

Performance Paradox of Dynamic Bipartite Matching Models

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Joint work with I. Iriondo (UPV/EHU)

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October 9, 2024

Outline

- 1 Introduction
- 2 Model Description
- 3 Main Results
- 4 Conclusions

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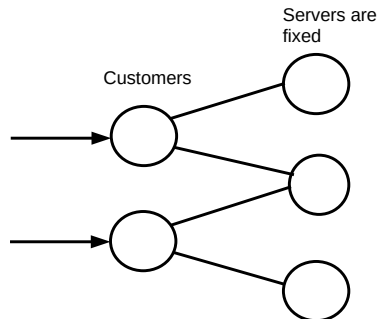
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2 Model Description

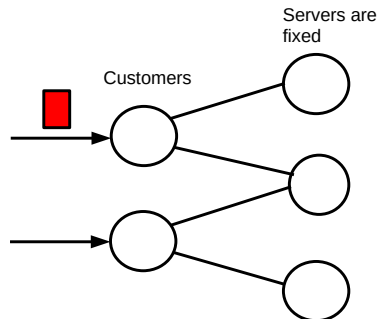
3 Main Results

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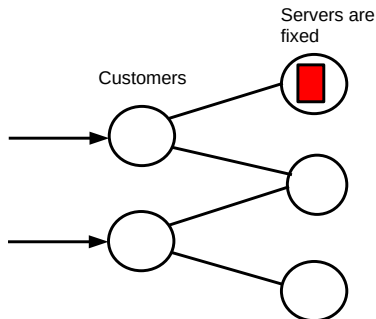
Multiskilled systems (call centers)



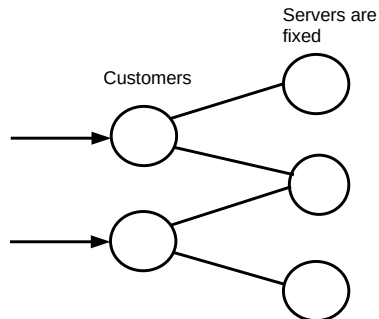
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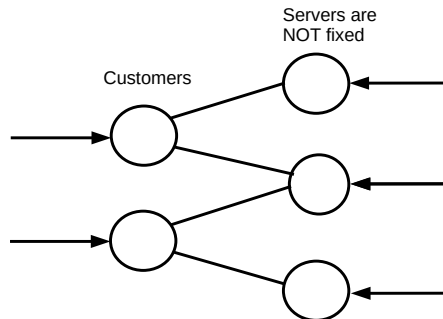
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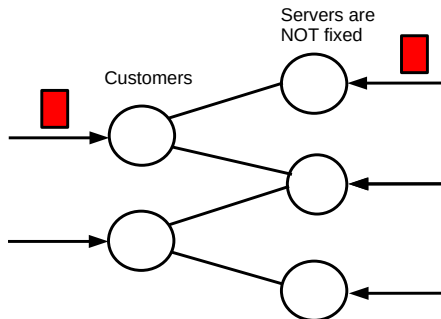
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Matching models

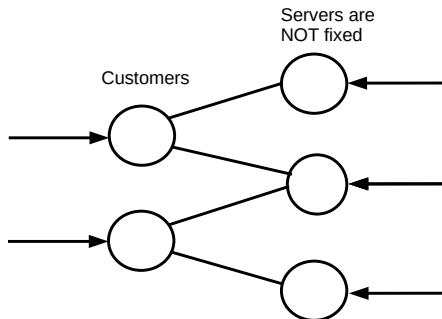


Matching models



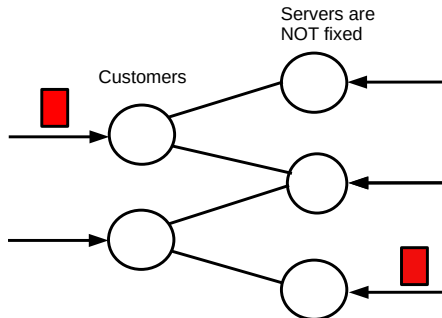
Matched customers and servers leave immediately

Matching models



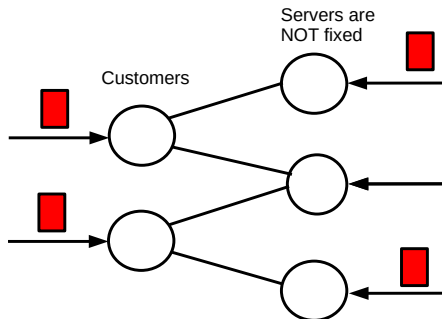
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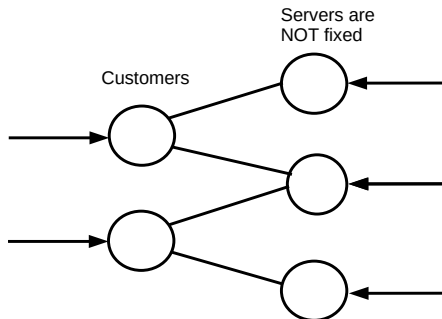
Unmatched customers and servers are stored in queues

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Bipartite Matching Models

Defined by:

- Compatibility graph \mathcal{G} .
- Distribution of arrivals of customers and servers
- Matching policy ψ

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$\mathcal{G} = (\mathcal{C} \cup \mathcal{S}, \mathcal{E})$, where $\mathcal{E} \subset \mathcal{C} \times \mathcal{S}$

Example: $\mathcal{C} = \{c_1, c_2\}$, $\mathcal{S} = \{s_1, s_2, s_3\}$ and
 $\mathcal{E} = \{(c_1, s_1), (c_1, s_2), (c_2, s_2), (c_2, s_3)\}$.

