

## XI Random process in continuous time

We'll consider  $\{N_t\}_{t \geq 0}$  a continuous family of integer-valued r.v.

Example: About emails received in a letter box.

Let  $N_t$  denote the # of emails received up to time  $t$ .

1)  $N_t$  is a r.v. taking values of  $0, 1, 2, \dots$

2)  $N_0 = 0$

3)  $N_s \leq N_t \Leftrightarrow s \leq t$

4) Independence: if  $0 \leq s < t$  then the emails received between  $(s, t]$  is indep. of the emails received before  $s$

5) Arrival rate:  $\exists \lambda > 0$  called arrival rate s.t. for  $h$  small enough

$$P(N_{t+h} = n+1 | N_t = n) = \lambda h + o(h)$$

$$P(N_{t+h} = n | N_t = n) = 1 - \lambda h + o(h) \leftarrow \begin{matrix} \text{not} \\ \text{equivalent} \end{matrix}$$

For the probability of receiving 2 or more emails in  $(t, t+h)$ :

$$P(N_{t+h} = n+2 | N_t = n) = 1 - P(N_{t+h} \in \{n, n+1\} | N_t = n)$$

$$= 1 - (\lambda h + o(h)) - (1 - \lambda h + o(h)) = o(h) \text{ for } h \text{ small enough}$$

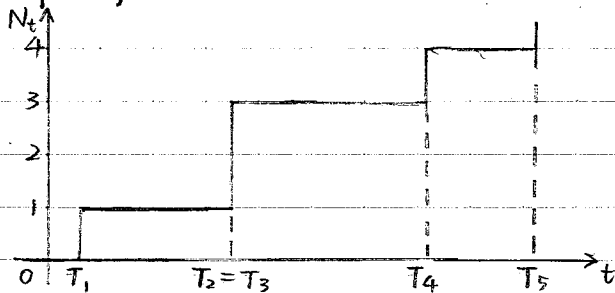
Def. A Poisson process with rate  $\lambda > 0$

is a family of integer valued r.v.  $\{N_t\}_{t \geq 0}$  satisfying conditions 1) ~ 5)

It can be used in models:

- the arrival of customers in a shop
- the clicks of Geiger counter for the detection of some particles  
(more realistic for long half-life of particles)

Graph of  $N_t$



Let  $T_i$  be the time of arrival of the  $i^{\text{th}}$  email

$$\Rightarrow T_i := \inf \{t | N_t = i\}$$

Then  $0 \leq T_0 \leq T_1 \leq T_2 \leq \dots$

and  $\{T_i\}_{i \in \mathbb{N}}$  is a sequence of r.v. which determines

$$N_t = \max\{n \mid T_n \leq t\}$$

The sequence of  $\{T_i\}_{i \in \mathbb{N}}$  is the inverse process of  $\{N_t\}_{t \geq 0}$

Thm. For each  $t > 0$ , the r.v.  $N_t$  has a Poisson distribution

with parameter  $\lambda t$ . That is for  $k \in \mathbb{N}$  and  $t > 0$

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

Corollary:  $\mathbb{E}(N_t) = \lambda t$ ,  $\text{var}(N_t) = \lambda t$

Proof: Let  $P_k(t) = P(N_t = k)$

Consider  $h > 0$  small enough

Then by the partition thm,

$$\begin{aligned} P_k(t+h) &= P(N_{t+h} = k) = \sum_{i=0}^k P(N_{t+h} = k \mid N_t = i) P(N_t = i) \\ &= P(N_{t+h} = k \mid N_t = k) P_k(t) + P(N_{t+h} = k \mid N_t = k-1) P_{k-1}(t) + o(h) \\ &= (1 - \lambda h + o(h)) P_k(t) + (\lambda h + o(h)) P_{k-1}(t) + o(h) \end{aligned}$$

$$\Leftrightarrow P_k(t+h) - P_k(t) = \lambda h (P_{k-1}(t) - P_k(t)) + o(h)$$

$$\Leftrightarrow \frac{1}{h} (P_k(t+h) - P_k(t)) = \lambda (P_{k-1}(t) - P_k(t)) + o(1)$$

$$\xrightarrow{h \rightarrow 0} \left. \begin{aligned} P'_k(t) &= \lambda (P_{k-1}(t) - P_k(t)) \\ \text{For } k=0, P'_0(t) &= -\lambda P_0(t) \end{aligned} \right\} \text{system of difference-differential equations}$$

Boundary condition:

$$P_k(0) = \begin{cases} 1, & \text{if } k=0 \\ 0, & \text{if } k \in \mathbb{N}_+ \end{cases}$$

3 ways of solving this system:

