averaged excess time distribution)

$$\text{ERT: } (\lim_{t \rightarrow \infty} \frac{m(t)}{t}) + \frac{1}{E[X_2]} =$$

for large enough t ,

$$m(t) \approx t$$

$$dm(t) = \frac{dt}{E[X_2]}$$

Def :- A proper variable X is called lattice if $\exists d > 0$ s.t.

$$(a) \sum_{n=-\infty}^{+\infty} P(X=nd) = 1 \quad \text{o/w } X \text{ is called non-lattice}$$

→ largest 'd' for which (a) holds is called the span of X .

Blackwell's Renewal Theorem (CBRT)

→ For a delayed renewal process with arbitrary A and non-lattice F , and any $h > 0$

$$\lim_{t \rightarrow \infty} [m(t+h) - m(t)] = h\mu$$

→ For an ordinary renewal process with lattice F having a span d for $h=d, 2d, \dots$

$$\lim_{t \rightarrow \infty} [m(t+h) - m(t)] = h\mu$$

$$H(t) = a(t) + \int_0^t a_0(t-x) dm(x)$$

$$\lim_{t \rightarrow \infty} H(t) = \lim_{t \rightarrow \infty} a(t) + \lim_{t \rightarrow \infty} \int_0^t a_0(t-x) dm(x)$$

↳ as $t \rightarrow \infty$
should converge to

$$u \int_0^\infty a_0(x) dx$$

KEY RENEWAL THEOREM

Let A and F be proper distributions s.t.

$$\int_0^\infty F'(u) du = \frac{1}{u}$$

① If F is non-lattice, then,

$$\lim_{t \rightarrow \infty} \int_0^t a_0(t-u) dm(u) = u \int_0^\infty a_0(u) du$$

(and a_0 is direct
Riemann integrable.)

② If $A=F$ and F is lattice with span 'd', then, $\forall t > 0$

$$\lim_{n \rightarrow \infty} \int_0^{t+nd} a_0(t+nd-u) dm(u) = u d \sum_{n=0}^\infty a_0(t+nd)$$

Lemma: If a function $a_0: \mathbb{R}^+ \rightarrow \mathbb{R}$ is non-negative, monotone non-increasing, reimann integrable, then it is Direct Riemann Integrable (DRI).

(sufficient but not necessary).

$$\text{e.g. } P(Y(t) > y) = A^c(t+y) + \int_0^t F^c(t+y-x) dm(x)$$

finding $\lim_{t \rightarrow \infty} P(Y(t) > y)$

Note :- $\lim_{t \rightarrow \infty} A^c(t+y) = 0$

and $\lim_{t \rightarrow \infty} \int_0^t F^c(t+y-x) dm(x)$ ↗ monotone, non-increasing,
non-negative

using KRT $= u \lim_{t \rightarrow \infty} \int_0^t F^c(x-y) dx = u \int_y^\infty F^c(x) dx$

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So, $\lim_{t \rightarrow \infty} P(Y(t) > y) = \int_y^{\infty} F^C(x) \cdot dx$ ∞

$$\int_y^{\infty} F^C(x) \cdot dx = \int_y^{\infty} (1 - F(x))^p \cdot dx$$

(KRT)

MAGNET JAWAHAR KERI

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