

Optimal State-Free, Size-aware Dispatching for Heterogeneous $M/G/-$ type systems

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Outline

The model

- Dispatching model
- Static Strategies
- Analysis model

Optimal static strategies

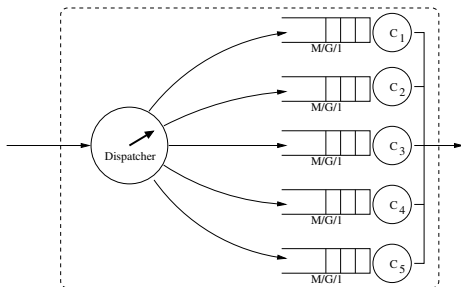
- Homogeneous servers
- Heterogeneous Servers
- Sketch of the proofs

Mappings of queues

Conclusion

Dispatching model

- ▶ One dispatcher, followed by multiple queues
- ▶ Heterogeneous: different server speeds
- ▶ FCFS policy for each queue
- ▶ Static dispatching strategy



Static strategies

- ▶ (Scheduling) Policies: algorithms used by each queue
- ▶ (Dispatching) Strategies: algorithms used by dispatcher

A class of static dispatching strategies

- ▶ Static: dispatcher is state-free
- ▶ Size-aware: dispatcher knows the job size on arrival
- ▶ Stochastic: dispatcher may randomly assign jobs

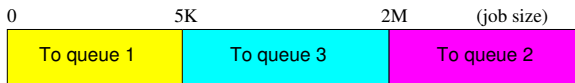
Why static?

- ▶ Easy to be implemented
- ▶ When collecting dynamic data is hard
- ▶ For the baseline of dynamic strategies

Static strategies

Examples of static strategies

- ▶ Random
- ▶ Size Interval (SI) [Harchol-Balter 1999]



- ▶ Nested Size Interval (NSI) (more generalized than SI)



(Queue 3 is nested in Queue 1)

Analysis with $M/G/-$ model

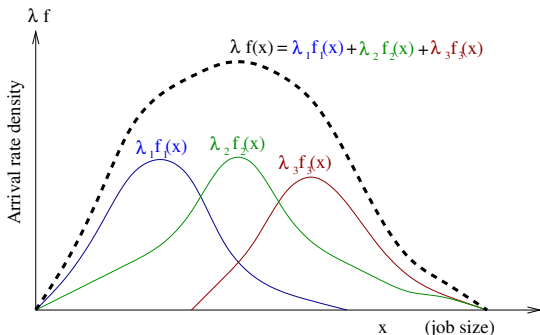
Analysis Assumptions

- ▶ Poisson arrival with rate λ
- ▶ General known job-size distribution with PDF $f(x)$

Implications

- ▶ $\lambda f(x)$ is the arrival rate density function (ARDF)
- ▶ Each queue is an $M/G/1$ whose ARDF is $\lambda_j f_j(x)$
- ▶ We have $\lambda f(x) = \sum_{i=1}^n \lambda_i f_i(x)$.
- ▶ The optimal static strategy means the optimal partitioning of function $\lambda f(x)$ to $\lambda_j f_j(x)$ such that the overall mean waiting time is minimized.

Analysis with $M/G/-$ model: example of three queues



- ▶ To find $\lambda_1 f_1(x)$, $\lambda_2 f_2(x)$, and $\lambda_3 f_3(x)$ such that $\lambda f(x) = \sum_{i=1}^3 \lambda_i f_i(x)$ and the mean waiting time is minimized.

Optimal static strategy in homogeneous case

Assume:

- ▶ Static strategy: state free, size-aware
- ▶ Homogeneous: all servers have the same speed
- ▶ $M/G/1$ -FCFS queues
- ▶ Objective measure: overall per-job mean waiting/response time

Result:

The optimal static strategy is a Size-Interval (SI) strategy.

Optimal static strategy in heterogeneous case

Assume:

- ▶ Static strategy: state free, size-aware
- ▶ Heterogeneous: servers have different speeds
- ▶ $M/G/1$ -FCFS queues
- ▶ Objective measure: overall per-job mean waiting time

Results:

- ▶ The optimal strategy may be a non-SI strategy.
(counter-example)
- ▶ It is a Nested Size-Interval (NSI) strategy.
- ▶ Slower queue can be nested in a faster queue.

Mean waiting time under static strategies

The per-job mean waiting time:

$$E[W] = \frac{1}{2\lambda} \sum_{i=1}^n \left[\frac{\lambda_i \omega_i}{c_i(c_i - \rho_i)} \right],$$

where

- ▶ λ_i is arrival rate of queue i ,
- ▶ ρ_i is the (first-order) load of queue i ,
- ▶ $\omega_i = \lambda \int_0^\infty t^2 f_i(t) dt$ is called “second-order load” of queue i ,
- ▶ c_i is the capacity (processing speed) of queue i .

Mean waiting time under static strategies (cont'd)

Mean waiting time

$$E[W] = \frac{1}{2\lambda} \sum_{i=1}^n \left[\frac{\lambda_i \omega_i}{c_i(c_i - \rho_i)} \right]$$

Objective: minimizing $E[W]$

- ▶ Recall that: partitioning $\lambda f(\cdot)$ to a sum of $\lambda_i f_i(\cdot)$'s
- ▶ Objective $E[W]$ depends on λ_i , ρ_i and ω_i .
- ▶ λ_i , ρ_i and ω_i are respectively zeroth-, first-, and second-order moments of $\lambda_i f_i(\cdot)$.

Proof for homogeneous case ($c_1 = c_2 = \dots$)

First prove for two queues, then extend to multiple queues.

