

CSC 2232: Topics in Computer System Performance and Reliability

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THE EXPONENTIAL DISTRIBUTION, THE POISSON PROCESS AND THE REAL WORLD

- What are the characteristics of the exponential distribution and the Poisson process?
- Why do people like it so much?
- What does the real world look like?

A QUICK REVIEW

- The cumulative distribution function (CDF)
 - $F(x)$ is the probability that $X \leq x$
 - $P(a < X \leq b) = F(b) - F(a)$
 - $\bar{F}(x) = 1 - F(x)$ is the probability that $X > x$
- The probability density function (PDF)
 - $f(x)$ is the derivative of $F(x)$
 - $f(x)dx$ is the probability that X is in $\{x, x+dx\}$
 - $P(a < X \leq b) = \int_a^b f(x)dx$
- The mean of X is:
 - $E[X] = \int_{-\infty}^{\infty} x f(x)dx$

THE EXPONENTIAL DISTRIBUTION

- If X is distributed exponentially with rate λ then

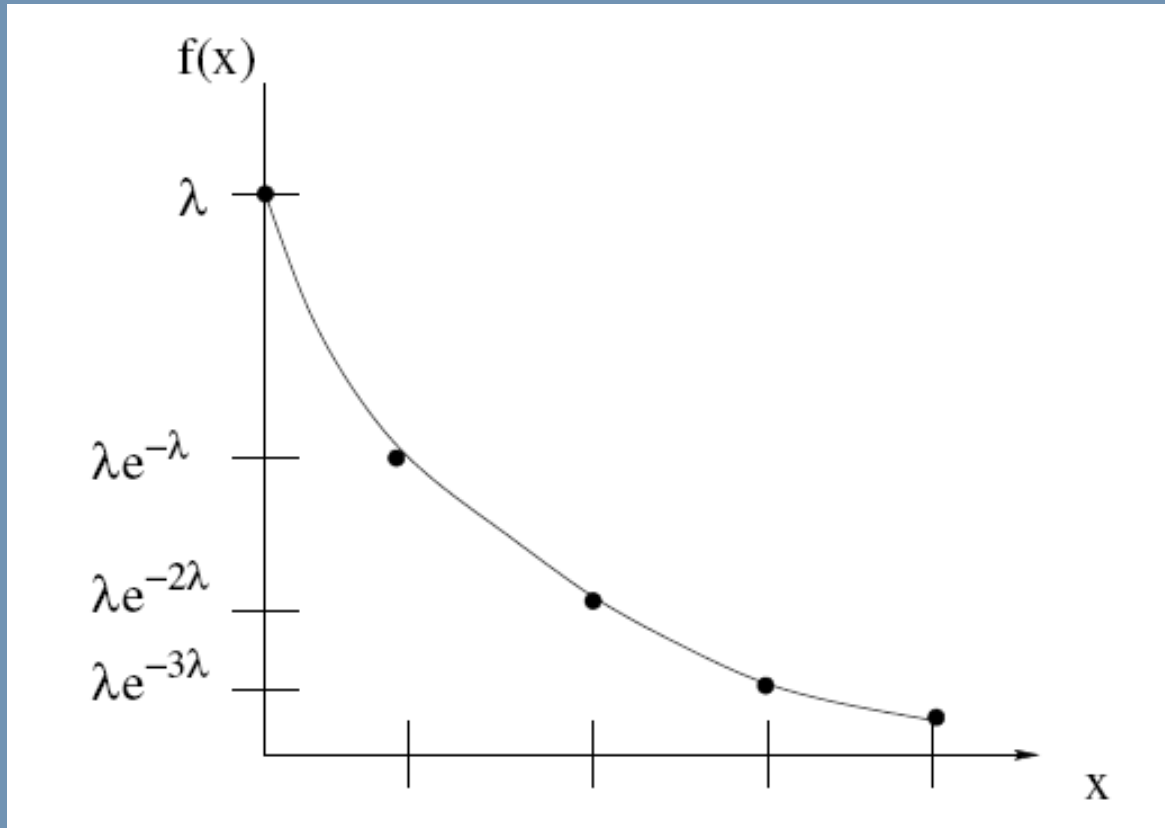
- $f(x) = \begin{cases} \lambda e^{-\lambda x} & \text{for } x \geq 0 \\ 0 & \text{for } x < 0 \end{cases}$

- $F(x) = \begin{cases} 1 - e^{-\lambda x} & \text{for } x \geq 0 \\ 0 & \text{for } x < 0 \end{cases}$

- $\bar{F}(x) = \begin{cases} e^{-\lambda x} & \text{for } x \geq 0 \\ 1 & \text{for } x < 0 \end{cases}$

THE EXPONENTIAL DISTRIBUTION

$$f(x) = \begin{cases} \lambda e^{-\lambda x} & \text{for } x \geq 0 \\ 0 & \text{for } x < 0 \end{cases} \quad \bar{F}(x) = \begin{cases} e^{-\lambda x} & \text{for } x \geq 0 \\ 1 & \text{for } x < 0 \end{cases}$$



