LECTURE 15

Poisson process — II

- Readings: Finish Section 6.2.
- Review of Poisson process
- Merging and splitting
- Examples
- Random incidence

Review

- Defining characteristics
- Time homogeneity: $P(k, \tau)$
- Independence
- Small interval probabilities (small δ):

$$P(k,\delta) \approx \begin{cases} 1 - \lambda \delta, & \text{if } k = 0, \\ \lambda \delta, & \text{if } k = 1, \\ 0, & \text{if } k > 1. \end{cases}$$

• N_{τ} is a Poisson r.v., with parameter $\lambda \tau$:

$$P(k,\tau) = \frac{(\lambda \tau)^k e^{-\lambda \tau}}{k!}, \qquad k = 0, 1, \dots$$

$$\mathbf{E}[N_{\tau}] = \mathsf{var}(N_{\tau}) = \lambda \tau$$

• Interarrival times (k = 1): exponential:

$$f_{T_1}(t) = \lambda e^{-\lambda t}, \quad t \ge 0, \qquad \mathbf{E}[T_1] = 1/\lambda$$

• Time Y_k to kth arrival: Erlang(k):

$$f_{Y_k}(y) = \frac{\lambda^k y^{k-1} e^{-\lambda y}}{(k-1)!}, \qquad y \ge 0$$

Poisson fishing

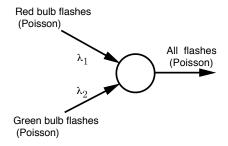
- Assume: Poisson, $\lambda = 0.6/\text{hour}$.
- Fish for two hours.
- if no catch, continue until first catch.
- a) P(fish for more than two hours)=
- $\begin{tabular}{ll} \bf P(fish\ for\ more\ than\ two\ and\ less\ than \\ five\ hours)= \end{tabular}$
- c) P(catch at least two fish)= $E[\#|\text{catch is } \lambda \text{ heres}] P(...) + E(\#|\text{he catch is } \lambda \text{ heres}) P(...)$ $E[\text{mumber of fish}] = \sum_{k=0}^{p(k)} \frac{k}{k} P(k, \lambda) = \chi_{\Sigma} \cdot n_{0} \cdot \lambda$ $E[\#|\text{two hears}] P(\lambda \text{ heres}) + E(\#|\text{nere than } \lambda) P(\text{here } \lambda) = \sum_{k=0}^{p(k)} \frac{k}{k} P(k, \lambda) = \chi_{\Sigma} \cdot n_{0} \cdot \lambda$ $= \sum_{k=0}^{p(k)} \frac{k}{k} P(k, \lambda) = \chi_{\Sigma} \cdot n_{0} \cdot \lambda$ $= \sum_{k=0}^{p(k)} \frac{k}{k} P(k, \lambda) = \chi_{\Sigma} \cdot n_{0} \cdot \lambda$ $= \sum_{k=0}^{p(k)} \frac{k}{k} P(k, \lambda) = \chi_{\Sigma} \cdot n_{0} \cdot \lambda$ $= \sum_{k=0}^{p(k)} \frac{k}{k} P(k, \lambda) = \chi_{\Sigma} \cdot n_{0} \cdot \lambda$ $= \sum_{k=0}^{p(k)} \frac{k}{k} P(k, \lambda) = \chi_{\Sigma} \cdot n_{0} \cdot \lambda$ $= \sum_{k=0}^{p(k)} \frac{k}{k} P(k, \lambda) = \chi_{\Sigma} \cdot n_{0} \cdot \lambda$ $= \sum_{k=0}^{p(k)} \frac{k}{k} P(k, \lambda) = \chi_{\Sigma} \cdot n_{0} \cdot \lambda$ $= \sum_{k=0}^{p(k)} \frac{k}{k} P(k, \lambda) = \chi_{\Sigma} \cdot n_{0} \cdot \lambda$ $= \sum_{k=0}^{p(k)} \frac{k}{k} P(k, \lambda) = \chi_{\Sigma} \cdot n_{0} \cdot \lambda$ $= \sum_{k=0}^{p(k)} \frac{k}{k} P(k, \lambda) = \chi_{\Sigma} \cdot n_{0} \cdot \lambda$ $= \sum_{k=0}^{p(k)} \frac{k}{k} P(k, \lambda) = \chi_{\Sigma} \cdot n_{0} \cdot \lambda$ $= \sum_{k=0}^{p(k)} \frac{k}{k} P(k, \lambda) = \chi_{\Sigma} \cdot n_{0} \cdot \lambda$ $= \sum_{k=0}^{p(k)} \frac{k}{k} P(k, \lambda) = \chi_{\Sigma} \cdot n_{0} \cdot \lambda$ $= \sum_{k=0}^{p(k)} \frac{k}{k} P(k, \lambda) = \chi_{\Sigma} \cdot n_{0} \cdot \lambda$ $= \sum_{k=0}^{p(k)} \frac{k}{k} P(k, \lambda) = \chi_{\Sigma} \cdot n_{0} \cdot \lambda$ $= \sum_{k=0}^{p(k)} \frac{k}{k} P(k, \lambda) = \chi_{\Sigma} \cdot n_{0} \cdot \lambda$ $= \sum_{k=0}^{p(k)} \frac{k}{k} P(k, \lambda) = \chi_{\Sigma} \cdot n_{0} \cdot \lambda$ $= \sum_{k=0}^{p(k)} \frac{k}{k} P(k, \lambda) = \chi_{\Sigma} \cdot n_{0} \cdot \lambda$ $= \sum_{k=0}^{p(k)} \frac{k}{k} P(k, \lambda) = \chi_{\Sigma} \cdot n_{0} \cdot \lambda$ $= \sum_{k=0}^{p(k)} \frac{k}{k} P(k, \lambda) = \chi_{\Sigma} \cdot n_{0} \cdot \lambda$ $= \sum_{k=0}^{p(k)} \frac{k}{k} P(k, \lambda) = \chi_{\Sigma} \cdot n_{0} \cdot \lambda$ $= \sum_{k=0}^{p(k)} \frac{k}{k} P(k, \lambda) = \chi_{\Sigma} \cdot n_{0} \cdot \lambda$ $= \sum_{k=0}^{p(k)} \frac{k}{k} P(k, \lambda) = \chi_{\Sigma} \cdot n_{0} \cdot \lambda$ $= \sum_{k=0}^{p(k)} \frac{k}{k} P(k, \lambda) = \chi_{\Sigma} \cdot n_{0} \cdot \lambda$ $= \sum_{k=0}^{p(k)} \frac{k}{k} P(k, \lambda) = \chi_{\Sigma} \cdot n_{0} \cdot \lambda$ $= \sum_{k=0}^{p(k)} \frac{k}{k} P(k, \lambda) = \chi_{\Sigma} \cdot n_{0} \cdot \lambda$ $= \sum_{k=0}^{p(k)} \frac{k}{k} P(k, \lambda) = \chi_{\Sigma} \cdot n_{0} \cdot \lambda$ $= \sum_{k=0}^{p(k)} \frac{k}{k} P(k, \lambda) = \chi_{\Sigma} \cdot n_{0} \cdot \lambda$ $= \sum_{k=0}^{p(k)} \frac{k}{k} P(k, \lambda) = \chi_{\Sigma} \cdot n_{0} \cdot \lambda$ $= \sum_{k=0}^{p(k)} \frac{k}{k} P(k, \lambda) = \chi_{\Sigma} \cdot n_{0} \cdot \lambda$ $= \sum_{k=0}^{p(k)} \frac{k}{k} P(k, \lambda) = \chi_{\Sigma} \cdot n_{0} \cdot \lambda$ $= \sum_{k=0}^{p(k)} \frac{k}{k} P$

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f) E[total fishing time]=

Merging Poisson Processes (again)

 Merging of independent Poisson processes is Poisson



– What is the probability that the next arrival comes from the first process?

Light bulb example

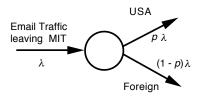
- Each light bulb has independent, exponential(λ) lifetime
- Install three light bulbs.
 Find expected time until last light bulb dies out.