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$\vec{\alpha}$ for customers and $\vec{\beta}$ for servers

Example: $\vec{\alpha} = (\alpha_1, \alpha_2)$ and $\vec{\beta} = (\beta_1, \beta_2, \beta_3)$

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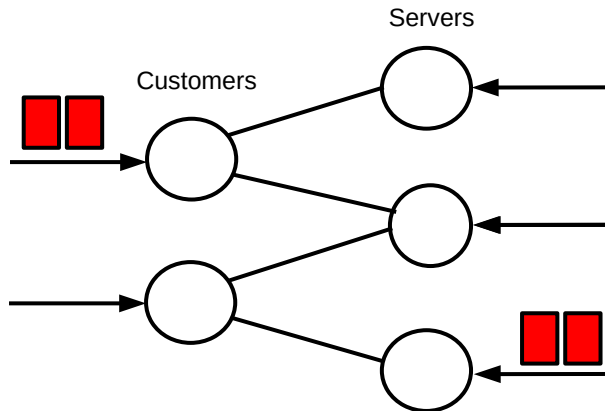
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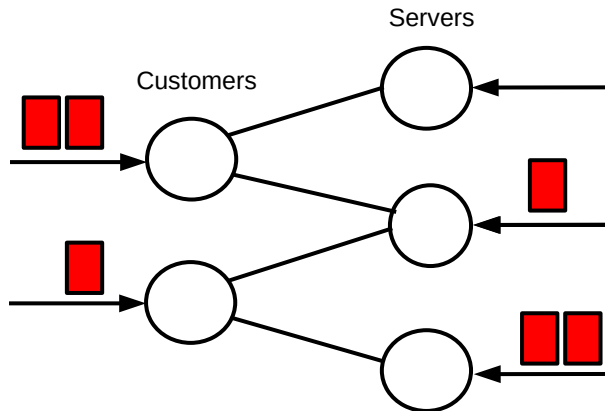
How compatible customers and servers are matched?

- In order of arrivals
- Prioritize the class with the longest/shortest number of elements...

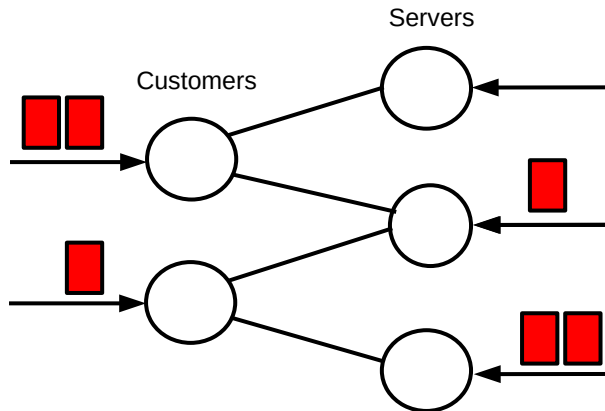
Matching Policy: Example



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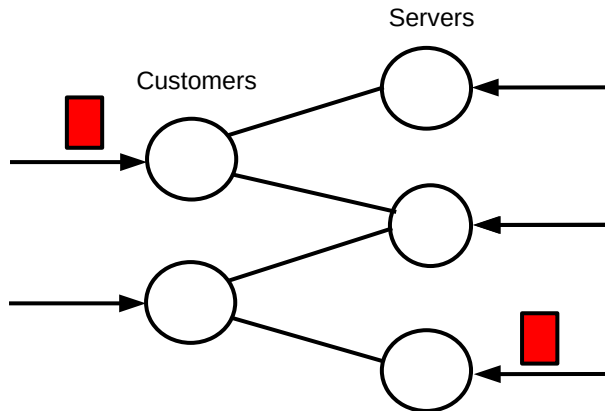


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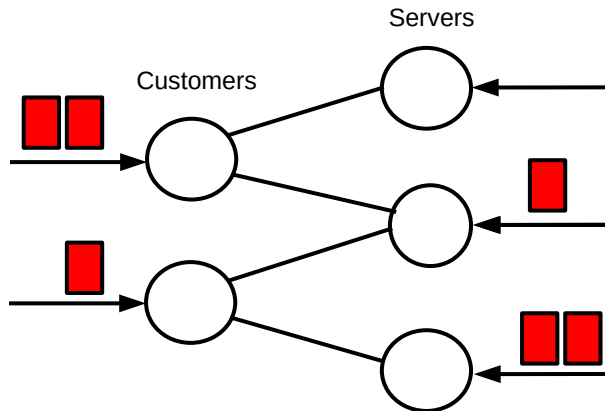
First-Come-First-Matched (FCFM)