THE MEAN AND VARIANCE OF THE EXPONENTIAL DISTRIBUTION

- $f(x) = \begin{cases} \lambda e^{-\lambda x} & \text{for } x \geq 0 \\ 0 & \text{for } x < 0 \end{cases}$

- $E[X] = \int_{-\infty}^{\infty} x f(x) dx = 1/\lambda$

- $\text{Var}(X) = 1/\lambda^2$ $(\text{Var}(X) = E[X^2] - (E[X])^2)$

- Squared coefficient of variation:

- $C.V.^2 = \text{variance}/\text{mean}^2 = 1$

THE MINIMUM OF TWO EXPONENTIALS

- Given independent variables $X_1 \sim \text{Exp}(\lambda_1)$ and $X_2 \sim \text{Exp}(\lambda_2)$
- Let $X = \min\{X_1, X_2\}$, then $X \sim \text{Exp}(\lambda_1 + \lambda_2)$

• Proof:

- $\Pr(X > t) = \Pr\{\min\{X_1, X_2\} > t\}$
- $= \Pr\{X_1 > t \text{ and } X_2 > t\}$
- $= \Pr\{X_1 > t\} * \Pr\{X_2 > t\}$
- $= e^{-\lambda_1 t} * e^{-\lambda_2 t}$
- $= e^{-(\lambda_1 + \lambda_2)t}$

THE MEMORYLESS PROPERTY

- A random variable X is said to be memoryless if

- $$\Pr\{X > s + t \mid X > s\} = \Pr\{X > t\}$$

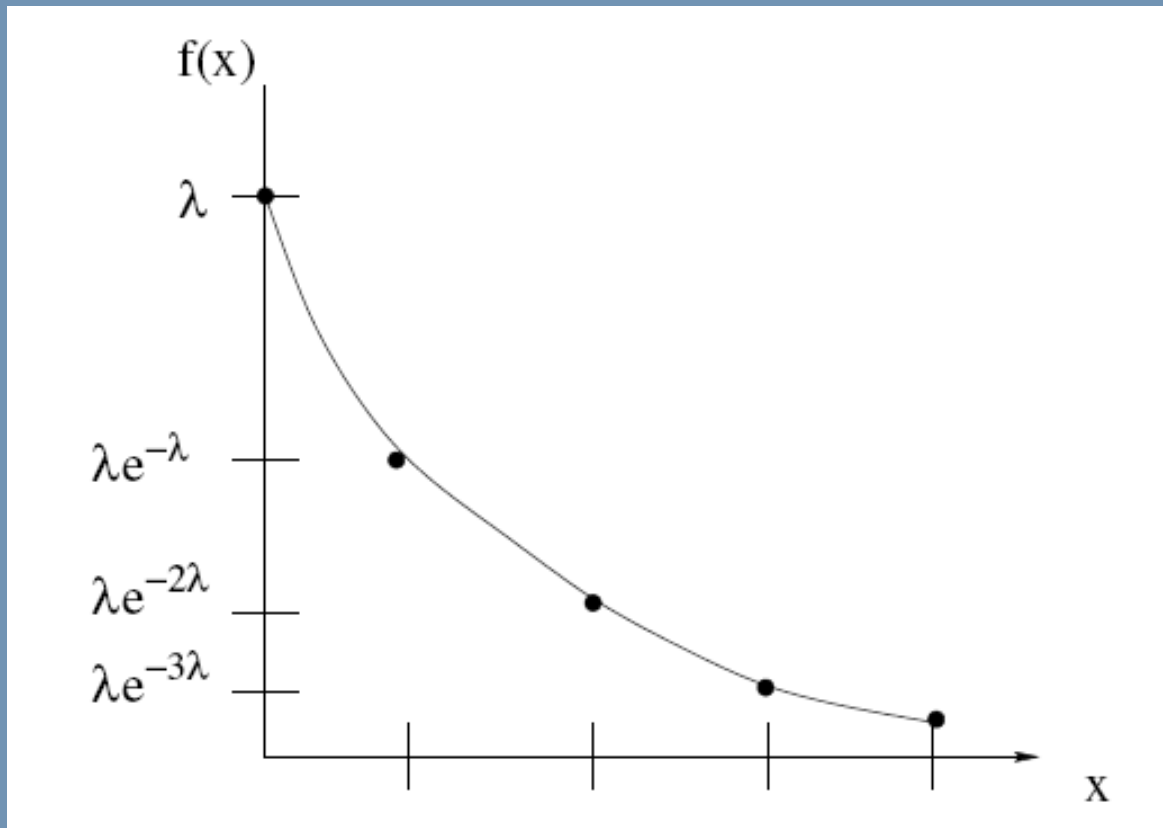
- Show that $X \sim \text{Exp}(\lambda)$ is memoryless

- $$\Pr\{X > s + t \mid X > s\} = \Pr\{X > s+t\} / \Pr\{X > s\}$$



$$\begin{aligned} &= e^{-\lambda(s+t)} / e^{-\lambda s} &&= e^{-\lambda t} \\ &&&= \Pr\{X > t\} \end{aligned}$$