For two queues: two cases

- ▶ Case 1: their loads are severely unbalanced: either of
 - ▶ $\frac{\lambda_1}{c_1(c_1 - \rho_1)} > \frac{\lambda_2}{c_2(c_2 - \rho_2)}$ and $\frac{\omega_1}{c_1(c_1 - \rho_1)} > \frac{\omega_2}{c_2(c_2 - \rho_2)}$
 - ▶ $\frac{\lambda_1}{c_1(c_1 - \rho_1)} < \frac{\lambda_2}{c_2(c_2 - \rho_2)}$ and $\frac{\omega_1}{c_1(c_1 - \rho_1)} < \frac{\omega_2}{c_2(c_2 - \rho_2)}$
- ▶ Case 2: otherwise.

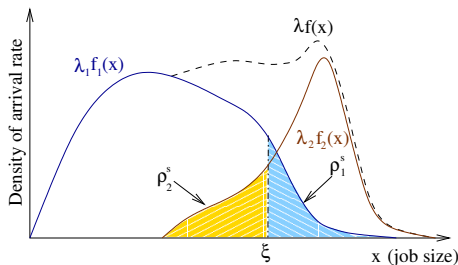
Case 1: Severely unbalanced

- ▶ Transfer some jobs (of any size) from the high-loaded queue to the other.
- ▶ $E[W]$ is lower.
- ▶ Repeat doing so until Case 2.

Proof for homogeneous case (cont'd)

Case 2: not severely unbalanced

- ▶ Find a threshold such that shaded areas have same ρ .
- ▶ Swapping shaded areas yields an SI strategy.
- ▶ We show that $E[W]$ is lower.



Proof for heterogeneous case ($c_1 < c_2 < \dots$)

For two queues: $c_1 < c_2$

- ▶ Case 1: their loads are severely unbalanced: either of
 - ▶ Case 1a: $\frac{\lambda_1}{c_1(c_1 - \rho_1)} > \frac{\lambda_2}{c_2(c_2 - \rho_2)}$ and $\frac{\omega_1}{c_1(c_1 - \rho_1)} > \frac{\omega_2}{c_2(c_2 - \rho_2)}$
 - ▶ Case 1b: $\frac{\lambda_1}{c_1(c_1 - \rho_1)} < \frac{\lambda_2}{c_2(c_2 - \rho_2)}$ and $\frac{\omega_1}{c_1(c_1 - \rho_1)} < \frac{\omega_2}{c_2(c_2 - \rho_2)}$
- ▶ Case 2: otherwise.

Cases 1b and Case 2

- ▶ Use the same arguments as in a homogeneous system.

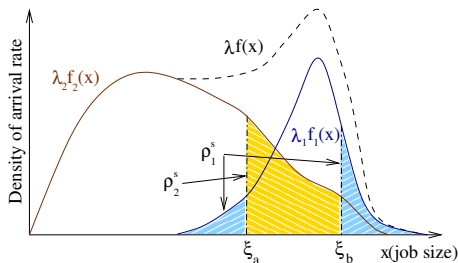
Case 1a

- ▶ Previous argument fails.

Proof for heterogeneous case (cont'd)

Case 1a (Slower queue has severely higher load)

- ▶ find two thresholds: shaded areas have same λ 's and ρ 's.
- ▶ Swapping shaded areas yields an NSI strategy
- ▶ We show that $E[W]$ gets lower.



Remaining issues for finding an optimal static strategy

Homogeneous case

- ▶ To find the optimal thresholds of size intervals.
- ▶ Mapping (which queue gets which interval) is irrelevant.

Heterogeneous case

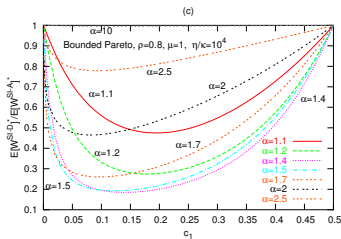
- ▶ Nested size intervals are more complicated.
- ▶ Mappings between queues and intervals matter.

What is the best mapping in heterogeneous case?

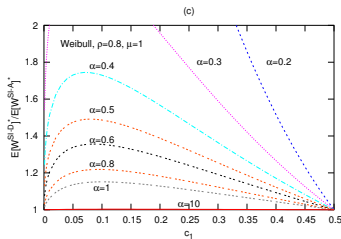
- ▶ No fixed rules(e.g. slower queue gets the interval of shorter jobs)
- ▶ Depending on job size distributions

Mapping of intervals to queues: examples

- ▶ Only SI strategy (more simpler than NSI) and two queues
- ▶ Compare two mappings (ascending and descending) with corresponding optimal thresholds



Bounded Pareto



Weibull

Best: faster server gets short jobs Best: faster server gets long jobs
 (Y-axis: ratio between optimal mean response times of ascending and descending
 mapping. X-axis: speed fraction of the slower queue)

Mapping of intervals to queues: examples (cont'd)

- ▶ Two queues, SI strategy
- ▶ Only two mappings
 - ▶ Ascending mapping: faster server gets interval of long jobs
 - ▶ Descending mapping: faster server gets interval of short jobs
- ▶ For each mapping, an optimal threshold (or load partitioning) can be found.
- ▶ Log-normal distribution is mapping-invariant here (two mappings obtains the same optimal) response time.
- ▶ All distributions such that

$$m(x) = E[X] - m\left(\frac{\psi}{x}\right), \text{ where } m(x) = \int_0^x t dF(t), \psi: \text{constant}$$

is mapping-invariant (for two queues and SI strategy).

Conclusion

- ▶ For FCFS, optimal static strategy is an NSI.
- ▶ For FCFS homogeneous queues, optimal static strategy is an SI.
- ▶ Other scheduling policies can be used on each queue.
- ▶ For PS, size information is useless (w.r.t. mean response time).
- ▶ Dynamic case is more complicated. [Whitt 1984]
- ▶ It is difficult to find the best mapping.