Performance Modeling of Computer Systems and Networks

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Memoryless property and probability distributions

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Memoryless property as lifetime

Analytical models Memoryless property

A random variable X is said to be memoryless if

$$Prob(X > s + t|X > s) = Prob(X > t) \quad \forall s, t > 0$$

Example

 \overline{X} is the lifetime of a lightbulb.

The property says that the probability that the lightbulb survives for at least another t seconds before burning out, given that the lightbulb has survived for s seconds already, is the same as the probability that the lightbulb survives at least t seconds independent of s.

Does this seem realistic for a lightbulb???

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Lifetime and failure rate

Analytical models Failure rate

Distributions for which $Prob\{X > S + t \mid X > S\}$ goes down as s goes up are said to have *increasing failure rate*.

The device is more and more likely to fail as time goes on.

Example

A car's lifetime. The older a car is the less likely that it will survive another t = 6 years.

Distributions for which Prob(X > s + t(X > s)) goes up as s goes up are said to have decreasing failure rate.

The device is less likely to fail as time goes on.

Example

- UNIX job CPU lifetimes. The more CPU a job has used up so far, the more it is likely to use up.
- The same for computer chips. If they're going to fail, they'll do it early. That's why chip manufacturers test them for a long while.

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Analytical models Failure rate

Hazard rate function

Let X be a continuous random variable with probability density function f(t) and cumulative distribution function $F(t) = Pr\{X < t\}$.

Then r(t) is formally defined as:

$$r(t) = \frac{f(t)}{F(t)}$$

where $F(t) = 1 - F(t) = Pr\{X > t\}$

Consider the probability that a t-year old item will fail during the next dt seconds:

$$Pr\{X \in (t,t+dt) | X > t\} = \frac{Pr\{X \in (t,t+dt)\}}{Pr\{X > t\}} \approx \frac{f(t)dt}{F(t)} = r(t)dt$$

the istantaneous failure rate

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Analytical models Failure rate

Hazard rate function

If r(t) is constant then f(t) must be exponential Indeed for the exponential

$$r(t) = \frac{f(t)}{F(t)} = \frac{\lambda e^{-\lambda t}}{e^{-\lambda t}} = \lambda$$

We use the failure rate concept when we study scheduling

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Analytical models Failure rate

Why the remaining lifetime is so important?

CPU load balancing in a Network of Workstations

- It may help to migrate a job to a <u>less-loaded</u> workstation (one with fewer jobs) in order to improve mean response times
- migration can be expensive if the job has <u>a lot of "state"</u> that has to be migrated with the job (lots of memory).

two types of migration used in load balancing techniques:

- non-preemptive migration (NP) only relocates "newborn" processes (initial placement, or remote execution)
- preemptive migration (P) migrates processes that are already active (running)

 (active process migration)

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Analytical models

- 1. Should we bother with P migration, or is NP enough?
- 2. If we are going to bother with P migration, which processes are worth migrating?
 That is, what is a good migration policy?

terminology:

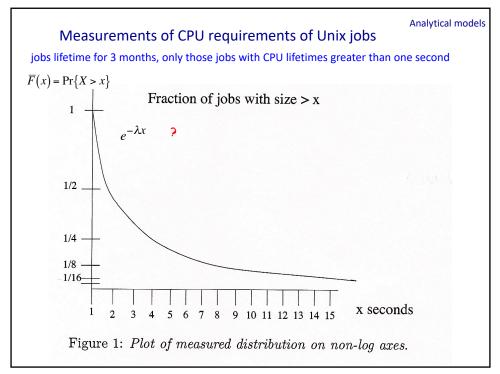
- a job's "size": its total CPU demand
- a job's "age": its total CPU usage thus far
- a job's "lifetime" refers to its total CPU requirement
- a job's "remaining lifetime" refers to its remaining CPU requirement