THE EXPONENTIAL DISTRIBUTION

$$f(x) = \begin{cases} \lambda e^{-\lambda x} & \text{for } x \geq 0 \\ 0 & \text{for } x < 0 \end{cases} \quad \bar{F}(x) = \begin{cases} e^{-\lambda x} & \text{for } x \geq 0 \\ 1 & \text{for } x < 0 \end{cases}$$



THE MEMORYLESS PROPERTY

- Is the memoryless property realistic for lightbulbs?
- Real-life examples whose lifetime can be modeled by X such that $\Pr\{X > s+t \mid X > s\}$ goes down as s goes up?  Increasing failure rate
- Real-life examples whose lifetime can be modeled by X such that $\Pr\{X > s+t \mid X > s\}$ goes up as s goes up?  Decreasing failure rate
- Failure rate $r(t) := f(t) / (1-F(t))$

EXAMPLE 1

- The time a customer spends waiting in a bank is exponentially distributed with mean 10 min.
- What is $\Pr[\text{cust. waits} > 5 \text{ min}]$
 - $\bar{F}(5\text{min}) = e^{-5/10} = e^{-0.5}$
- $\Pr[\text{cust. waits} > 15 \text{ min} \mid \text{still in bank after 10 min}]$
 - $\bar{F}(5\text{min}) = e^{-5/10} = e^{-0.5}$

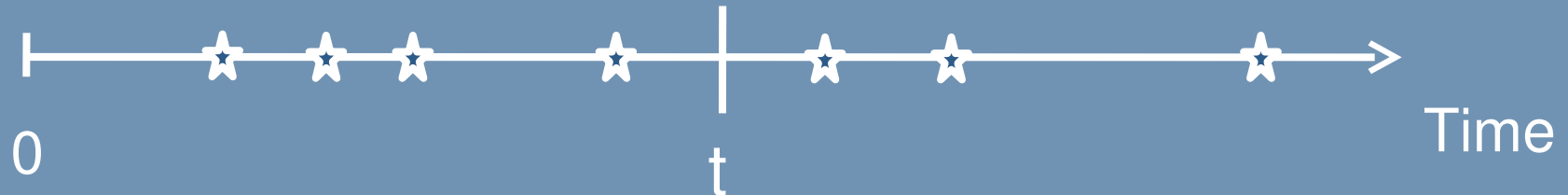
EXAMPLE 2

- A post office has 2 clerks. Cust. B is served by one clerk and cust. C is served by the other clerk, when A walks in. All service times are exponentially distributed.
- What is the probability that A is last to leave?
- $\frac{1}{2}$

NEXT:

- The Poisson process
- But need some definitions first.

DEFINITIONS



- $N(t) :=$ number of events that occurred by time t
- An event sequence has independent increments if the number of events at different time intervals are independent.
- I.e. for all $t_0 < t_1 < t_2 < \dots < t_n$ the rand. variables $N(t_1)-N(t_0)$, $N(t_2)-N(t_1)$, $N(t_n)-N(t_{n-1})$ are independent.

EXAMPLES

- Consider the following event processes:
 - People entering a store
 - Births of children
 - Goals scored by a particular soccer player?
- Do the above processes have independent increments?

DEFINITION 1 OF POISSON PROCESS

- A Poisson process having rate λ is a sequence of events such that:
 - $N(0) = 0$
 - The process has independent increments
 - The number of events in any interval of length t is Poisson distributed with mean λt , i.e.

$$\Pr\{N(t+s) - N(s) = n\} = e^{-\lambda t} \frac{(\lambda t)^n}{n!}$$

DEFINITION 2 OF POISSON PROCESS

- A Poisson process with rate λ is a sequence of events such that the inter-arrival times are i.i.d. exponential random variables with rate λ and $N(0)=0$.

MERGING POISSON PROCESSES

 Process 1 with rate λ_1

A horizontal timeline with an arrow pointing right. It contains seven yellow stars. The first two stars are close together, followed by a gap, then another star, followed by a gap, and finally three stars close together.

 Process 2 with rate λ_2

A horizontal timeline with an arrow pointing right. It contains three orange stars. The first star is at the beginning, followed by a large gap, then a second star, followed by another large gap, and finally a third star at the end.

 Poisson with rate $(\lambda_1 + \lambda_2)$

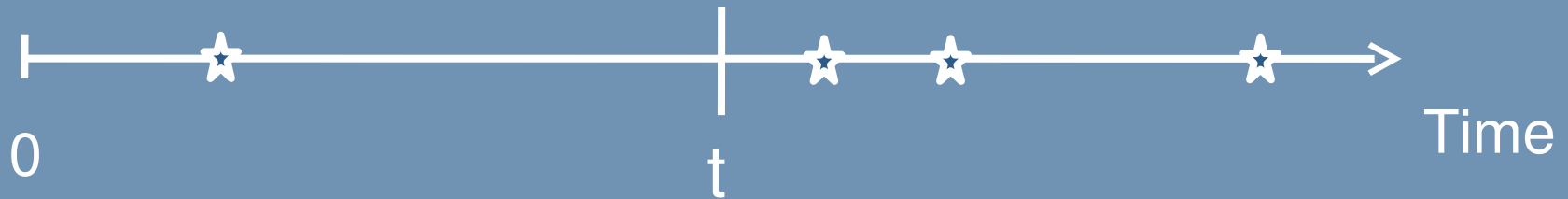
A horizontal timeline with an arrow pointing right. It contains ten stars in total, alternating in color: orange, yellow, yellow, orange, yellow, yellow, yellow, yellow, orange, orange.