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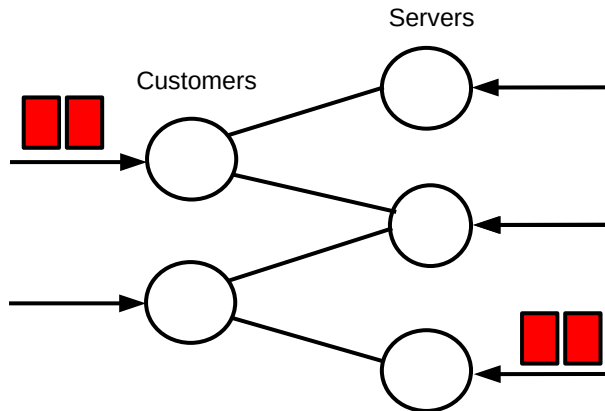
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Last-Come-First-Matched

Matching Policy: Example



Last-Come-First-Matched

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- Distribution of arrivals of customers and servers (independent)
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c_1 and s_3 arrive with probability $\alpha_1\beta_3$.

- Matching policy ψ

How compatible customers and servers are matched?

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When \mathcal{G} , $(\vec{\alpha}, \vec{\beta})$ and ψ are fixed

The number of unmatched items is a Discrete Time Markov Chain.

Stability of Markov Chain

Busic, Gupta and Mairesse, 2013

For FCFM, the Markov chain is stable iff $\forall C \subset \mathcal{C} \forall S \subset \mathcal{S}$

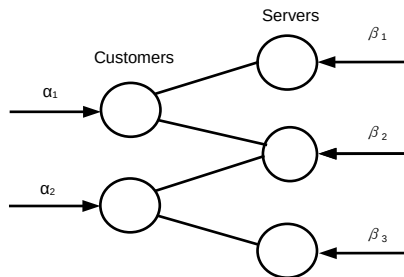
$$\sum_{c_i \in C} \alpha_i < \sum_{s_i \in S(C)} \beta_i \text{ and } \sum_{s_i \in S} \beta_i < \sum_{c_i \in C(S)} \alpha_i,$$

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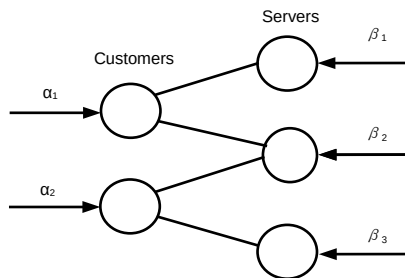


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- $\alpha_1 < \beta_1 + \beta_2$
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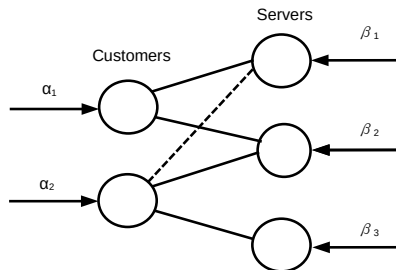
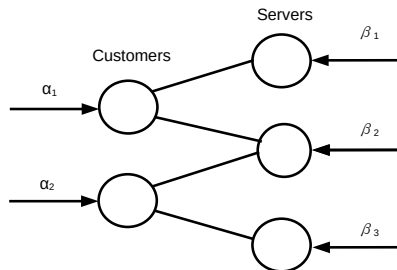
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Model Description

FCFM matching policy

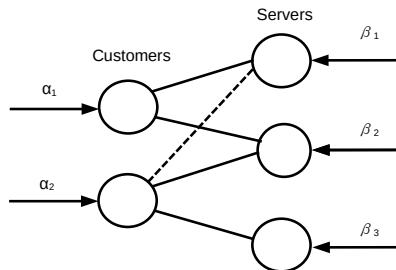
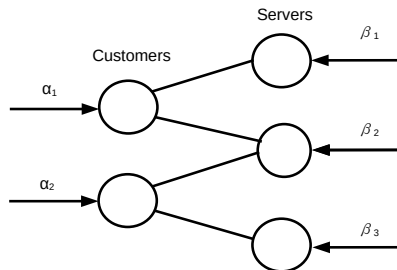
Objective: study the influence of adding an edge to the compatibility graph



Model Description

FCFM matching policy

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Definition: Performance paradox

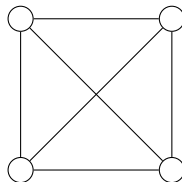
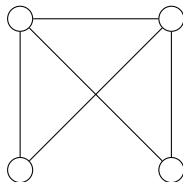
Adding an edge the mean number of unmatched customers and servers increases (analogue of Braess paradox).

