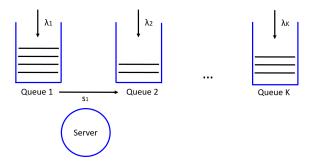
Optimal Control of Polling Systems with Large Switchover Times

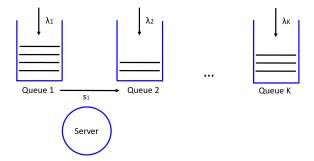
Yue Hu (Columbia Business School)

Joint work with Jing Dong (Columbia Business School) and Ohad Perry (Northwestern University) A polling system is a stochastic queueing network where several queues are served by one switching server



The Model

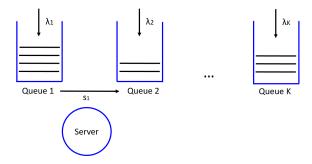
A polling system is a stochastic queueing network where several queues are served by one switching server



- A single server switching between $K \ge 2$ queues
- The server attends to the queues in a periodic cycle of *a* stages defined by the polling table *p* : {1,...,*a*} → {1,...,*K*} (*a* ≥ *K*)
 - p(j) = i the server serves queue *i* at stage *j* within a cycle

The Model

A polling system is a stochastic queueing network where several queues are served by one switching server



- Stationary arrival process of jobs with rate λ_i to queue *i*
- IID service times with mean $1/\mu_i$ at queue $i \in \{1, \ldots, K\}$
- Necessary stability condition: $\sum_{i=1}^{K} \lambda_i / \mu_i < 1$
- Large switchover time s_i is incurred when server switches from queue i

Stochastic Dynamics

- $Q_i(t)$ number in queue *i* at time *t*, $1 \le i \le K$
- Z(t) the stage of the server at time t:
 Z(t) = j server is at stage j, j = 1, ..., a
 Z(t) = ⊖_j server is switching from stage j to stage j + 1 (modulo a)

The stochastic dynamics are characterized via the process

$$X(t) := (Q_i(t), Z(t); i = 1, \dots, K), \quad t \ge 0$$

<u>Remark</u>: Given a routing order and service policy, X(t) is a well-defined stochastic process

Applications

Polling systems are used to model computer, communication, and production systems

Stochastic economic lot scheduling: make-to-stock production of multiple products on a single machine



Polling System:

- Zero Switchover Time: Coffman et al. (1995), Remerova et al. (2014)
- <u>Small Switchover Time</u>: Boxma et al. (1990), Fricker and Jaibi (1993), Borst and Boxma (1997), Coffman et al. (1998), Olsen and Van der Mei (2003, 2005)
- Large Switchover Time: Van der Mei (1999), Olsen (2001), Lan and Olsen (2006)

Hybrid Dynamical Systems:

Matveev et al. (2016), Feoktistova et al. (2012), Perry and Whitt (2016), Dong and Perry (2018)

Our goal is to find a control that minimizes long run average holding costs among all "admissible" controls that are

- non-idling
- discrete-Markov (Given the queue length at the polling instant, the service policies do not depend on the history of the service process up to that instant)

Remark:

- X(t) is not necessarily a markov process under an admissible control
- It is prohibitively hard to solve the optimal control problem exactly.
- We carry out optimality analysis in an asymptotic sense.

Consider a hybrid dynamic system (HDS) characterized by

$$x(t) = (q(t), z(t)),$$

where x(t) is a hybrid of the piece-wise linear continuous number-in-system process $q(t) \in [0, \infty)$ and the discrete server-position process $z(t) \in \{j, \ominus_j, 1 \le j \le a\}$, characterized via

$$\dot{q}_i(t) = (\lambda_i - \mu_i) \mathbf{1}_{\{p(z(t))=i,q(t)>0\}} + \lambda_i \mathbf{1}_{\{p(z(t))\neq i\}}, \quad 1 \le i \le K$$

A Fluid Approximation to Simplify the Problem

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Remark:

• Under a fixed routing order and service policy, x(t) is well-defined

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Fluid Optimal Control Problem: To find a control that minimizes

$$c := \limsup_{t \to \infty} rac{1}{t} \int_0^t \sum_{i=1}^K \psi_i(q_i(s)) ds,$$

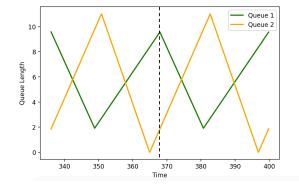
where $\psi_i(\cdot)$ is the cost rate for queue *i*

A Road Map of Our Approach:

- Find the optimal periodic equilibrium among all the possible periodic equilibria of the HDS
- Design a fluid control that achieves the optimal equilibrium
- "Interpret" the fluid control into a control for the stochastic system
- Prove: The control is asymptotically optimal

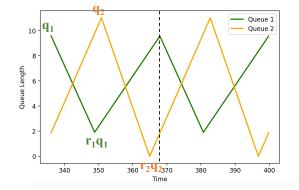
Characterization of Periodic Equilibria

E.x. periodic equilibrium with two queues and server routing $\{1, 2|1, 2|...\}$



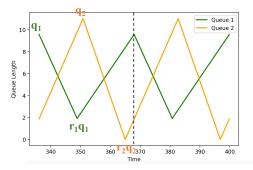
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Fixed-Proportion Reduction Control (FPR)