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\begin{split} & P\left( \left. \mathsf{Mar}\left\{T_1, T_2, T_3\right\} \leqslant t \right. \right) = P\left(T_1 \leqslant t \text{ and } T_4 \leqslant t \text{ ond } T_4 \leqslant t \right. \\ & = P\left(T_1 \leqslant t \right) P\left(T_4 \leqslant t \right) P\left(T_4 \leqslant t \right) \\ & = \left( \left( -e^{2 \epsilon^2} \right) \left( \left( 1 - e^{2 \epsilon^2} \right) \left( 1 - e^{2 \epsilon^2} \right) \right) \right. \\ & \left. \left. \left( T_4 \leqslant t \right) P\left(T_4 \leqslant t \right) \left( \left( 1 - e^{2 \epsilon^2} \right) \left( 1 - e^{2 \epsilon^2} \right) \right) \right. \\ & \left. \left. \left( T_4 \leqslant t \right) P\left(T_4 \leqslant t \right) \left( \left( 1 - e^{2 \epsilon^2} \right) \left( 1 - e^{2 \epsilon^2} \right) \right) \right. \\ & \left. \left. \left( T_4 \leqslant t \right) P\left(T_4 \leqslant t \right) \left( 1 - e^{2 \epsilon^2} \right) \left( 1 - e^{2 \epsilon^2} \right) \right. \\ & \left. \left( T_4 \leqslant t \right) P\left(T_4 \leqslant t \right) \left( 1 - e^{2 \epsilon^2} \right) \left( 1 - e^{2 \epsilon^2} \right) \right. \\ & \left. \left( T_4 \leqslant t \right) P\left(T_4 \leqslant t \right) \left( 1 - e^{2 \epsilon^2} \right) \left( 1 - e^{2 \epsilon^2} \right) \right. \\ & \left. \left( T_4 \leqslant t \right) P\left(T_4 \leqslant t \right) \left( 1 - e^{2 \epsilon^2} \right) \left( 1 - e^{2 \epsilon^2} \right) \right. \\ & \left. \left( T_4 \leqslant t \right) P\left(T_4 \leqslant t \right) \left( 1 - e^{2 \epsilon^2} \right) \left( 1 - e^{2 \epsilon^2} \right) \left( 1 - e^{2 \epsilon^2} \right) \right. \\ & \left. \left( T_4 \leqslant t \right) P\left(T_4 \leqslant t \right) \left( 1 - e^{2 \epsilon^2} \right) \left( 1 - e^{2 \epsilon^2} \right) \right. \\ & \left. \left( T_4 \leqslant t \right) P\left(T_4 \leqslant t \right) \left( 1 - e^{2 \epsilon^2} \right) \left( 1 - e^{2 \epsilon^2} \right) \left( 1 - e^{2 \epsilon^2} \right) \right. \\ & \left. \left( T_4 \leqslant t \right) P\left(T_4 \leqslant t \right) \left( 1 - e^{2 \epsilon^2} \right) \left( 1 - e^{2 \epsilon^2} \right) \left( 1 - e^{2 \epsilon^2} \right) \right] \right. \\ & \left. \left( T_4 \leqslant t \right) P\left(T_4 \leqslant t \right) \left( T_4 \leqslant t \right) \left( T_4 \leqslant t \right) \left( T_4 \leqslant t \right) \right. \\ & \left. \left( T_4 \leqslant t \right) P\left(T_4 \leqslant t \right) \left( T_4 \leqslant t \right) \left( T_4 \leqslant t \right) \right. \\ & \left. \left( T_4 \leqslant t \right) P\left(T_4 \leqslant t \right) \left( T_4 \leqslant t \right) \left( T_4 \leqslant t \right) \right) \left( T_4 \leqslant t \right) \right. \\ & \left. \left( T_4 \leqslant t \right) P\left(T_4 \leqslant t \right) P\left(T_4 \leqslant t \right) \left( T_4 \leqslant t \right) \left( T_4 \leqslant t \right) \right. \\ & \left. \left( T_4 \leqslant t \right) P\left(T_4 \leqslant t \right) P\left(T_4 \leqslant t \right) \left( T_4 \leqslant t \right) \left( T_4 \leqslant t \right) \right. \\ & \left. \left( T_4 \leqslant t \right) P\left(T_4 \leqslant t \right) P\left(T_4
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Splitting of Poisson processes

 Assume that email traffic through a server is a Poisson process.
 Destinations of different messages are

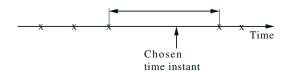
Destinations of different messages are independent.



• Each output stream is Poisson.

Random incidence for Poisson

- Poisson process that has been running forever
- Show up at some "random time" (really means "arbitrary time")



• What is the distribution of the length of the chosen interarrival interval?

Random incidence in "renewal processes"

- Series of successive arrivals
- i.i.d. interarrival times(but not necessarily exponential)

• Example:

Bus interarrival times are equally likely to be 5 or 10 minutes

- If you arrive at a "random time":
- what is the probability that you selected
 a 5 minute interarrival interval?
- what is the expected time to next arrival?

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