Chapter 3.3

Poisson Processes

Performance Evaluation of the Internet of Things (IoT)

Module Course: Performance Evaluation of Distributed Systems

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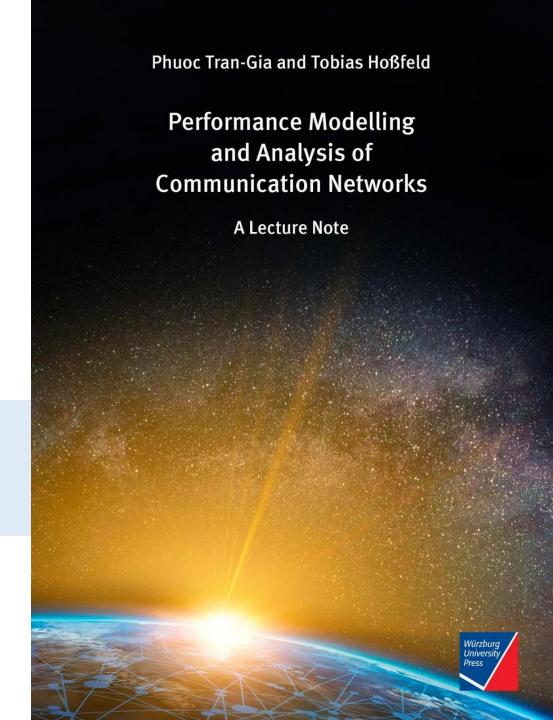
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Tran-Gia, P. & Hossfeld, T. (2021).
Performance Modeling and Analysis of Communication
Networks - A Lecture Note. Würzburg University Press.
https://doi.org/10.25972/WUP-978-3-95826-153-2

Website to download book, exercises, slides and scripts: https://modeling.systems/





Chapter 3

3 Elementary Random Processes

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 - 3.1.1 Definition
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DEFINITION OF A POISSON PROCESS





Poisson Process as Counting Process

- ▶ Counting process $\{N(t), t \ge 0\}$ is a Poisson process with rate λ if being fulfilled
 - \bullet 1. N(0) = 0.
 - 2. N(t) has independent increments.
 - 3. N(t) has stationary increments.

▶ Independent increments

- Number of arrivals during non-overlapping time intervals are independent r.v.s.
- Consider time series: $0 \le t_0 < t_1 < \dots < t_n$
- $N(t_0), N(t_1) N(t_0), N(t_2) N(t_1), ...$ are independent

▶ Stationary increments

- Number of arrivals between t and $t + \tau$ depends only on the length of the interval τ , not on the starting point t.
- ► For Poisson process: $N(t) \sim POIS(\lambda t)$ and $N(t + \tau) N(t) = N(\tau)$





Poisson Process as Renewal Process

- Poisson process with rate λ
 - renewal process with interarrival times following an exponential distribution
 - $A \sim \text{EXP}(\lambda)$
 - CDF $A(t) = 1 e^{-\lambda t}$
 - PDF $a(t) = \lambda e^{-\lambda t}$
 - Mean $E[A] = \frac{1}{\lambda}$
- ▶ Both definitions of a Poisson process (counting process, renewal process) are equivalent

PROPERTIES OF A POISSON PROCESS

Memoryless property, recurrence time, PASTA property





Poisson Process as Renewal Process

- ightharpoonup Poisson process with rate λ
 - renewal process with interarrival times following an exponential distribution
 - $A \sim \text{EXP}(\lambda)$
 - CDF $A(t) = 1 e^{-\lambda t}$
 - PDF $a(t) = \lambda e^{-\lambda t}$
 - Mean $E[A] = \frac{1}{\lambda}$
- ▶ Both definitions of a Poisson process (counting process, renewal process) are equivalent