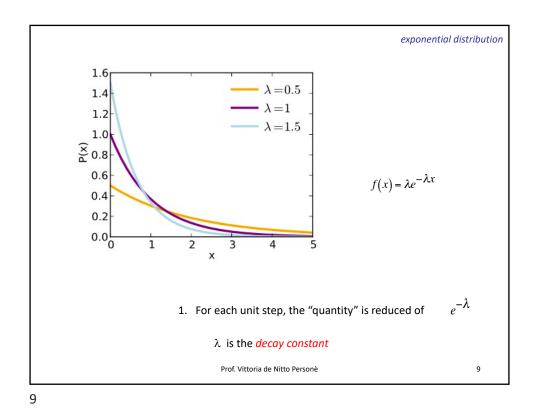
Observe that commonly, at any point in time, you don't know the job's remaining lifetime, just it's current CPU age.

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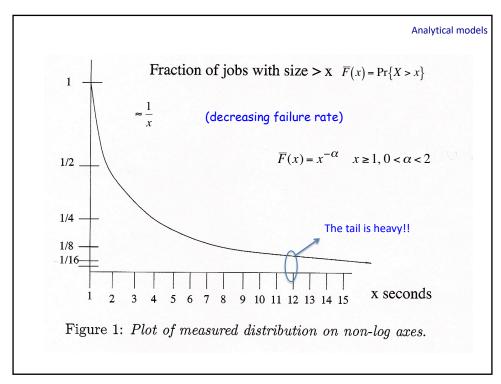
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Measurements of CPU requirements of Unix jobs jobs lifetime for 3 months, only those jobs with CPU lifetimes greater than one second $\overline{F}(x) = \Pr\{X > x\}$ Fraction of jobs with size > x $e^{-\lambda x}$ No! With each unit increase in x, the fraction should drop by a constant factor $e^{-\lambda}$ (constant failure rate) 1/2 1/4 1/8 1/16 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 x seconds Figure 1: Plot of measured distribution on non-log axes.



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Analytical models Heavy tail distributions

Pareto distributions

$$f(x) = \alpha k^{\alpha} x^{-\alpha - 1}$$
 $k \le x < \infty, 0 < \alpha < 2$

lpha a measure of the distribution variability and of the "heavy-tailedness":

 $\alpha \rightarrow 0$ +variability, +heavy

 $\alpha \rightarrow 2$ -variability, -heavy

Problem: i-th moment is finite just for $\alpha > i$

$$E[X] = \frac{\alpha k}{\alpha - 1} \qquad \alpha > 1$$

$$var[X] = \frac{\alpha k^2}{(\alpha - 1)^2 (\alpha - 2)} \quad \alpha > 2$$

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Analytical models Heavy tail distributions

Properties of Pareto distributions

Decreasing Failure Rate

The more cpu you have used so far, the more you will continue to use

completely different from the exponential distribution, where your cpu usage after any point in time is completely independent of the amount of cpu used up to that point (memoryless property)

Infinite Variance

"Heavy-Tail Property"

A miniscule fraction of the very largest jobs comprise half of the load on the system.

For example, when α = 1.1, the largest 1% of the jobs comprise 1/2 of the load

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Analytical models Heavy tail distributions

Bounded Pareto distributions

the measured data has a *minimum* job lifetime and a *maximum* job lifetime.

Thus the measured data has all finite moments.

To model the measured data, we therefore want a Pareto distribution which has been truncated.

$$f(x) = \alpha x^{-\alpha - 1} \frac{k^{\alpha}}{1 - \left(\frac{k}{p}\right)^{\alpha}} \quad k \le x \le p, \ 0 < \alpha < 2$$

all of the moments are finite

The actual measured squared coefficient of variation values were (obviously) finite and were between 25 and 49!!!

$$\mathbf{C}^2 = \frac{var}{mean^2}$$
, expo=1

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Analytical models

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the DFR property leads us to think that it may pay to migrate old jobs. The reasoning is that although an old job may have high migration cost, because it has accumulated a lot of memory, if the job is really old then it has a high probability of using a lot more cpu in the future, which means that the cost of migration can be amortized over a very long lifetime.

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