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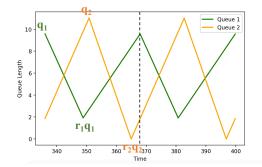


The FPR Control (Definition):

- For fixed {r_j ∈ [0, 1], j = 1, ..., a}, if p(j) = i, then reduce q_i (the value of queue i at the beginning of service) to r_iq_i
- After serving station j, switch to station j + 1 in the polling table

Fixed-Proportion Reduction Control (FPR)

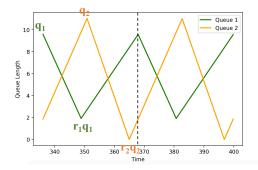
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<u>Remark:</u> Any periodic equilibrium can be translated to some control parameters of the FPR

Fixed-Proportion Reduction Control (FPR)

E.x. periodic equilibrium with two queues and server routing $\{1, 2|1, 2|...\}$



Theorem

Independently of initialization,

- the HDS converges to a unique periodic equilibrium under FPR with any non-trivial control parameters
- the HDS converges to the desired equilibrium under the translated FPR

There is a bijective mapping from the possible periodic equilibria of the HDS to the FPR control.

Therefore, minimizing the long-run average $\cot c$ is equivalent to finding the optimal FPR control:

$$\min_{\{r_j, j=1,...,a\}} \quad \frac{1}{\tau} \int_0^\tau \sum_{i=1}^K \psi_i\left(q_i(s, r)\right) ds \quad s.t. \quad r_j \in [0, 1], \quad j = 1, ..., a,$$

where τ is the equilibrium cycle length

The Optimal Control

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- E.x.1: Under routing $\{1, 2, 3 | 1, 2, 3 | \dots\}$, optimal ratio is $\{0\%, 0\%, 0\%\}$
- E.x.2: Under routing $\{1, 3, 2, 3, 2|1, 3, 2, 3, 2| \dots \}$, optimal ratio is $\{0\%, 48.86\%, 0\%, 0\%, 0\%\}$

 $\lambda_i = 2, \mu_i = 8, s_i = 2, i = 1, 2, 3$ linear cost rates $\psi_1(\cdot) = 1, \psi_2(\cdot) = 4, \psi_3(\cdot) = 1$.

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Proposition

It is optimal to exhaust each queue at least once during a cycle

Corollary

In the case of cyclic routing, the exhaustive policy is optimal

The Road Map Revisited:

- Find the optimal periodic equilibrium among all the possible periodic equilibria of the HDS
- Design a fluid control that achieves the optimal equilibrium
- "Interpret" the fluid control into a control for the stochastic system
 - Reduce queue *i* from q_i to $\lceil r_j q_i \rceil$ if the server is at queue *i* in stage *j*
- Prove: The control is asymptotically optimal

Consider a sequence of systems indexed by *n* under an admissible control π^n

- Large switchover times: switchover times satisfy $s_i^n = ns_i$ in n^{th} system
- Temporal and spacial scalings: $\bar{X}_{\pi^n}^n(t) := (Q_{\pi^n}(nt)/n, Z_{\pi^n}(nt))$
- <u>Scaled long-run average costs</u>: $\bar{C}_{\pi^n}^n = \lim_{t \to \infty} \frac{1}{t} \int_0^t \psi(\bar{Q}_{\pi^n}^n(s)) ds$

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Let π_*^n denote the optimal fluid FPR control translated for the n^{th} stochastic system, and c_* denote the optimal fluid objective value

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Theorem (Asymptotic Optimality)

- Under any sequence of policies π^n , $\liminf_{n\to\infty} \overline{C}_{\pi^n}^n \ge c_*$
- Under π_*^n , $\bar{C}_{\pi_*}^n \Rightarrow c_*$ as $n \to \infty$ (The FPR is asymptotically optimal)

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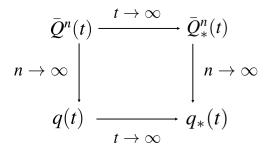
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 \Rightarrow For cyclic routing, the exhaustive policy is asymptotically optimal

Asymptotic Optimality (Proof)



Theorem (Limit Interchange)

Under any FPR control parameters,

$$\bar{C}^n = \lim_{t \to \infty} \frac{1}{t} \int_0^t \psi\left(\bar{Q}^n(s)\right) ds \Rightarrow c \text{ as } n \to \infty$$

In particular, $\bar{C}^n_{\pi^n_*} \Rightarrow c_* \text{ as } n \to \infty \text{ under } \pi^n_*$

- We considered the optimal-control problem of a polling system with large switchover times when long-run average holding costs are to be minimized
- We proposed the Fixed-Proportion Reduction Control (FPR):
 - all possible periodic equilibria are achievable under FPR
 - stable under all control parameters
- We find the optimal FPR and prove that the control is asymptotically optimal under large-switchover-time scaling
- In the special case of cyclic routing, the exhaustive policy is optimal

Thank You