

NHP  $\Rightarrow \lambda(t)$  intensity

①  $N(0) = 0$

② independent increments

③  $N(t) - N(s) \sim \text{Poisson} \left( \int_s^t \lambda(u) du \right)$

$\Lambda(t) =$  Expected number of arrivals by time  $t$

$$= \int_0^t \lambda(s) ds$$

$$\Lambda(t) - \Lambda(s) = \int_s^t \lambda(u) du$$

①  $\lambda(t) \leq \lambda_{\max} \quad 0 \leq t \leq T$

$\Rightarrow$  ① generate the arrivals of a homogeneous poisson process w/ intensity  $\lambda_{\max}$

$$s_1, \dots, s_n$$

② keep  $s_i$  w/ prob

$$p(s_i) = \frac{\lambda(s_i)}{\lambda_{\max}}$$

(thinning)

② order statistic

→ sample  $N(T) \sim \text{Poisson}(\Lambda(T))$

→ given  $N(T) = k$   
sample  $\gamma_1, \dots, \gamma_k$

$$F(t) = \frac{\Lambda(t)}{\Lambda(T)}$$

↗ inverse transform

$$f(t) = \frac{\lambda(t)}{\Lambda(T)}$$

↘ rejection sampling

③ Inverse method

$\Lambda(t)$  is invertible

$\Lambda^{-1}(t)$  exists

① sample arrival times  
Poisson Process (1) on  $[0, \Lambda(T)]$

$s_1, \dots, s_k$

② arrival times MHP

$\Lambda^{-1}(s_1), \dots, \Lambda^{-1}(s_k)$

Thinning

Good

$\lambda(t)$  is  
bounded

Bad

$\lambda(t)$  is not  
bounded

$\lambda_{\max} \gg \lambda(t)$  for  
most  $t$

Order

works nicely  
if comput  $\Lambda$  is  
slow

if we don't have  
a good proposal  
density for rejection  
sampling

Inverse

$\Lambda^{-1}$  exists  
and easily  
comput

$\lambda(t) = 0$  on some  
interval  
(not invertible)  
computing inverse  
computationally

slow

I >  $\mathcal{L}$  invertible  $\rightarrow$  Inverse method  
and  $\mathcal{L}^{-1}$  fast to compute?

$\downarrow N$

do we have a good  
proposal density  $\rightarrow$  order statistics  
for  $\frac{\lambda(t)}{\mathcal{L}(T)}$ ?

$\downarrow N$

I >  $\lambda$  bounded?  $\rightarrow$  thinning

$\downarrow N$

need to deal  $\rightarrow$  inverse method  
w/ computation  
 $\mathcal{L}^{-1}$  being slow

Simulate from a MHP  
using data

$$t_1 = 0.25$$

$$t_2 = 0.7$$

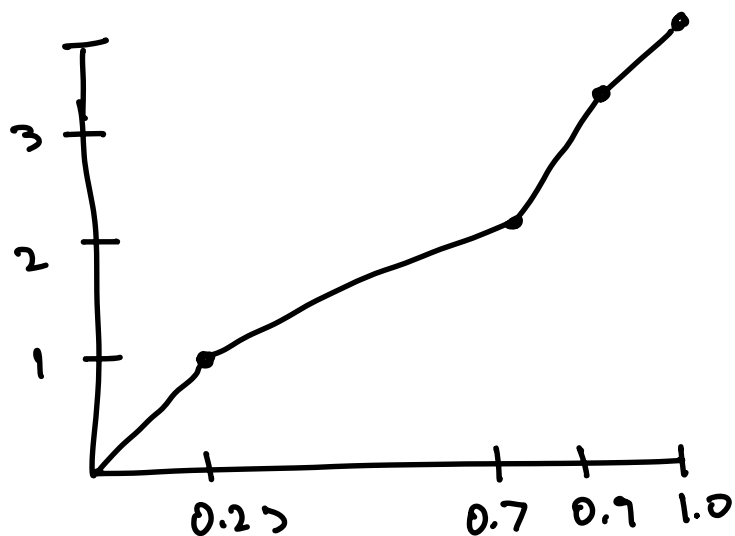
$$t_3 = 0.9$$

$$\mathcal{L}(0.25) = 1$$

$$\mathcal{L}(1) = 4$$

$$\mathcal{L}(0.7) = 2$$

$$\mathcal{L}(0.9) = 3$$



I observe one set of arrivals  
on time  $[0, 1]$

$$t_1, \dots, t_n$$

$$t_0 = 0$$

$$t_{n+1} = 1$$

$$\Lambda(t) = \begin{cases} i + \frac{1}{t_{i+1} - t_i} (t - t_i) & , t \in [t_i, t_{i+1}) \end{cases}$$

$$\text{day } 1: [t_{11}, \dots, t_{1n_1}]$$

$$n = \sum_{i=1}^k n_i$$

$\vdots$

$$\text{day } k: [t_{k1}, \dots, t_{kn_k}]$$

$$t_{c1}, \dots, t_{cn}$$

$$t_{c0} = 0$$

$$t_{c(n+1)} = 1$$

$$\Lambda(t_{c1}) = \frac{1}{k}$$

$$\Lambda(t_{c2}) = \frac{2}{k}$$

$$\Lambda(t_{cn}) = \frac{n}{k}$$

$\Rightarrow$  all we need to do is  
instead of linearly interpolating pieces  
between  $(0,0), (t_1, 1), \dots$

$$(0,0), (t_{cn}, \frac{1}{k}), (t_{cn}, \frac{2}{k}), \dots$$

$\Delta \rightarrow \Delta y$   
x