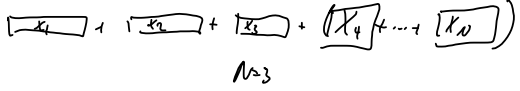


Lemma 1 If X_1, \dots, X_n are independent $X_i \sim \text{Exp}(\lambda_i)$ $\lambda_i \neq \lambda_j$
 then $f_{X_1, \dots, X_n}(x) = \prod_{i=1}^n (c_i \lambda_i \exp(-\lambda_i x))$
 where $c_i = \prod_{j \neq i} \frac{\lambda_j}{\lambda_j - \lambda_i}$
 Hence $1 - F(t) = \frac{\prod_{i=1}^n (c_i \lambda_i \exp(-\lambda_i t))}{\sum_{i=1}^n (c_i \lambda_i \exp(-\lambda_i t))}$
 failure rate \dots
 $\lim_{t \rightarrow \infty} r(t) = \lambda_j = \min(\lambda_i)$

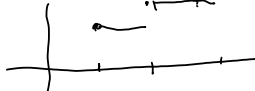
feb. 27-10:14

So if we know that the survival is up to t
 then the conditional probability of dying
 has approximately p.d.f $\lambda_j \exp(-\lambda_j t)$.
 Coxian distributions have a random
 number of terms in the sum.
 Let X_1, \dots, X_n be independent $X_i \sim \text{Exp}(\lambda_i)$ $\lambda_i \neq \lambda_j$
 Let N be independent of X_1, \dots, X_n and
 $P_n = P(N=n)$. Then $Y = \sum_{i=1}^N X_i$ is a Coxian
 random variable.
 The n.d.f of Y is
 $f_Y(t) = \sum_{n=1}^{\infty} f_Y(t|N=n) P_n$
 $= \sum_{n=1}^{\infty} f_{X_1, \dots, X_n}(t|N=n) P_n$
 $= \sum_{n=1}^{\infty} \prod_{i=1}^n f_{X_i}(t) P_n$
 $= \sum_{n=1}^{\infty} P_n \cdot \prod_{i=1}^n (c_i \lambda_i \exp(-\lambda_i t))$

feb. 27-10:23

Visualization

 5.3 The Poisson process
 The Poisson process is the simplest process
 in continuous time. The process describes
 the occurrence of events and now the
 occurrence can be described by continuously
 distributed random variables.
 Examples: (i) Customers entering
 a shop
 (ii) Children born in a city/country
 (iii) Goals scored by a soccer
 player.

feb. 27-10:32

A basic tool is the counting process $N(t)$
 which indicates the number of events up to
 and including t . Thus
 (i) $N(t) \geq 0$
 (ii) $N(t)$ integer valued
 (iii) $N(t)$ is nondecreasing so if $s < t$ $N(s) \leq N(t)$
 (iv) For $s < t$, $N(t) - N(s)$ is the
 number of events in the half open
 interval $(s, t]$. Thus $N(t)$ is
 continuous from the right.

 There are two important properties
 that a counting process may have
 or not

feb. 27-10:40

(i) Independent increments which mean that
 number of events in disjoint intervals
 are independent.
 (ii) Stationary increments mean that the
 number of events in an interval only depends
 on the length of the interval.
 Notation: f function is $o(h)$ if
 $\frac{f(h)}{h} \rightarrow 0$ as $h \rightarrow 0$
 Examples, x^2 is $o(h)$, x is not $o(h)$,
 If f, g are $o(h)$ then $f+g$ is $o(h)$
 $c \cdot f$ is $o(h)$ if f is $o(h)$. All linear
 combination of function which are $o(h)$
 are also $o(h)$.
 Example: $X \sim f(x)$
 $P(t < X < t+h) = \frac{F(t+h) - F(t)}{h} \cdot h$
 $= [f(t) + o(h)] \cdot h = f(t) \cdot h + o(h)$

feb. 27-10:45

Definition: The counting process $\{N(t); t \geq 0\}$
 is a Poisson process with rate λ if
 (i) $N(0) = 0$
 (ii) $\{N(t); t \geq 0\}$ has independent increments
 (iii) $P(N(t+h) - N(t) = 1) = \lambda \cdot h + o(h)$
 (iv) $P(N(t+h) - N(t) \geq 2) = o(h)$
 Remark $P(N(t+h) - N(t) = 0) = 1 - \lambda \cdot h + o(h)$
 Theorem 5.1 If $\{N(t); t \geq 0\}$ is a Poisson
 process with rate λ , then $N(s+t) - N(s) \geq 0$
 i.e. the number of events in $(s, s+t]$
 is Poisson distributed with rate λ .
 Proof: $g(t) = E[e^{-u N(t)}]$
 Then $g(t+h) = E[\exp(-u N(t+h))]$
 $= E[e^{-u [N(t+h) - N(t) + N(t)]}]$
 $= E[e^{-u [N(t+h) - N(t)]} \cdot e^{-u N(t)}]$
 $= E[e^{-u [N(t+h) - N(t)]}] \cdot e^{-u N(t)}$

