

A Resource-Sharing Game with Relative Priorities

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File hosting \Rightarrow premium packages

- Higher payment, higher bandwidth

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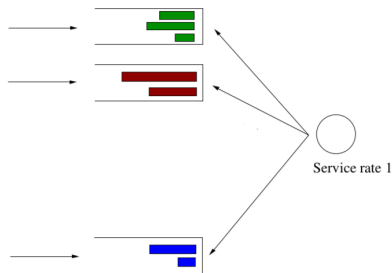
\Rightarrow Better service increasing payment

- 1 Model Description
- 2 Solution of Original Game
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Model: system parameters

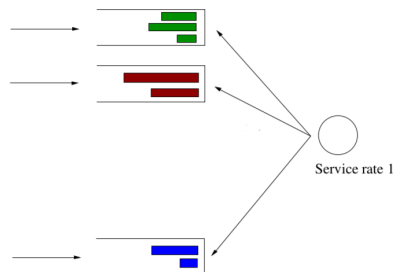
$\mathcal{C} = \{1, 2, \dots, R\}$ set of players (classes of users) paying for service



⇒ **Quality of service** of classes: function of processing time

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⇒ **Quality of service** of classes: function of processing time

- λ_i : arrival rate
- B_i : service requirement r. v.
- $\rho_i = \lambda_i \mathbb{E}(B_i)$: class- i load
- $T_i(\mathbf{g})$: response time of tasks of class i
- $\mathbb{E}(T_i(\mathbf{g})) = \bar{T}_i(\mathbf{g})$

Model: resource sharing

- *Strategy* of players: amount that they pay $g_i \in [\epsilon, \infty)$.
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Objective:

Minimize payment ensuring the QoS requirement

Known Results of DPS

Difficult model

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- for exponentially distributed required service times and $R \geq 2$:

$$\bar{T}_k(\mathbf{g}) \left(1 - \sum_{j=1}^R \frac{\lambda_j \mathbf{g}_j}{\mu_j \mathbf{g}_j + \mu_k \mathbf{g}_k} \right) - \sum_{j=1}^R \frac{\lambda_j \mathbf{g}_j \bar{T}_j(\mathbf{g})}{\mu_j \mathbf{g}_j + \mu_k \mathbf{g}_k} = \frac{1}{\mu_k}, \quad k \in \mathcal{C}.$$

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- For non-exponentially distributed service times, $\bar{\mathbf{T}}(\mathbf{g})$ is a solution to a set of integro-differential equations (Fayolle et al 80).
- when $\rho \rightarrow 1$,

$$(1 - \rho) T_i(\mathbf{g}) \xrightarrow{d} T_i(\mathbf{g}; 1) = X \cdot \frac{\mathbb{E}(B_i)}{g_i}, \quad i \in \mathcal{C}, \quad (1)$$

where X is an exponentially distributed random variable

Game (OPT-M)

Each player i

$$\begin{array}{ll} \min_{g_i \geq \epsilon} & \rho_i g_i \\ \text{subject to} & \bar{T}_i(\mathbf{g}) \leq c_i. \end{array} \quad (\text{OPT-M})$$

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Question:

What price should a player pay?

Summary of Results

	Game (OPT-M)		Game (OPT-HT)	
	N. Classes	Serv. Times	N. Classes	Serv. Times
Feasibility	Arbitrary	Exponential	Arbitrary	General
Existence of NE	Arbitrary	General	Arbitrary	General
Uniqueness of NE	2	General	Arbitrary	General
NE Characterization	2	Exponential	Arbitrary	General
Price of Anarchy	2	General	Arbitrary	General
BR Convergence (feasible point)	Arbitrary	General	Arbitrary	General
BR Convergence (any point)	2	Exponential	2	General

Summary of main results

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Definition (Feasibility)

The game (OPT-M) is feasible if and only if it exists a performance vector such that $\bar{T}_i(\mathbf{g}) \leq c_i, i \in \mathcal{C}$.

Existence of Equilibrium

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Proposition

With general service time distributions, if the game is feasible, then

- there exists a **Nash Equilibrium**
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Since:

- $\bar{T}_i(\mathbf{g})$ decreases with g_i and increases with $g_j, \forall j \neq i$
- $\bar{T}_i(c\mathbf{g}) = \bar{T}_i(\mathbf{g})$

For exponential service times

Proposition

The game (OPT-M) is *feasible* if and only if

$$\sum_{i \in r} \rho_i c_i \geq W_r, \quad \forall r \text{ subset of } \mathcal{C}.$$

where $\bar{\rho}_r = \sum_{i \in r} \rho_i$ and $W_r = \frac{1}{1 - \bar{\rho}_r} \sum_{i \in r} \frac{\rho_i}{\mu_i}$.

Particular case

If $\exists \mathbf{g}$ such that $\bar{T}_i(\mathbf{g}) = c_i \Rightarrow$ infinite equilibria

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Two players and exponential service times

Let $\mathbb{E}(B_i) = 1/\mu_i$. Assume $c_1\mu_1 \leq c_2\mu_2$.