- Theorem: The merge of process 1 and process 2 is a single Poisson process with rate $(\lambda_1 + \lambda_2)$
- Proof:
 - Process 1 has $\text{Exp}(\lambda_1)$ inter-arrival times
 - Process 2 has $\text{Exp}(\lambda_2)$ inter-arrival times
 - Time until first event has $\text{Exp}(\lambda_1 + \lambda_2)$

POISSON SPLITTING

- Given Poisson process with rate λ , suppose each event is randomly classified as type A with prob. p and type B with prob $1-p$.
- Then type A events form a Poisson process with rate $p * \lambda$
- And type B events form a Poisson process with rate $(1-p) * \lambda$

UNIFORMITY

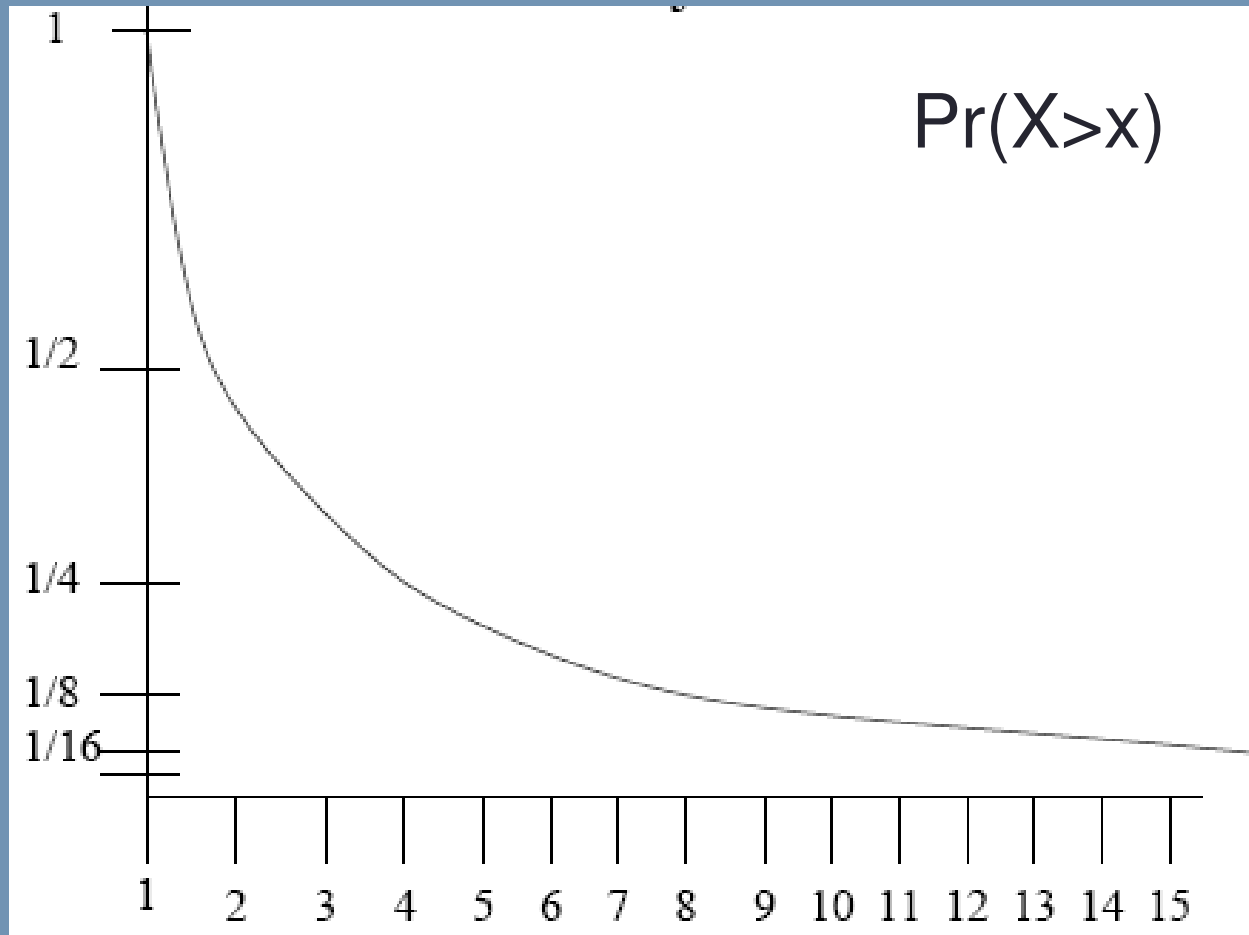


- Given that one event of a Poisson process has occurred by time t , that event is equally likely to have occurred anywhere in $[0, t]$
- Generalization:
- If k events of a Poisson process occur by time t , the k events are distributed independently and uniformly in $[0, t]$.