① Induction over  $k$ ; ② Generating functions; ③ Inspection. □

↓  
Obtain one differential equation of generating function

Recall  $T_i = \inf\{t \mid N_t = i\}$  and set  $X_i := T_i - T_{i-1}$  (inter-arrival times)

Thm. In a Poisson process with parameter  $\lambda$ ,  $X_1, X_2, \dots$  are independent random variables, each having the exponential distribution with parameter  $\lambda$ .

$$X(t) = \begin{cases} 0 & \text{if } t \leq 0 \\ 1 - e^{-\lambda t} & \text{if } t > 0 \end{cases}$$

Note that the  $X_i$  are independent (loss of memory)

Def. A positive r.v.  $X$  has the loss of memory property if  $P(X > u+v \mid X \geq u) = P(X > v) \quad \forall u, v \geq 0$

Prop.

A continuous positive r.v. has the loss of memory property iff  $X$  is the exponential distribution.

Proof: If  $\lambda > 0$ ,  $u, v \geq 0$ ,  $X$  is the exp. distribution,

$$\begin{aligned} P(X > u+v \mid X > u) &= \frac{P(X > u+v \text{ and } X > u)}{P(X > u)} = \frac{P(X > u+v)}{P(X > u)} = \frac{e^{-\lambda(u+v)}}{e^{-\lambda u}} \\ &= e^{-\lambda v} = P(X > v) \end{aligned}$$

Conversely, set  $G(u) := P(X > u) \quad \forall u \geq 0$

$$\left. \begin{aligned} P(X > u+v \mid X > u) &= \frac{P(X > u+v)}{P(X > u)} = \frac{G(u+v)}{G(u)} \\ \text{By assumption } \rightarrow &= P(X > v) = G(v) \end{aligned} \right\} \Leftrightarrow G(u)G(v) = G(u+v)$$

Since  $G$  is non- $\nearrow$ , the only solution is  $G(u) = e^{-\lambda u}$  for  $\lambda > 0$ .  $\square$

Essentially, Property 4) of Poisson process

$\Rightarrow X$  must have the loss of memory property

Remark

If  $\{X_i\}_{i \in \mathbb{N}}$  is a family of ind. exp. dist. with parameter  $\lambda$

We set  $T_1 = X_1, T_2 = X_1 + X_2, \dots, T_n = \sum_{j=1}^n X_j$ , and  $N_t = \max\{k \mid T_k \leq t\}$

Then  $\{N_t\}_{t \geq 0}$  is a Poisson process with parameter  $\lambda$ .

### Application to population growth:

Time  $t=0$ ,  $I \in \mathbb{N}$  of amoebas in a pond, each of them can divide at a random time with the rules:  $\exists \lambda > 0$  birth rate s.t.

- 1)  $P(\text{division})$  in  $(t, t+h)$  is  $\lambda h + o(h)$
- 2)  $P(\text{no division})$  in  $(t, t+h)$  is  $1 - \lambda h + o(h)$
- 3) Each one is independent

Let  $M_t = \#$  of amoebas at time  $t$  and set  $P_k(t) = P(M_t = k)$

$$P(M_{t+h} = k | M_t = k) = P(\text{no division})^k = (1 - \lambda h + o(h))^k = 1 - k\lambda h + o(h)$$

for  $k$  small enough

$$P(M_{t+h} = k+1 | M_t = k) = P(1 \text{ division}) = \binom{k}{1} (\lambda h + o(h)) (1 - \lambda h + o(h))^{k-1} = k\lambda h + o(h)$$

$$P(M_{t+h} \geq k+2 | M_t = k) = o(h)$$

Thm. If  $M_0 = I$  and  $t > 0$

$$P(M_t = k) = \binom{k-1}{I-1} e^{-\lambda I t} (1 - e^{-t})^{k-1} \quad \text{for } k \geq I$$

### Application to population growth + death:

Division rate as before; death rate with  $\mu > 0$

For each amoeba, for the interval  $(t, t+h)$  one has

- death with prob.  $P = \mu h + o(h)$
- single division with  $P = \lambda h + o(h)$
- no change with prob.  $P = (1 - \mu h + o(h)) (1 - \lambda h + o(h)) = 1 - (\mu + \lambda) h + o(h)$

Set  $M_t, P_k(t)$  as before

$$\Rightarrow P'_k(t) = \lambda(k-1)P_{k-1}(t) - (\lambda + \mu)kP_k(t) + \mu(k+1)P_{k+1}(t)$$

Here the iteration method doesn't work because of  $\uparrow$

One needs a differential equation for the moment generating function,

and we find  $E(e^{sM_t}) = \dots$

$$\underset{\parallel}{G(s, t)}$$