If the game is feasible, then the **unique equilibrium** is

- Let $\mathbf{g}^{PS} = (\epsilon, \epsilon)$. If $\bar{T}_i(\mathbf{g}^{PS}) \leq c_i$, then $\mathbf{g}^{NE} = \mathbf{g}^{PS}$,
- otherwise, $\mathbf{g}^{NE} = (g_1^{NE}, \epsilon)$, where $g_1^{NE} = \epsilon \frac{-\mu_1\rho_2 + \mu_2(1-\rho_2)[\mu_1 c_1(1-\rho) - 1]}{-\mu_1\rho_2 - \mu_1(1-\rho_1)[\mu_1 c_1(1-\rho) - 1]}$.

Example:

2 classes and exp serv times

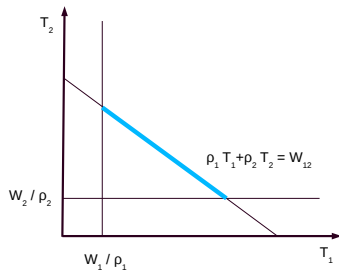


Figure: Set of performance vectors in a DPS queue

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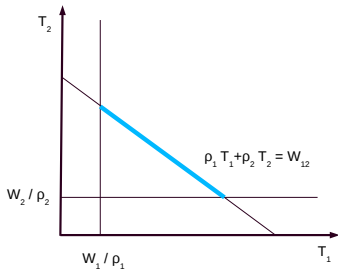


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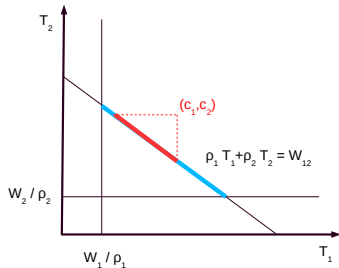


Figure: Set of performance vectors such that $\bar{T}_i(\mathbf{g}) \leq c_i$ (feasibility)

$$\text{Feasibility} \iff \begin{aligned} \rho_i c_i &\geq W_i, i = 1, 2 \\ \rho_1 c_1 + \rho_2 c_2 &\geq W_{12} \end{aligned}$$

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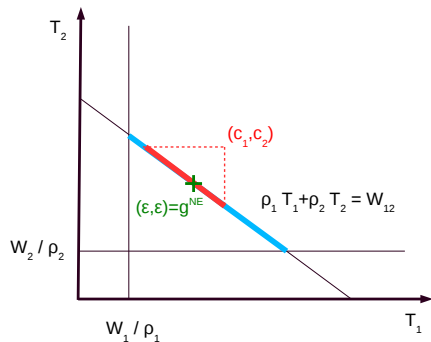


Figure: The case $\bar{T}_i(g^{PS}) \leq c_i$.

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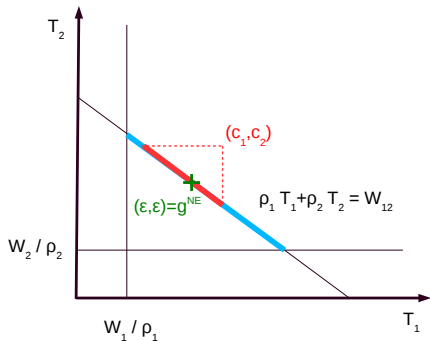


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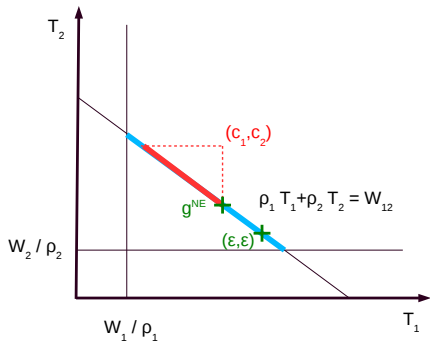


Figure: The case $\bar{T}_1(\mathbf{g}^{PS}) > c_1$.

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The HT Game

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Proposition (Verloop et al 2011)

When scaled with $1 - \rho$, the response time of class- i jobs has a proper distribution as $\rho \rightarrow 1$.

$$(1 - \rho) T_i(\mathbf{g}) \xrightarrow{d} T_i(\mathbf{g}; 1) = X \cdot \frac{\mathbb{E}(B_i)}{g_i}, \quad i \in \mathcal{C}, \quad (2)$$

where \xrightarrow{d} denotes convergence in distribution and X is an exponentially distributed random variable with mean

$$\mathbb{E}(X) = \frac{\sum_k \lambda_k \mathbb{E}(B_k^2)}{\sum_k \lambda_k \mathbb{E}(B_k^2) \frac{1}{g_k}}. \quad (3)$$

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Assume convergence in mean:

$$(1 - \rho) \bar{T}_i(\mathbf{g}; \rho) = \bar{T}_i(\mathbf{g}; 1) = \frac{\mathbb{E}(B_i)}{g_i} \frac{\sum_k \lambda_k \mathbb{E}(B_k^2)}{\sum_k \lambda_k \mathbb{E}(B_k^2) \frac{1}{g_k}}$$

Theorem

Assume $\mathbb{E}(B_i)/\tilde{c}_i$ decreasing with i , where $\tilde{c}_i = c_i (1 - \rho)$.

