

# Lec 9

## Review of Poisson Approximation

## Further applications of Poisson Process

### Recall

Thus, Conditioned on  $X(t) = n$ , the  $n$  arrival times  $T_1, \dots, T_n$  has the same distribution as  $(V_1, \dots, V_n)$ , where  $(V_1, \dots, V_n)$  is the ordered version of  $(U_1, \dots, U_n)$ , and each  $U_i \sim \text{Uniform}(0, t)$  independently.

### A generalization of thinning

$k$  types of events. each time  $s$ .  $P_i(s) \quad \sum_{i=1}^k P_i(s) = 1.$

Thm \*  $N_i(t)$ : arrival of type  $i$  events.

independent Poisson variables with mean

$$E[N_i(t)] = \lambda \int_0^t P_i(s) ds.$$

Example M/G/∞ queue.

Customer arrival  $\sim$  Poisson( $\lambda$ ).

infinite number of servers

Service time: independent  $G$ .

type 1

$X(t)$ : # of customers completed services before  $t$ .

$Y(t)$ : # of customers being served at time  $t$ .

type 2

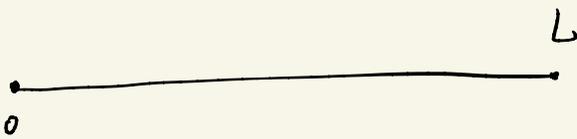
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By \*  $P_X(s) = G(t-s)$ .  $P_Y(s) = 1 - G(t-s) =: \overline{G(t-s)}$

$$E[X(t)] = \lambda \int_0^t P_X(s) ds = \lambda \int_0^t G(t-s) ds = \lambda \int_0^t G(y) dy$$

$$E[Y(t)] = \lambda \int_0^t P_Y(s) ds = \lambda \int_0^t \overline{G(t-s)} ds = \lambda \int_0^t \overline{G(y)} dy$$

Example # of encounters.

highway 

cars arrives  $\sim$  Poisson( $\lambda$ ).

constant speed  $\sim$   $G$ .

What is the speed minimizing encounters?

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Choose speed  $x$  Travel time  $t_0 = L/x$

$\Rightarrow$  Enter at time  $s$ , leave at time  $s + t_0$ .

Distribution of the travel time.

$$F(t) = \Pr[T \leq t] = \Pr[L/x \leq t]$$
$$= \bar{G}(L/t).$$

Type 1 car: encounter with you:

① enter earlier than  $s$ , leave later than  $s+t_0$ .

② enter later than  $s$ , leave earlier than  $s+t_0$ .

$$\textcircled{1} \quad t < s \Rightarrow t + \bar{T} > s + t_0.$$

$$\textcircled{2} \quad t > s \Rightarrow t + \bar{T} < s + t_0.$$

$$p_i(t) = \begin{cases} P(t + \bar{T} > s + t_0) = \bar{F}(s + t_0 - t) & \text{if } t < s. \\ P(t + \bar{T} < s + t_0) = F(s + t_0 - t) & \text{if } s < t < s + t_0. \\ 0, & \text{if } t > s + t_0. \end{cases}$$

$$E[\text{type 1 cars}] = \lambda \int_0^{\infty} p_i(t) dt$$

$$= \lambda \left( \int_0^s \bar{F}(s + t_0 - t) dt + \int_s^{s+t_0} F(s + t_0 - t) dt \right)$$

$$= \lambda \int_{t_0}^{s+t_0} \bar{F}(y) dy + \lambda \int_0^{t_0} F(y) dy. \quad \dots (\heartsuit)$$

$$\frac{d\heartsuit}{dt_0} = \lambda \left( \bar{F}(s + t_0) - \bar{F}(t_0) + F(t_0) \right) = 0$$

$\bar{F}(s+t_0) \approx 0$  for large  $s$ .

$$\Rightarrow \bar{F}(t_0) = \overline{\bar{F}(t_0)}$$

$$\Rightarrow \bar{F}(0_0) = \frac{1}{2}.$$

The optimal speed is the median of  $G$ .

Example Tracking the # of HIV infections.

# of individuals contract HIV  $\sim \text{Poi}(\lambda)$ .

unknown

incubation time  $\sim G$  — known.

type 1. symptoms shown.

type 2. symptoms not shown.

$$E[N_1(\lambda)] = \lambda \int_0^t G(t-s) ds = \lambda \int_0^t G(y) dy$$

$$E[N_2(\lambda)] = \lambda \int_0^t \bar{G}(y) dy.$$

$$\text{Let } \hat{n}_1 = E[N_1(\lambda)]. \quad \hat{\lambda} = \frac{\hat{n}_1}{\int_0^t G(y) dy}$$
$$\Rightarrow \hat{n}_2 = \hat{n}_1 \cdot \frac{\int_0^t \bar{G}(y) dy}{\int_0^t G(y) dy}.$$

\*. Each  $N_i(t)$  is a non-homogeneous Poisson Process  
 with  $\lambda(t) = \lambda P_i(t)$ .

## Nonhomogeneous Poisson Processes

$\{N(t) : t \geq 0\}$  rate  $\lambda(s)$ . if

①  $N(0) = 0$

②  $N(t)$  has independent increments.

③  $N(t) - N(s)$  is Poisson with mean  $\int_s^t \lambda(r) dr$ .

proof of \*

$$P_r [N_1(t) = n_1, \dots, N_k(t) = n_k]$$

$$= P_r [N_1(t) = n_1, \dots, N_k(t) = n_k \mid N(t) = \sum_{i=1}^k n_i] \cdot P_r [N(t) = \sum_{i=1}^k n_i]$$

$$= \left( \frac{\left( \sum_{i=1}^k n_i \right)!}{n_1! \dots n_k!} P_1^{n_1} \dots P_k^{n_k} \right) \cdot e^{-\lambda t} \frac{(\lambda t)^{\sum n_i}}{\left( \sum n_i \right)!}$$

$$= \prod_{i=1}^k e^{-\lambda t P_i} \frac{(\lambda t P_i)^{n_i}}{n_i!}$$