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Previous works of performance paradox in matching models focus on a general matching models

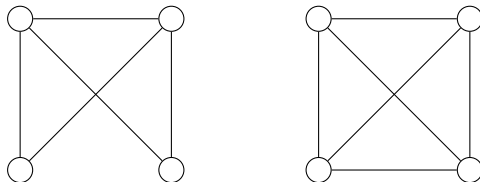


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Necessary and sufficient conditions on the arrivals such that performance paradox exists in a quasicomplete graph under FCFM (Cadas et al, 2021) and under greedy policies (Busic et al, 2024)

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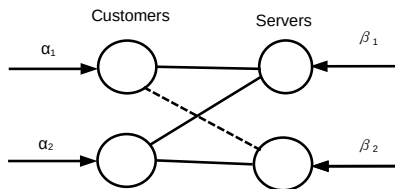
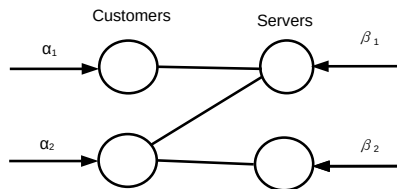
Performance Paradox Analysis

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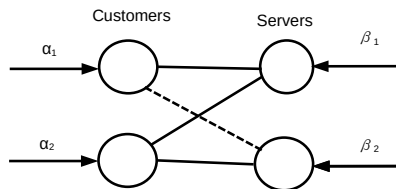
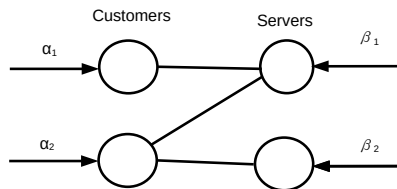
First (and simplest) approach:



Performance Paradox Analysis

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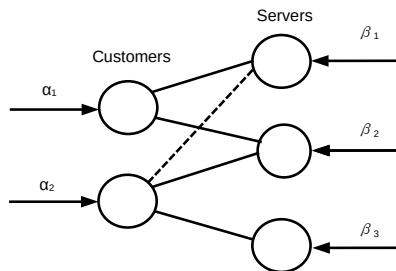
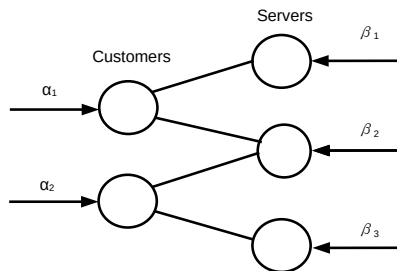
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The performance paradox does NOT exist for this case

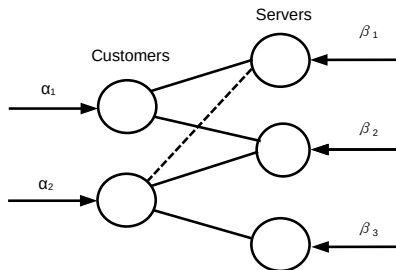
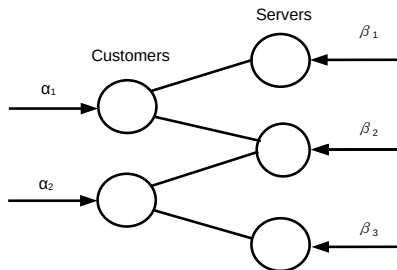
Performance Paradox Analysis

Second approach:



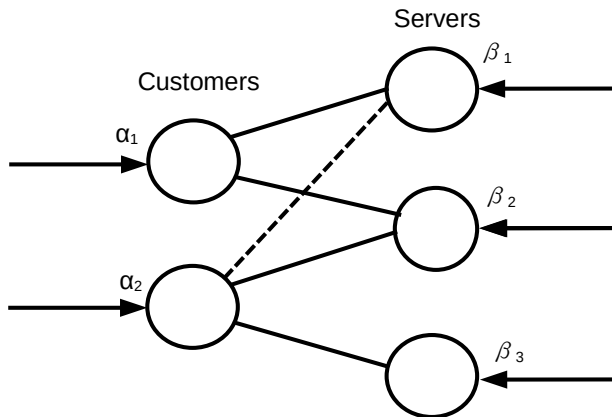
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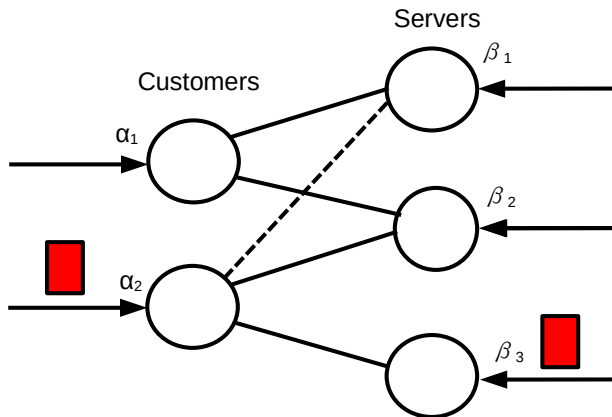


Under FCFM, the Markov chains of the unmatched elements are not difficult to analyze