feb. 27-10:59

By independent increments
 $= E[e^{-\lambda N(t+h)}] \cdot E[e^{-\lambda(N(t+h)-N(t))}]$
 $= g(t) \cdot E[e^{-\lambda(N(t+h)-N(t))}]$
 But $E[e^{-\lambda(N(t+h)-N(t))}]$
 $= E\{e^{-\lambda(N(t+h)-N(t))} \mid N(t+h)-N(t)=0\}$
 $+ E\{e^{-\lambda(N(t+h)-N(t))} \mid N(t+h)-N(t)=1\}$
 $+ E\{e^{-\lambda(N(t+h)-N(t))} \mid N(t+h)-N(t) \geq 2\}$
 $= [1 - \lambda h + o(h)] + e^{-\lambda h} [\lambda h + o(h)] + o(h)$
 $= 1 - \lambda h + e^{-\lambda h} \lambda h + o(h)$
 $= 1 + (e^{-\lambda h} - 1) \lambda h + o(h)$


feb. 27-11:27

Hence $g(t+h) = g(t) [1 + (e^{-\lambda h} - 1) \lambda h + o(h)]$
 $\frac{g(t+h) - g(t)}{h} = g(t) \lambda (e^{-\lambda h} - 1) + \frac{o(h)}{h}$
 Let $h \rightarrow 0$, then
 $g'(t) = g(t) \lambda (e^{-\lambda h} - 1)$ or $\frac{g'(t)}{g(t)} = \lambda (e^{-\lambda h} - 1)$
 Hence $\frac{d}{dt} \log g(t) = \lambda (e^{-\lambda h} - 1)$
 $\log g(t) = \lambda (e^{-\lambda h} - 1) \cdot t + C$
 But $g(0) = 1 \Rightarrow C = 0$ and $g(t) = e^{\lambda t (e^{-\lambda h} - 1)}$
 But if $X \sim \text{Po}(\lambda t)$, $E[e^{-\lambda X}] = e^{-\lambda t} \sum_{j=0}^{\infty} \frac{(\lambda t)^j}{j!} e^{-\lambda t}$
 $= e^{-\lambda t} e^{\lambda t (e^{-\lambda h} - 1)}$

feb. 27-11:34

Hence the two Laplace transforms are equal
 And since the Laplace transform determines
 the distribution $N(t)$ is Poisson distributed
 with parameter λt
 Let $N_s(t) = N(s+t) - N(s)$. Then
 $N_s(t)$ is a counting process starting in 0
 and satisfying the requirements in the
 definition of a Poisson process.
 Hence $N(s+t) - N(s) = N_s(t)$ is Poisson
 distributed with parameter λt .

feb. 27-11:40

Heuristic argument:
 If $Y \sim \text{Bin}(n, p)$ then $P(Y=y) = \binom{n}{y} p^y (1-p)^{n-y}$
 $\rightarrow \frac{n!}{y!} p^y (1-p)^{n-y} \quad n, p \rightarrow \lambda$
 see section 2.2.4, p 29 in Ross

 "success": event in interval of length t/k
 $P(\text{success}) = P(\text{event in interval of length } t/k)$
 $= \lambda \cdot \frac{t}{k} + o(t/k)$
 by stationarity
 Hence $P(y \text{ success}) = \binom{n}{y} [\lambda \cdot \frac{t}{k} + o(t/k)]^y$
 $[\lambda \cdot \frac{t}{k} + o(t/k)]^{n-y}$

feb. 27-11:45

since $R \cdot [\lambda \cdot \frac{t}{k} + o(\frac{t}{k})] \rightarrow \lambda t$ as $k \rightarrow \infty$
 the $P(y \text{ success}) \rightarrow \frac{(\lambda t)^y}{y!} e^{-\lambda t}$
 from the Poisson approximation to the Binomial
 distribution
 It so, the distribution of $N(t+s) - N(s)$
 is the same for all s , so $N(t+s) - N(s)$
 has stationary increments. Thus also the
 Poisson process has stationary increments.

feb. 27-11:52

Next we will look into the distribution of
 time to first event, T_1 , and the times between
 successive events which are denoted by
 T_1, T_2, \dots . The variables T_1, T_2, \dots
 are called the sequence of interarrival
 times.

feb. 27-11:58