SUMMARY OF EXPONENTIAL DISTRIBUTION & POISSON PROCESS

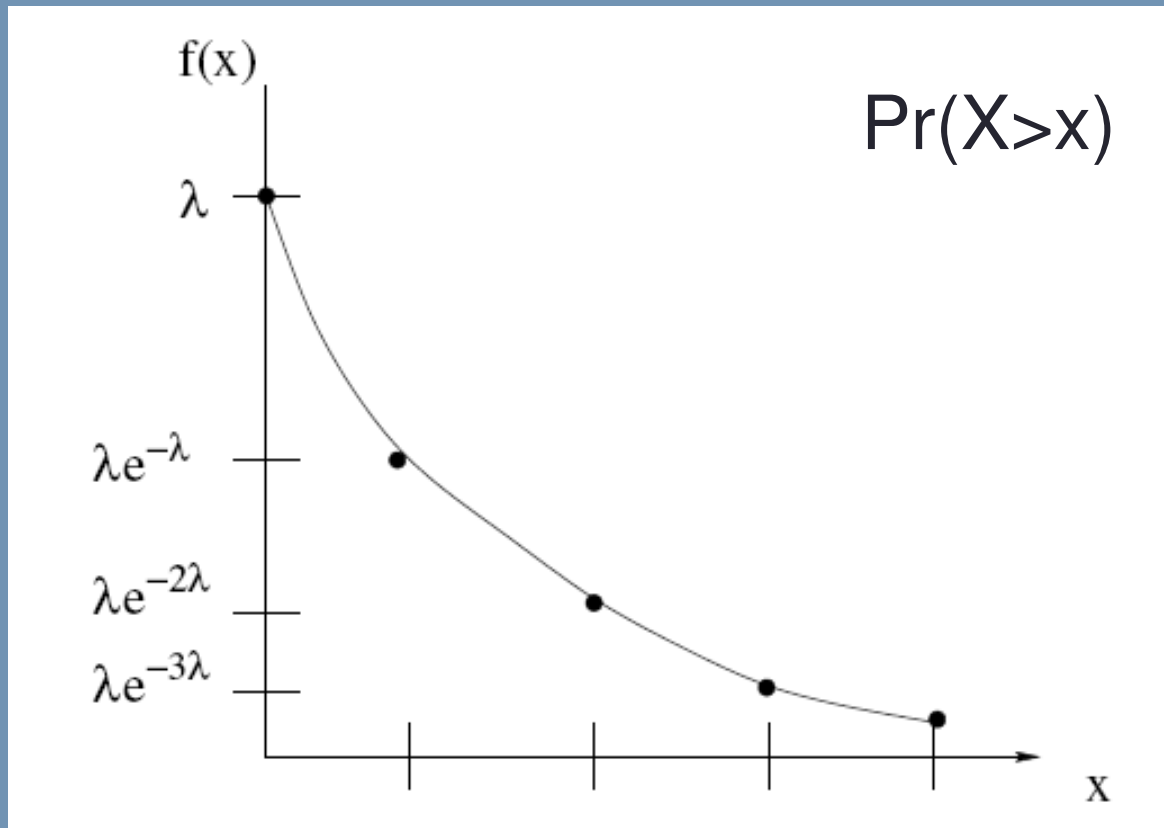
- The exponential distribution is memoryless
 - History doesn't matter!
- The Poisson process is an event sequence such that inter-arrival times are iid exponential rand. vars.
 - Poisson process can be merged
 - Poisson processes can be split
 - Uniformity of events

