



## Optimal scheduling discipline in a single-server queue with Pareto type service times

Samuli Aalto (Helsinki University of Technology) Urtzi Ayesta (LAAS-CNRS)

> ValueTools 2008 23 october, 2008

## Scheduling in an M/G/1 Queue



- Poisson arrivals with rate  $\lambda$ .

Service requirements are i.i.d. with distribution  $F(x)=P[X \le x]$ . Complementary cumulative distribution denoted by  $\overline{F}(x)=1-F(x)$ 

- Attained service is known (total service requirement unknown)
- Optimality criterion: Mean number of jobs in the system

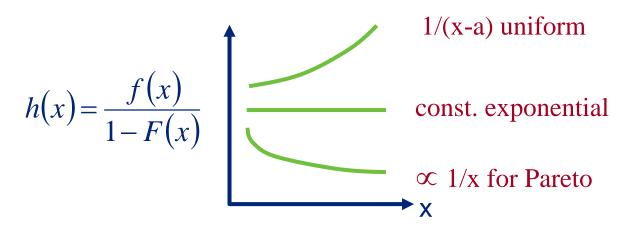
# **Scheduling disciplines**

Two important set of disciplines depending on whether the size of jobs is known.

- The size is known: Shortest-Remaining-Processing-Time (SRPT) is optimal with respect to the average response time of the system.
- The size is not known, but we know the *attained service* of jobs. The most appropriate scheduling discipline depends on the service time distribution characteristics

## **Monotone Hazard Rate**

Hazard rate of a distribution function:  $h(x)dx=P[x < X \le x+dx | X > x]$ 



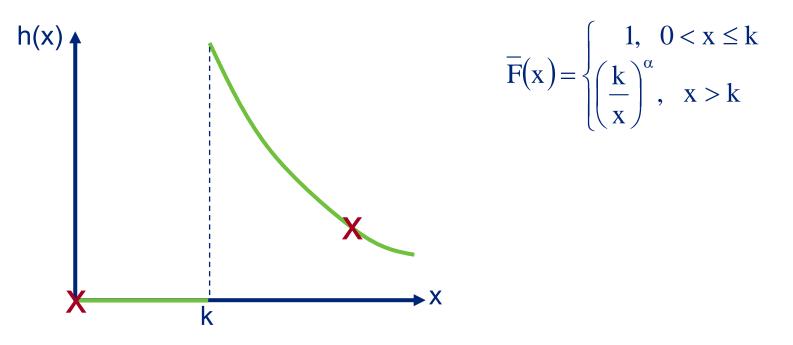
- Increasing Hazard Rate (IHR):

Non-preemptive discipline (FCFS etc.) is optimal

- Constant hazard rate, i.e. exponential distribution:
  Mean number of jobs is policy independent
  - Decreasing Hazard Rate (DHR):
    Least Attained Service (LAS) is optimal. Job(s) with the least attained service is served.

## Scheduling for non-monotone hazard rate?

What if the distribution is defined on a interval [k,∞).
 For example a Pareto-type distribution



- What if the support is bounded, that is, if  $\overline{F}(x) = 0$  for all p>x ?

# **Optimality of Gittins policy**

Theorem [Gittins89]:

Gittins index policy minimizes the mean number of jobs in the system among all non-anticipating scheduling policies

Introduced by Sevcik [1974] for static scheduling (Smallest-Rank policy) Optimality in an M/G/1 queue by Gittins [1989].

## **Gittins index**

Job with attained service **a** has the Gittins index  $G(a) = \sup_{\Delta \ge 0} J(a, \Delta)$ with  $J(a, \Delta) = \frac{a}{a + \Delta} \int_{a}^{a+\Delta} F(y) dy = \frac{\text{reward}}{\text{investment}}$ 

- reward:  $P[a \le X \le a + \Delta | X > a]$
- investment: E[min(X-a, ∆) |X>a]
- In particular: J(a,0)=h(a) and  $J(a,\infty)=1/E[X-a|X>a]$

Gittins index policy

Serve at every instant of time the job with highest value G(a).