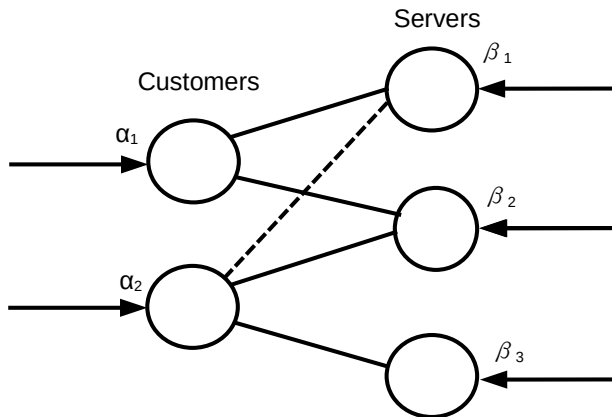
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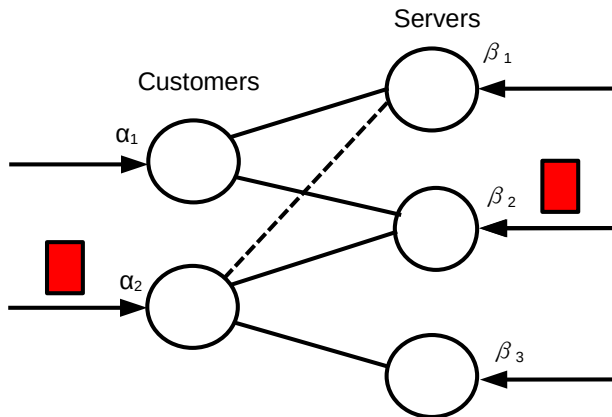
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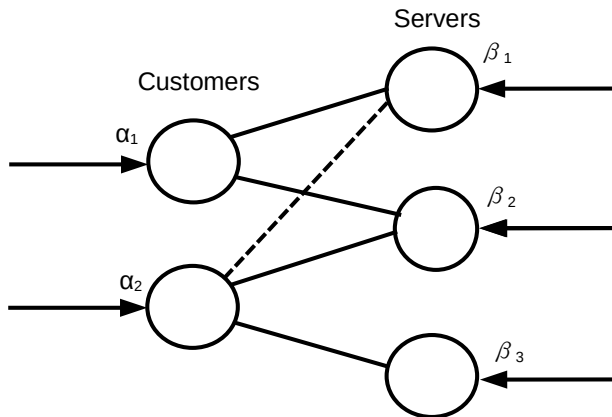
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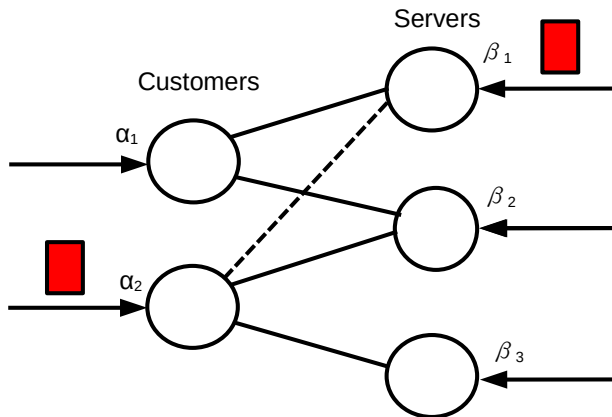
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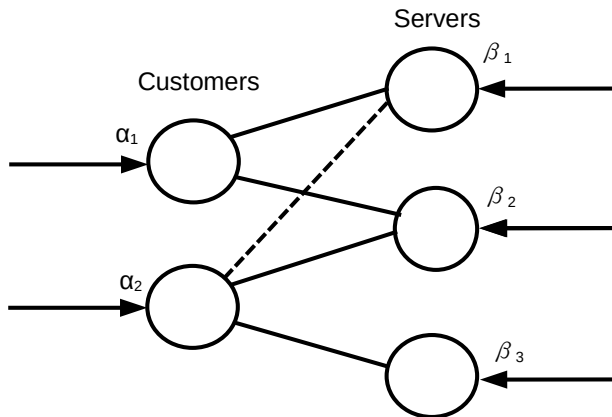
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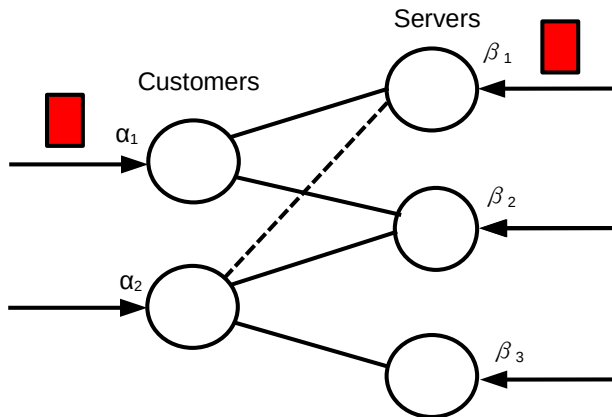
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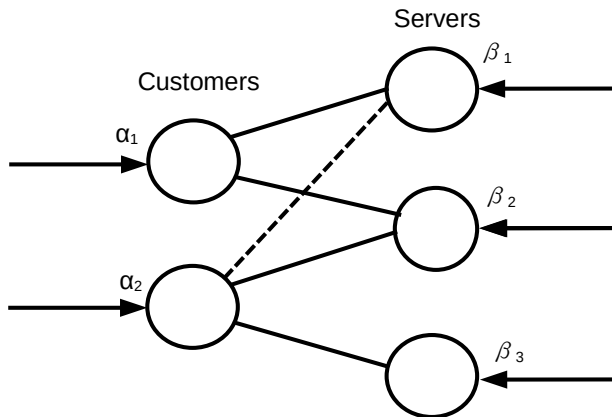
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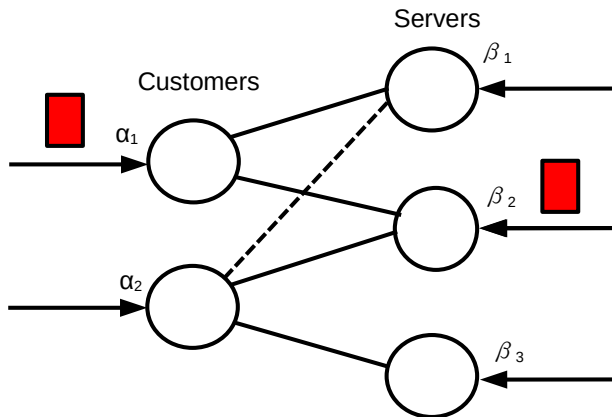