A COMMON REAL-WORLD DISTRIBUTION

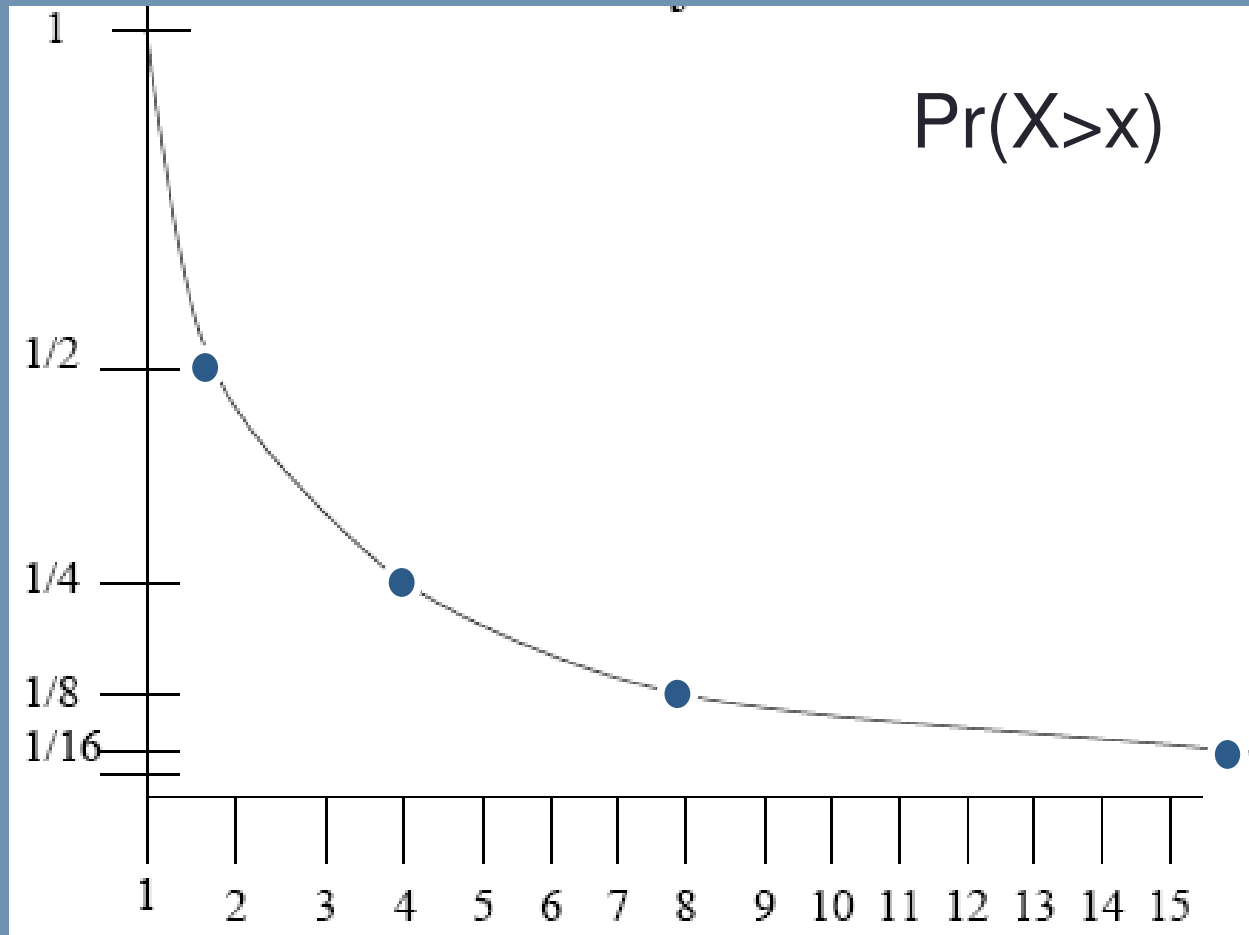


THE EXPONENTIAL DISTRIBUTION

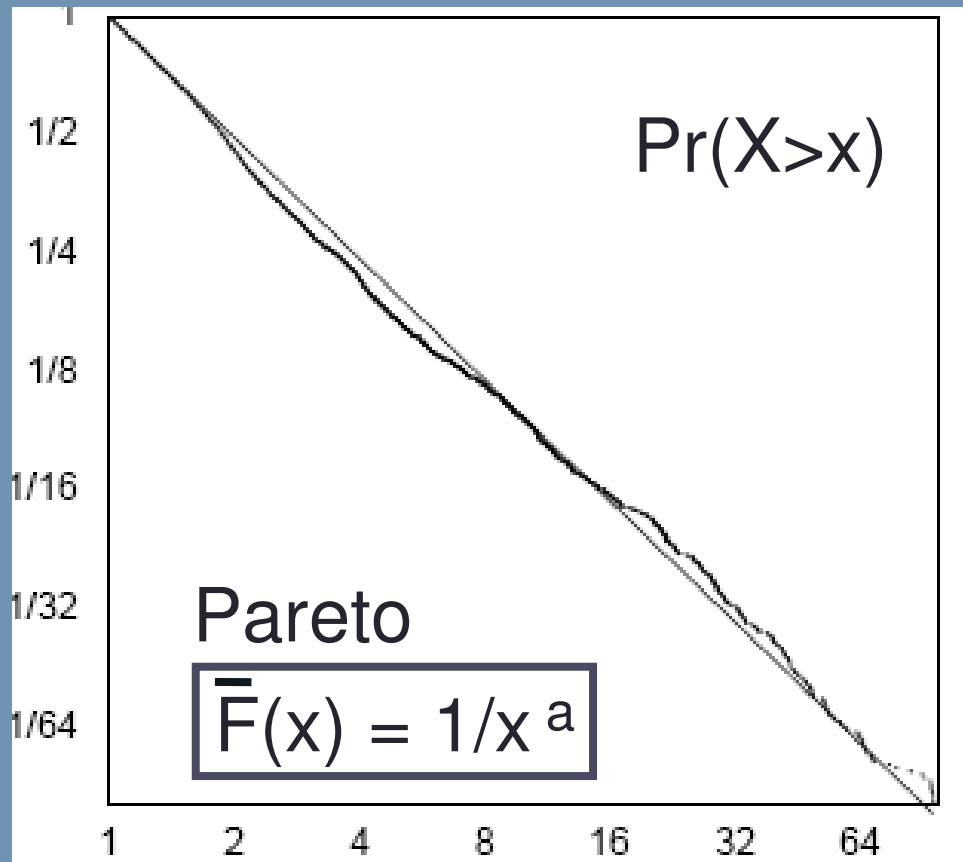
- $\overline{F}(x) = \begin{cases} e^{-\lambda x} & \text{for } x \geq 0 \\ 0 & \text{for } x < 0 \end{cases}$



