

An Optimal Service Ordering for a World Wide Web Server

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Overview

- Motivation
- System model
- Derivation of an optimal policy
- Two extensions
- A counterexample
- Conclusions





- Key motivation is the existence of an *impatient user population*
 - processor sharing is not effective if many users time out
 - is fairness really a good idea?
- Key metric is user-perceived performance, measured as a function of user response time at the web server



The model

- Web server modeled as a single-server queue
 - service times $\sim \exp(\mu)$, i.i.d, proportional to sizes of requested files
 - Request arrivals ~ Poisson(λ)
- Network delays not considered here
- Switching time between requests is negligible



- User-perceived performance ≡ 1 P{user aborts the download before the file is completely received}
 - decreasing function of response time
 - $\equiv revenue$
- Server is unaware if/when requests are aborted
- Server alternates between busy cycles, when there is at least one request in the system, and idle cycles (state 0), when there are no requests waiting or in service. These cycles are i.i.d.



• **Revenue** earned by the server under policy *p* from serving request *i* to completion (collected upon departure):

$$g_{ip}(t_{ip}) = e^{-c(t_{ip} - \bar{x}_i)}$$

- Potential revenue: $g_{ip}(t)$
- By the Renewal Reward Theorem, the average reward earned by the server per unit time under policy *p* is

$$V_p(0) = \frac{E\left[\sum_{i=0}^{N} g_{ip}(t_{ip})\right]}{E[Z_1]}$$

Objective

The optimal policy is the service policy that maximizes average revenue earned per unit time; it satisfies

$$V = \max_{p} V_{p}(0)$$



Derivation of the optimal policy

The optimal policy satisfies the following:

- non-idling
- switches between jobs in service only at arrival/departure epochs
- 3. non-processor-sharing
- 4. Markov (independent of both past and future arrivals)



Proof of (2)

If
$$g_{(k+1)}(a_1) \ge g_k(a_1)$$

use p'

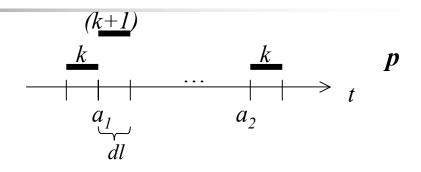
Revenue difference =

$$\mu d\ell e^{-c(a_1-x_{(k+1)})}(1-e^{-cd\ell})$$

Else use p"

Revenue difference =

$$\mu d\ell e^{-cd\ell} (e^{-ca_1} - e^{-ca_2}) (e^{cx_k} - e^{cx_{(k+1)}})$$



(k+1)

Proof of (3)

Consider a portion of the sample path over which the server processes two jobs to completion

Three possibilities:

- serve 1 then 2: expected revenue is $\frac{\mu}{c+\mu}c_1 + \left(\frac{\mu}{c+\mu}\right)^2 c_2$ (1)
- serve 2 then 1: expected revenue is $\frac{\mu}{c+\mu}c_2 + \left(\frac{\mu}{c+\mu}\right)^2 c_1 \quad (2)$
- serve both until one completes, then serve the other: expected revenue is

$$q(1)+(1-q)(2)$$

Definition of an optimal policy

State vector at time t under policy p:

$$G_{tp} = \{g_{1p}(t) \quad g_{2p}(t) \quad \dots \quad g_{Mp}(t)\}$$

M = number of requests that have not departed prior to t

best job:
$$\arg \max_{i} g_{ip}(t), i \in 1, 2, ... M$$



The optimal policy chooses the *best job* to serve at any time *t* regardless of the order of service chosen for the rest of the jobs in the system (greedy)

⇒ LIFO-PR



Two extensions

- Varying initial reward
 - applicable to different "classes" of users, QoS levels, etc.
- File sizes independently drawn from a set of exponential distributions
 - applicable to hosting several types of content at a server, each described by a different exponential distribution

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Extension 1

- *Initial reward* $\equiv C_i$
- Potential revenue:

$$g_{ip}(t) = C_i e^{-c(t-x_i)}$$

Natural extension of the original system model

Optimal policy: serve the request that has the highest potential revenue value at any time *t*

(function of the time in the system so far and the initial reward)



- Service time of request $i \sim \exp(\mu_i)$, independent
- Additional initial assumptions:
 - server only switches between jobs in service upon an arrival to or departure from the system
 - non-processor-sharing



Extension 2

We show that the optimal policy is

- non-idling
- Markov (independent of past and future arrivals)

Extension 2

State vector at time *t* under policy *p*:

$$G_{tp} = \{ (g_{1p}(t), \mu_1) \mid (g_{2p}(t), \mu_2) \dots (g_{Mp}(t), \mu_M) \}$$

Best job satisfies

$$\arg\max_{i} g_{ip}(t) \mu_{i}, i \in 1, 2, ...M$$

Optimal policy:

Serve the request with the highest $g_{ip}(t) \mu_i$ product



- Q: What happens if we remove the two additional initial assumptions in Extension 2?
 - specifically, what happens if we throw processorsharing policies back into the pool of possible policies?

Consider a sample path over which the server processes two jobs to completion. Job 1 arrives at time 0, job 2 arrives at time *a*.

Three possibilities from a until the end of the busy cycle:

- NPS1: serve 1 then 2
- NPS2: preempt 1, serve 2, serve 1
- PS: upon 2's arrival, serve both until one departs, then serve the other



Expected revenues for NPS1, NPS2:

$$E(g_{NPS1}) = c_1 \frac{\mu_1}{c + \mu_1} + c_2 \frac{\mu_1}{c + \mu_1} \frac{\mu_2}{c + \mu_2}$$

$$E(g_{NPS2}) = c_1 \frac{\mu_1}{c + \mu_1} \frac{\mu_2}{c + \mu_2} + c_2 \frac{\mu_2}{c + \mu_2}$$



However,

$$E(g_{PS}) \neq qE(g_{NPS1}) + (1-q)E(g_{NPS2})$$

There exists at least one numerical example in which

$$E(g_{PS}) > qE(g_{NPS1}) + (1-q)E(g_{NPS2})$$



A: We cannot clearly state that NPS policies are always better than PS policies. In fact, for at least one sample path, PS slightly outperforms both NPS alternatives.

(Note: We have not yet been able to reproduce this behavior in larger sample path simulations.)



Conclusions

- When network delays are ignored, impatient users are best served by a greedy policy
- This result does not always hold when file sizes are drawn from a set of exponential distributions
 - the implications of this are not yet completely understood