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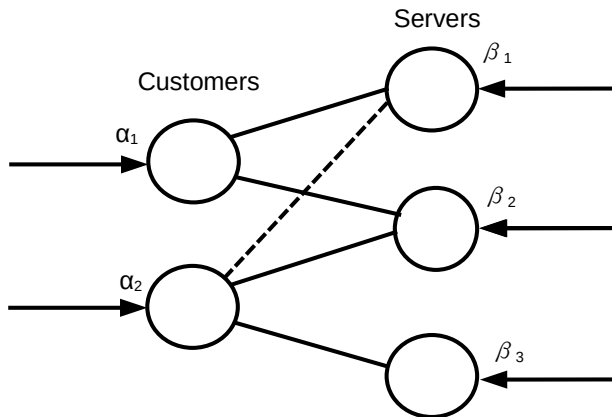
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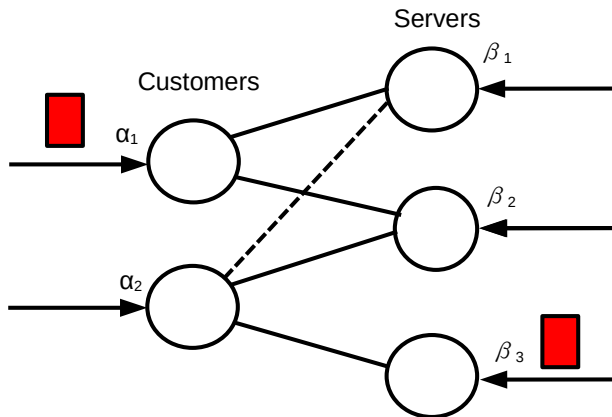
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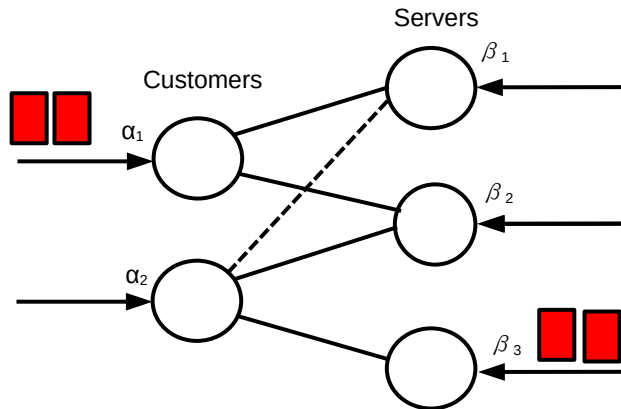
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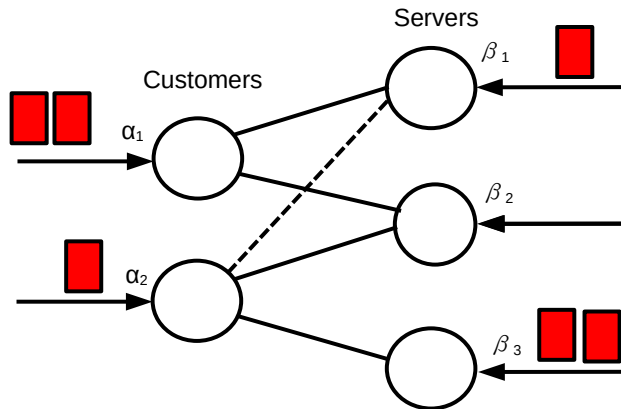
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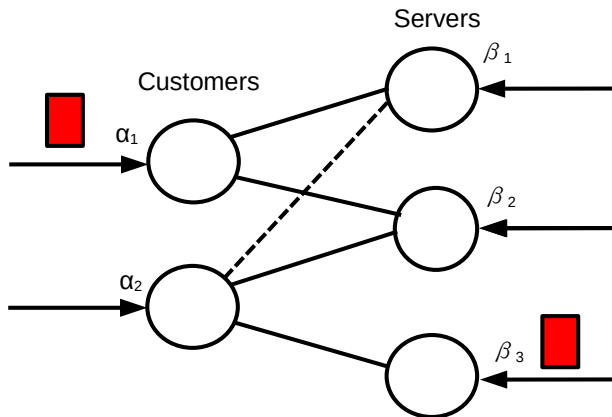
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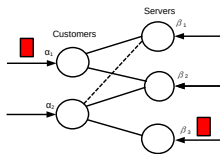
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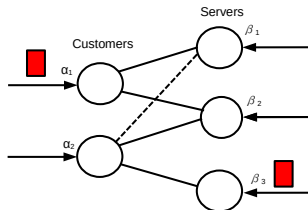