Memoryless Property for Poisson Process

► Memoryless property of a positive random variable A for every $t \ge 0$ and $s \ge 0$

$$P(A > s + t | A > s) = P(A > t)$$

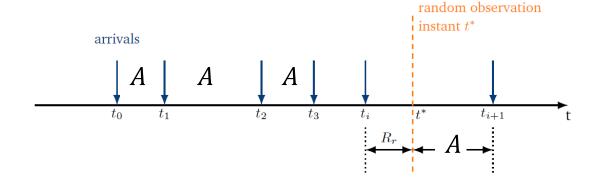
Exponential distribution has memoryless property

$$P(A > s + t | A > s) = e^{-\lambda t} = P(A > t)$$

▶ Recurrence time of Poisson process

$$r(t) = \lambda (1-A(t)) = \lambda e^{-\lambda t} = a(t)$$

$$R(t) = A(t)$$



- Interarrival time and recurrence time follow exponential distribution with rate λ
- Process develops completely independently of its past at any observation time
- Poisson process is memoryless or has the Markov property



PASTA Property

- ▶ PASTA = "Poisson Arrival Sees Time Average"
- \triangleright X^* number of customers in system at a random time t^* (r.v.)
- \triangleright X_A number of customers in system as observed by an arriving customer (r.v.)
- \triangleright PASTA property: probability to see *i* customers is identical for random and arriving customers

$$x^*(i) = x_A(i)$$
, $i = 0,1,...$

- PASTA property is often utilized
 - e.g., blocking probabilities of arriving customers in M/M/n loss systems
 - e.g., derivation of system state probabilities $x^*(i)$ for M/GI/1 delay systems
- ► PASTA requires Poisson arrival process



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PASTA Property





Counter-example: PASTA

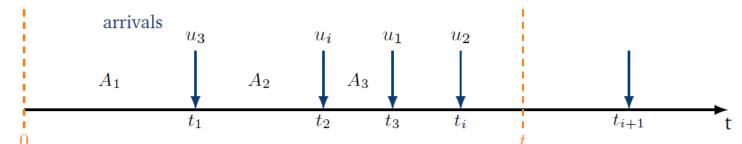
- Simple counter-example that PASTA is not valid for any arrival process
 - D/D/1 queue with E[A] = 2 and E[B] = 1
 - $x^*(i) \neq x_A(i)$





Arrivals in Fixed Interval

Number N(t) of Poisson arrivals in a fixed interval of length t with arrival rate λ



$$N(t) \sim \text{POIS}(\lambda t) \text{ with } E[N(t)] = \frac{1}{\lambda t}$$

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

Uniform distribution of Poisson arrivals in fixed interval

$$u_i \sim U(0, t)$$
 for $i = 1, 2, ..., n$

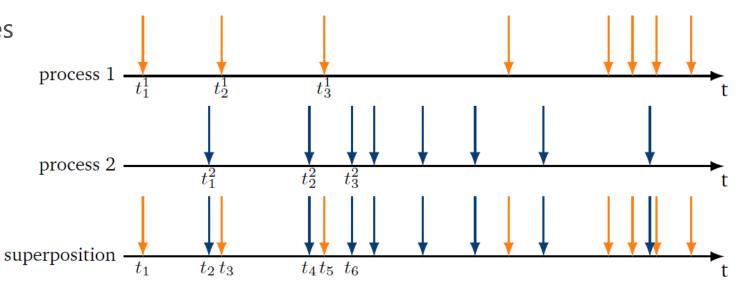
Merging and Splitting of Poisson Processes





Merging of Poisson Processes

- Superposition of Poisson processes remains a Poisson process
 - Process 1: $A_1 \sim \text{EXP}(\lambda_1)$
 - Process 2: $A_2 \sim \text{EXP}(\lambda_2)$
- ► Superposition $A \sim \text{EXP}(\lambda_1 + \lambda_2)$



- ▶ Proof: random observer sees minimum of recurrence times of both processes
 - recurrence time is exponentially distributed with same rate
 - $A = \min(A_1, A_2)$ with CDF $A(t) = 1 (1 A_1(t)) \cdot (1 A_2(t)) = 1 e^{-(\lambda_1 + \lambda_2)t}$
 - $A = \min(A_1, A_2) \sim \text{EXP}(\lambda_1 + \lambda_2)$
- \blacktriangleright In general: superposition of n Poisson processes