## **Relation between Gittins and SR**

Gittins index policy

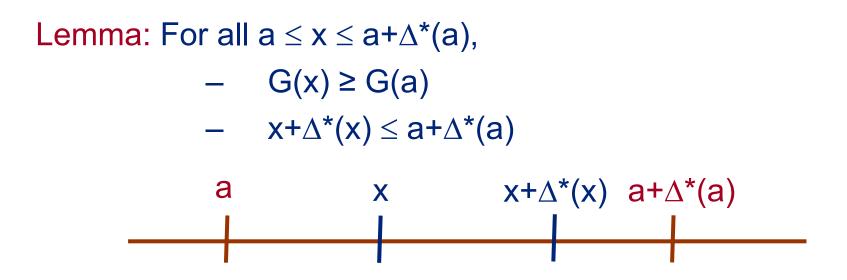
Serve at every instant of time the job with highest value G(a).

Sevcik's Smallest-Rank policy index policy: Pick the job with highest index value G(a) and assign him a service quota  $\Delta^*(a) = \inf \{\Delta \ge 0 \mid G(a) = J(a, \Delta)\}$ 

This job will be served until:

- It receives  $\Delta^*(a)$  units of service
- It departs from the system
- A new job with higher Gittins index arrives to the queue

# Gittins and SR (cont.)



Proposition: The Gittins discipline and SR are equivalent sample-path wise.

## - Not surprising result:

- Optimality of cµ-rule (without arrivals Smith'56, with arrivals Fife'65)
- Multi-class single server queue with feedback and non-preemptive policy: The optimal policy without arrivals is also optimal with Poisson arrivals [MW76]

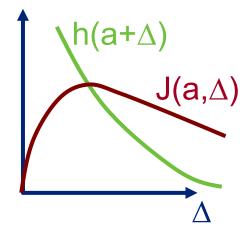
## **Gittins index policy**

**Theorem:** For any attained service  $a \ge 0$ ,

$$G(a) = h(a + \Delta^*(a))$$

Sketch of the proof:

$$\frac{\partial}{\partial \Delta} J(a, \Delta) = 0 \Longrightarrow J(a, \Delta^*(a)) = h(a + \Delta^*(a))$$



Proposition: G(a) is decreasing for all a if and only if the service time distribution is DHR

Sketch of the proof:

For any fixed a,  $J(a,\Delta)$  is decreasing with respect to  $\Delta$ .

- Then for all a, G(a)=J(a,0)=h(a), and note that h(a) is decreasing
- $\rightarrow$  It can be shown that G(a)=h(a)

Theorem: LAS minimizes stochastically the number of jobs if and only if the service time distribution is DHR Sketch of the proof:

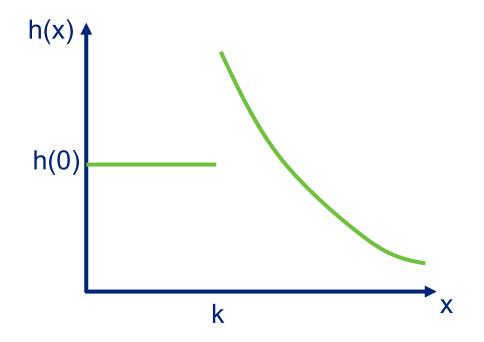
- ← By [RS,1989]
- → in particular LAS minimizes the mean, thus G(a) is decreasing, and hence distribution must be DHR.

Equivalent result for FCFS and NBUE distributions.

## **CDHR(k) or Pareto-type distributions**

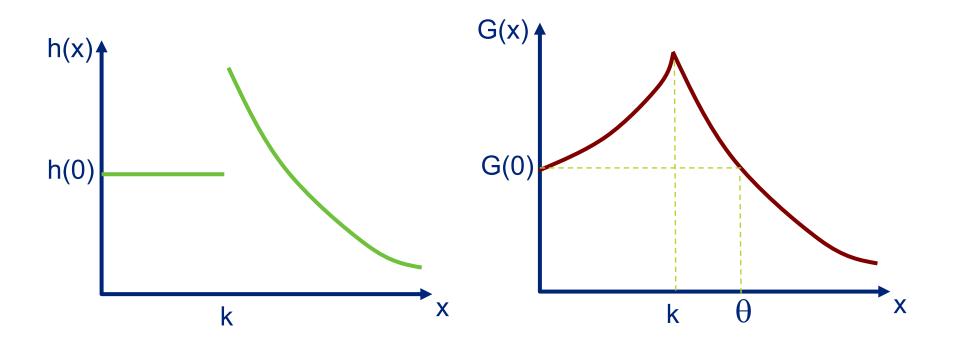
Definition of CDHR(k) distribution:

- A1: h(x) is constant for all x < k,
- A2: h(x) is decreasing for all  $x \ge k$ .
- A3: h(0) < h(k).