Performance Paradox Analysis



The couple (c_1, s_3) is the only possible unmatched pair of customer and server.

Performance Paradox Analysis



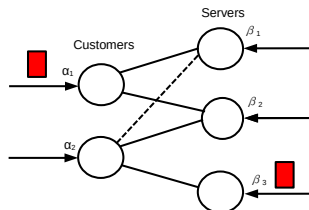
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The Markov chain is a birth-death process:

Birth probability: $\alpha_1 \beta_3$

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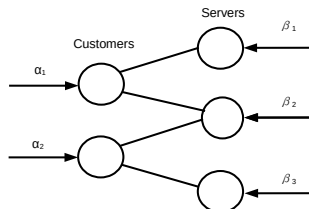
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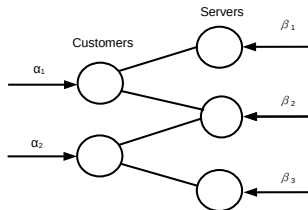
Let $\rho = \frac{\alpha_1 \beta_3}{\alpha_2 (1 - \beta_3)}$. If $\rho < 1$, the mean number of unmatched elements is $2 \frac{\rho}{1 - \rho}$

Performance Paradox Analysis



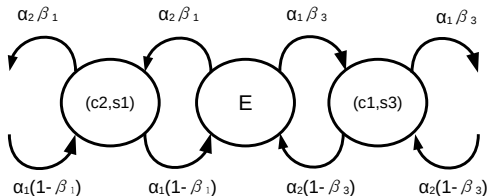
The couples (c_1, s_3) and (c_2, s_1) are the only possible unmatched pair of customer and server.

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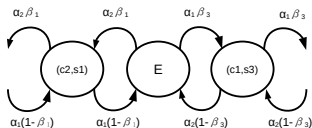
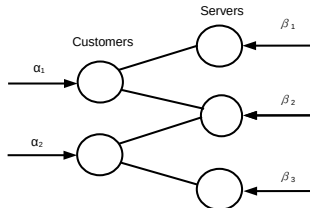


The couples (c_1, s_3) and (c_2, s_1) are the only possible unmatched pair of customer and server.

The Markov chain of unmatched couples is formed by two birth-death process connected in the empty state.



Performance Paradox Analysis



Let $\rho_1 = \frac{\alpha_1 \beta_3}{\alpha_2 (1 - \beta_3)}$ and $\rho_2 = \frac{\alpha_2 \beta_1}{\alpha_1 (1 - \beta_1)}$. If $\rho_1 < 1$ and $\rho_2 < 1$, the mean number of unmatched elements is

$$\frac{2(1 - \rho_1)(1 - \rho_2)}{1 - \rho_1 \rho_2} \left(\frac{\rho_1^2}{(1 - \rho_1)^2} + \frac{\rho_2^2}{(1 - \rho_2)^2} \right)$$

Performance Paradox Analysis

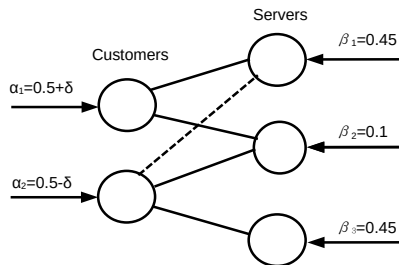
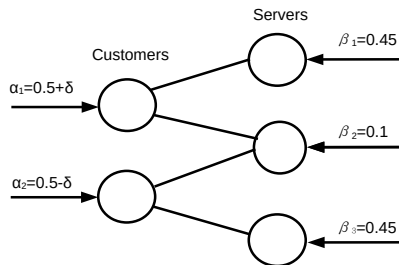
Objective: study the existence of the performance paradox

$$\frac{2(1 - \rho_1)(1 - \rho_2)}{1 - \rho_1\rho_2} \left(\frac{\rho_1^2}{(1 - \rho_1)^2} + \frac{\rho_2^2}{(1 - \rho_2)^2} \right) \stackrel{?}{>} \frac{2\rho_1}{1 - \rho_1}.$$