If the game is feasible, the *unique Nash equilibrium* is

$$g_i^{NE} = \epsilon \frac{\tilde{t}_m / \mathbb{E}(B_m)}{\tilde{c}_i / \mathbb{E}(B_i)}, \text{ for all } i < m,$$

$$g_i^{NE} = \epsilon, \text{ for all } i \geq m,$$

where m is the minimum value such that there exists $\tilde{t}_m \leq \tilde{c}_m$ verifying

$$\frac{\tilde{t}_m}{\mathbb{E}(B_m)} = \frac{\sum_{k=1}^R \lambda_k \mathbb{E}(B_k^2) - \sum_{k=1}^{m-1} \lambda_k \frac{\mathbb{E}(B_k^2)}{\mathbb{E}(B_k)} \tilde{c}_k}{\sum_{k=m}^R \lambda_k \mathbb{E}(B_k^2)}.$$

Arbitrary number of classes and general service times distribution.

Approximating (OPT-M)

Let $\mathbb{E}(B_i)/c_i \geq \mathbb{E}(B_j)/c_j$, if $i < j$.

Using $\bar{T}_i(\mathbf{g}) = \frac{\bar{T}_i(\mathbf{g};1)}{1-\rho} \Rightarrow$ **Approximated NE**:

Corollary

$$g_i^{NE} = \epsilon \frac{t_m / \mathbb{E}(B_m)}{c_i / \mathbb{E}(B_i)}, \text{ for all } i < m,$$

$$g_i^{NE} = \epsilon, \text{ for all } i \geq m,$$

where $m = 1, \dots, R$ is the minimum value such that there exists a value $t_m \leq c_m$ verifying

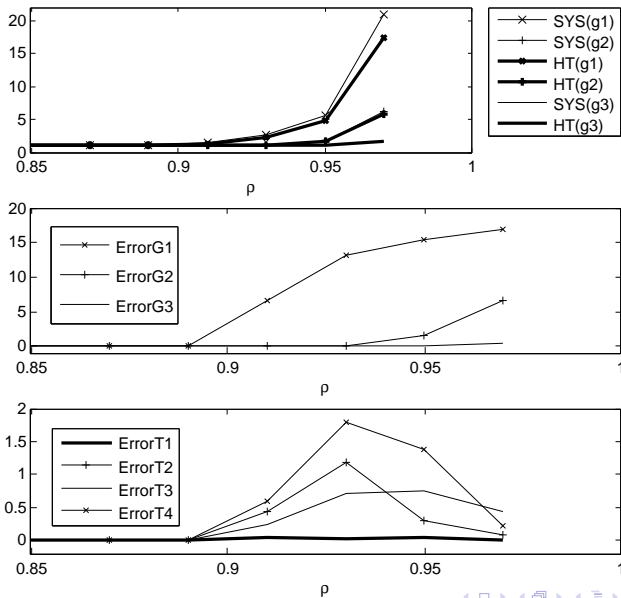
$$\frac{t_m}{\mathbb{E}(B_m)} = \frac{\sum_{k=1}^R \frac{\lambda_k \mathbb{E}(B_k^2)}{(1-\rho)} - \sum_{k=1}^{m-1} \lambda_k \frac{\mathbb{E}(B_k^2)}{\mathbb{E}(B_k)} c_k}{\sum_{k=m}^R \lambda_k \mathbb{E}(B_k^2)}. \quad (4)$$

$$\tilde{c}_i = c_i (1 - \rho),$$

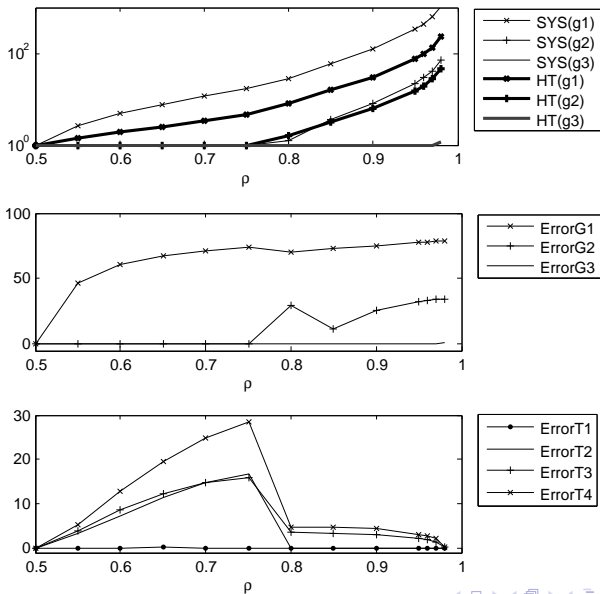
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4 classes and exp serv times: homogeneous players



4 classes and exp serv times: heterogeneous players



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Game theory and time-sharing systems: few previous work

Complicated model

- solved in some case
- for the rest, HT approximation

Future:

- Convergence of the Best Response
- Multiserver
- Users decreasing λ_i if $g_i > M$

Thank you

Thank you for your attention.