**Proposition:** Assume the service time distribution belongs to the class CDHR(k). Then there exists a  $\theta > k$  s.t,

- $G(x) \ge G(0)$  for all  $x < \theta$ ,
- $G(\theta) \leq G(0)$ , and
- G(x) is decreasing for all  $x \ge \theta$ .



Theorem: Assume a CDHR(k) service time distribution:

- (i) If A3 is not satisfied, then G(x) is decreasing for all x, hence LAS is optimal.
- (ii) If A3 is satisfied, then there is  $\theta > k$  such that FCFS+ LAS( $\theta$ ) is optimal.
  - The precise value of  $\theta$  depends only on the parameters of the service time distribution.

#### $FCFS+LAS(\theta)$ :

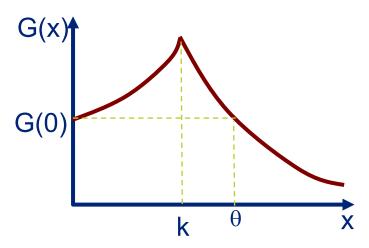
Classify jobs into two classes:

## **High Priority:**

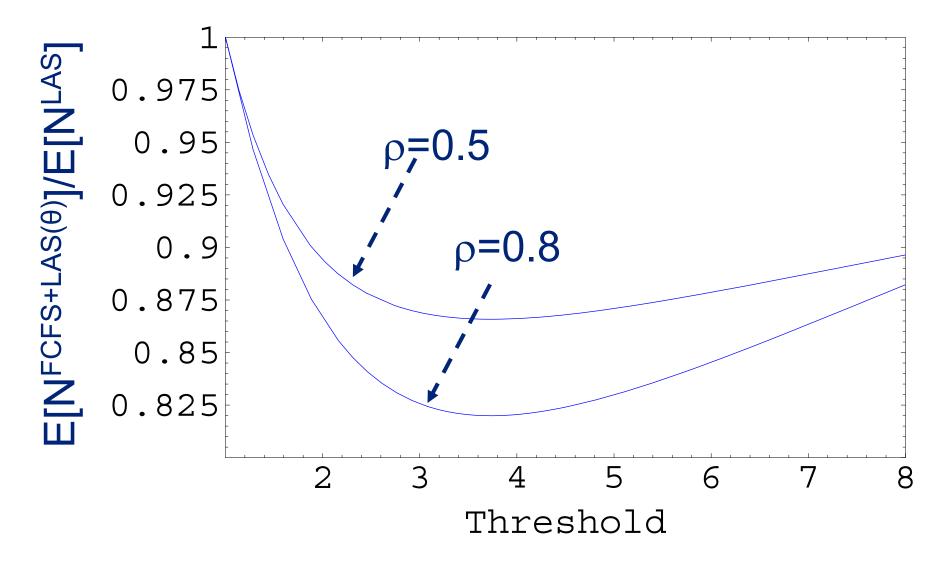
- Jobs with attained service less than  $\theta$
- Serve within this class according to FCFS

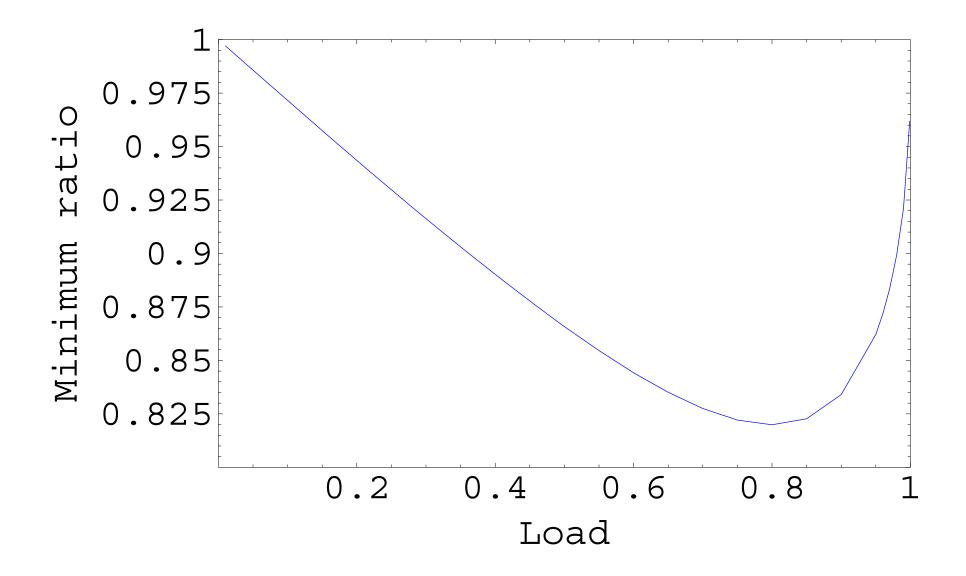
### Low Priority:

- Jobs with attained service more than  $\theta$
- Serve within this class according to LAS

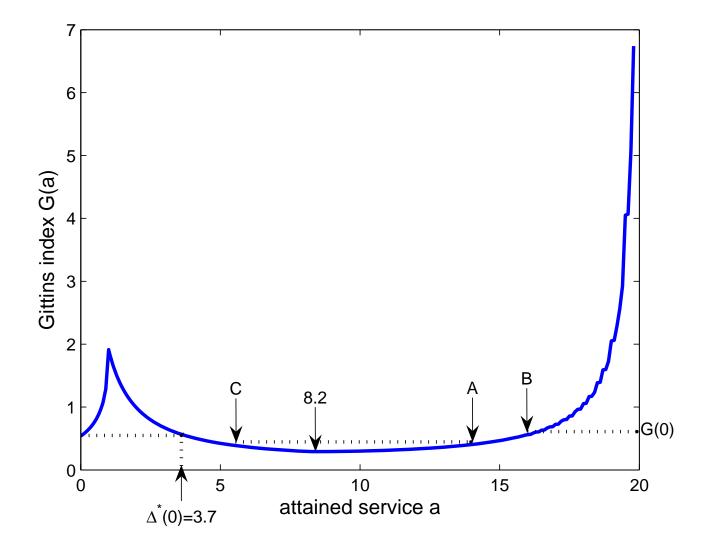


Numerical example: Pareto distribution with k=1 and  $\alpha$  =2





# Impact of an upper bound bounded distribution: Bounded Pareto

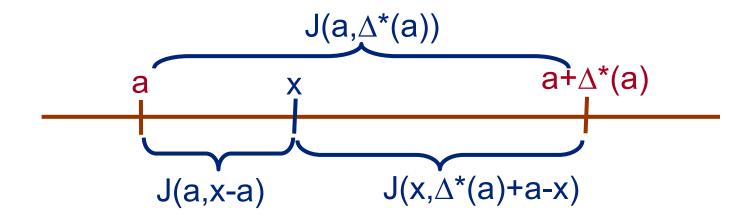


## **Conclusion and future research**

- Application of index policy for non work-conserving systems:
  - Multi-server systems
  - Time-varying capacity like in wireless systems
- Scheduling in a G/G/1 queue. LAS and FCFS are optimal with DHR and IHR respectively.
  - What if hazard-rate is not monotone?
- Calculate performance metrics for a given function G(a)?
- Relation between optimal scheduling in static and stochastic scenarios.
- Application of Gittins for multi-class queues
  - Optimal policy for cases that  $c\mu$ -rule does not cover

Sketch of the proof: For all  $a \le x \le a + \Delta^*(a)$ , there exists a function  $p(x) \le 1$  such that

 $J(a,\Delta^{*}(a))=p(x) J(a,x-a) + (1-p(x)) J(x,\Delta^{*}(a)+a-x).$ 



But  $J(a,\Delta^*(a)) \ge J(a,x-a)$ , thus  $J(x,\Delta^*(a)+a-x) \ge J(a,\Delta^*(a))$ .

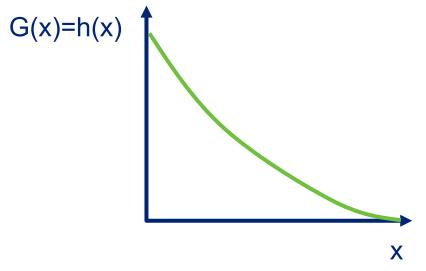
Now it follows that

 $G(x) \geq J(x, \Delta^*(a) + a - x) \geq J(a, \Delta^*(a)) = G(a).$ 

If the distribution is of type DHR, LAS (Least-Theorem: Attained Service) minimizes the mean number of jobs in the system  $J(a, \Delta) = \frac{\int f(y) dy}{\int \frac{a}{a + \Delta}}$ 

Sketch of the proof:

- For any fixed a,  $J(a,\Delta)$  is decreasing with respect to  $\Delta$ .
- Then for all a, G(a)=J(a,0)=h(a), and note that h(a) is decreasing



Similar result for IHR, then  $G(a) \ge G(0)$ , for all a Hence any non-preemptive policy is optimal