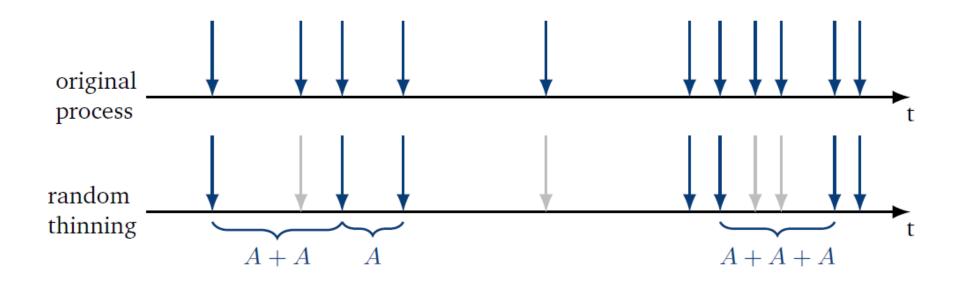
$$A \sim \text{EXP}(\sum_{i=1}^{n} \lambda_i)$$





Random Thinning of Poisson Processes

- \blacktriangleright Poisson process with rate λ
- \triangleright With probability p, an arrival event is considered in the thinned process
- With probability 1 p, an arrival event is not considered (gray arrows)



► Thinned process is a Poisson process with rate $p \cdot \lambda$

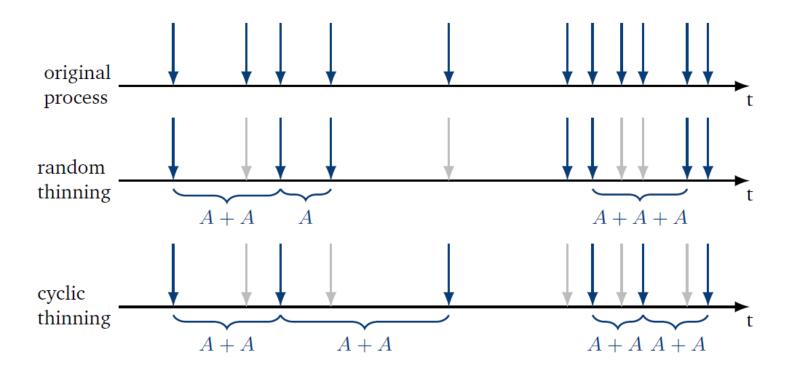
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Random Thinning of Poisson Processes



Cyclic Thinning of Poisson Processes

- ightharpoonup Poisson process with rate λ
- Cyclic or deterministic thinning
 - e.g., every second arrival is skipped
 - interarrival times follows an Erlang distribution, e.g., with k=2 and λ



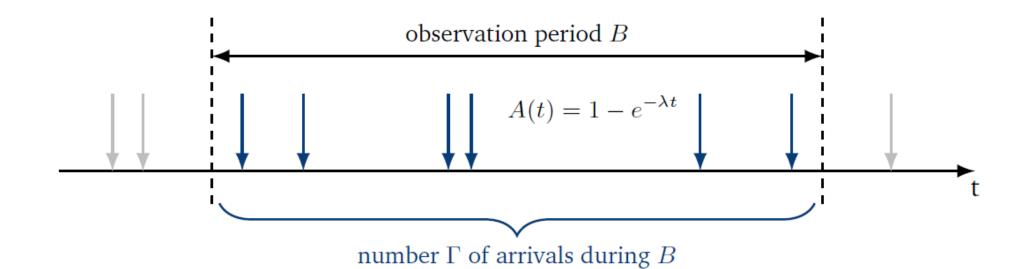
POISSON ARRIVALS DURING AN ARBITRARILY DISTRIBUTED INTERVAL





Arbitrarily Distributed Observation Interval

- ▶ Length B of observation window is a r.v. with Laplace transform $\Phi_B(s)$
- Number Γ of arrivals during B is of interest, e.g., analysis of M/GI/1 delay system



▶ **Generating function** of the number of arrivals $\Gamma_{GF}(z)$ depends on Laplace transform $\Phi_B(s)$

$$\Gamma_{GF}(z) = \Phi_B(\lambda(1-z))$$