A COMMON REAL-WORLD DISTRIBUTION

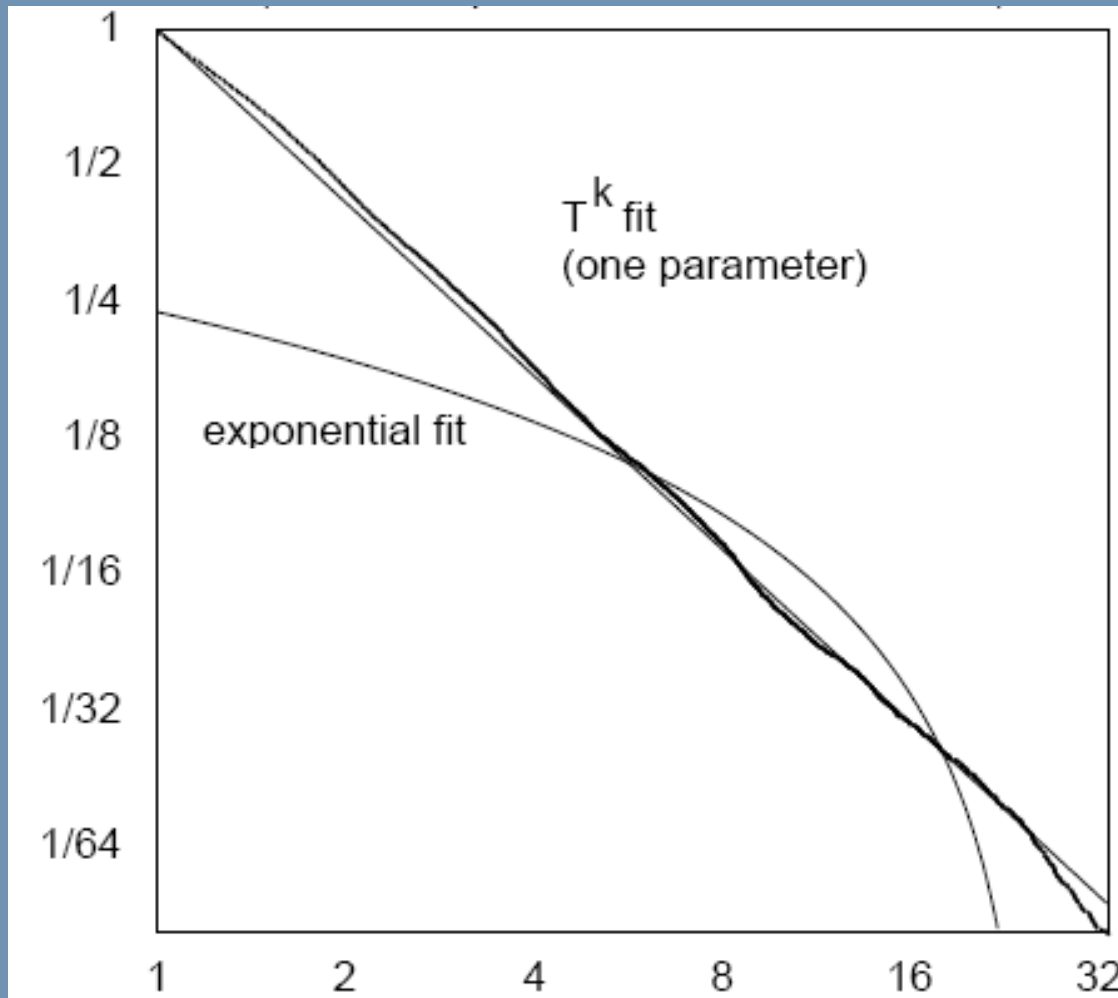


A COMMON REAL-WORLD DISTRIBUTION



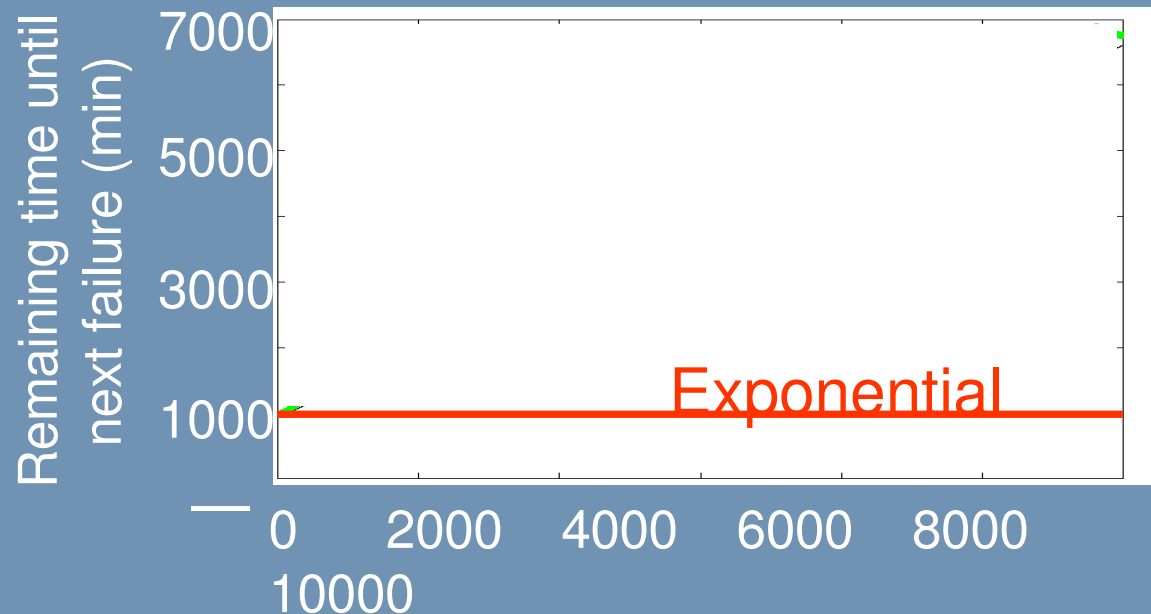
- a is commonly ~ 1

COMPARISON WITH EXPONENTIAL



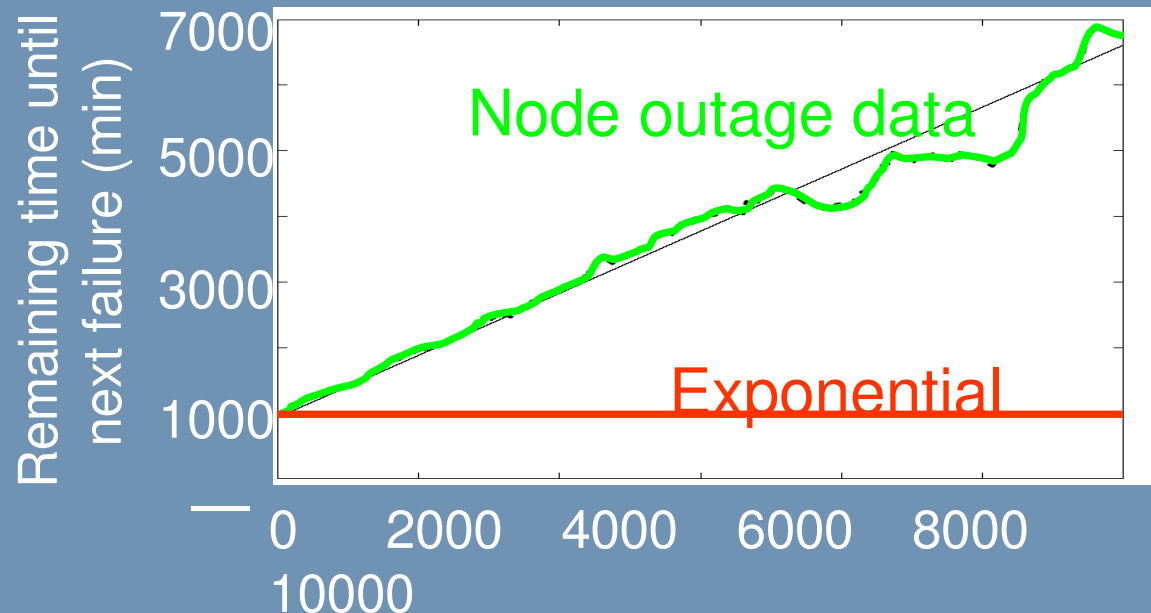
DECREASING HAZARD RATES

- *Example:* Time between failure of nodes in a cluster



DECREASING HAZARD RATES

- *Example:* Time between failure of nodes in a cluster



PROPERTY OF PARETO DISTRIBUTIONS

- Unbounded mean and variance ($a \leq 1$)
 - $E[X] = \infty$
 - $\text{Var}(X) = \infty$
 - \Rightarrow In practice, use bounded Pareto
- Decreasing failure rate
- Heavy-tail property

HEAVY-TAIL PROPERTY

- Miniscule fraction of the very largest jobs comprises half the load
 - E.g. if $a=1.1$ then the top 1% largest jobs make up 50% of the load
 - In comparison: in an exponential the top 1% jobs make up 5% of the load

REAL-WORLD EXAMPLES OF HEAVY-TAIL DISTRIBUTIONS

- Distribution of wealth in the population
- Number of visits to a web site
- Number of pages within a site
- Number of links to a web page
- Length of TCP flows
- File size distributions
- Job runtime distributions