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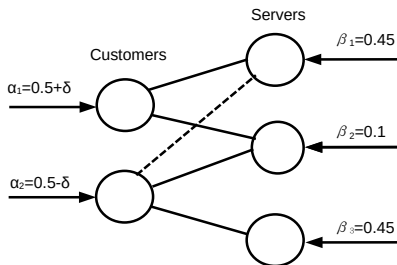
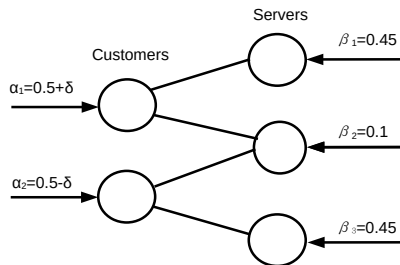


where $\delta \in (0, 0.05)$.

Performance Paradox Analysis

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$$\frac{2(1 - \rho_1)(1 - \rho_2)}{1 - \rho_1\rho_2} \left(\frac{\rho_1^2}{(1 - \rho_1)^2} + \frac{\rho_2^2}{(1 - \rho_2)^2} \right) \stackrel{?}{>} \frac{2\rho_1}{1 - \rho_1}.$$



where $\delta \in (0, 0.05)$.

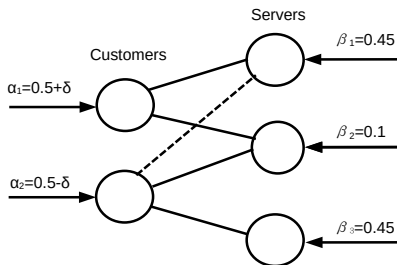
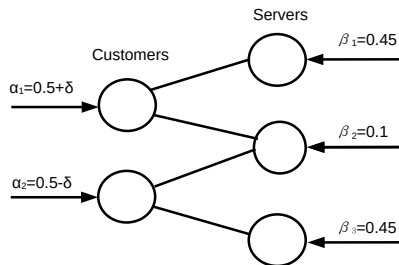
Stability

We check that $\beta_1 < \alpha_1$ and $\beta_3 < \alpha_2$ for all $\delta \in (0, 0.05)$

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We check that $\beta_1 < \alpha_1$ and $\beta_3 < \alpha_2$ for all $\delta \in (0, 0.05)$ and also that $\rho < 1$, $\rho_1 < 1$ and $\rho_2 < 1$.

Performance Paradox Analysis

Objective: study the existence of the performance paradox

$$\frac{99}{10} \frac{1 + 400\delta^2}{1 - 400\delta^2} \stackrel{?}{>} \frac{9(1 + 2\delta)}{1 - 20\delta}.$$

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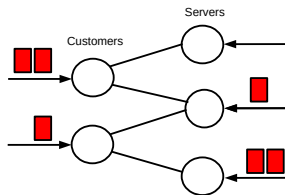
Proposition

When $\delta \rightarrow 0,05$, the difference due to the performance paradox tends to infinity.

Extensions

Matching policy: FCFM

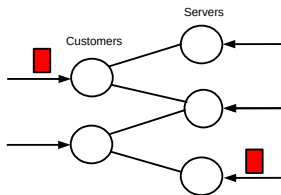
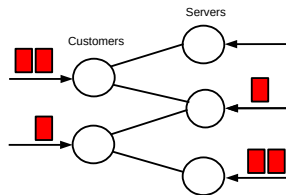
Other matching policies lead to the same Markov chains: MaxWeight



Extensions

Matching policy: FCFM

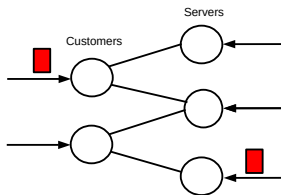
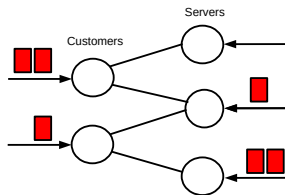
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The W-shaped compatibility graph

The same results are obtained for any compatibility graph which is complete minus two edges

Outline

- 1 Introduction
- 2 Model Description
- 3 Main Results
- 4 Conclusions**

Conclusions and Future Work

Adding an edge in bipartite matching models might hurt the performance of the system

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Adding an edge in bipartite matching models might hurt the performance of the system

Questions for future work

- Is the existence of a performance paradox related to this particular compatibility graph?
- Can we provide sufficient conditions on the existence of a performance paradox in arbitrary compatibility graphs (and FCFM)?
- Does the performance paradox exist in multigraphs? And when we consider self-loops?

Bipartite matching models:

- R. Caldentey, E. H. Kaplan, and G. Weiss. FCFS infinite bipartite matching of servers and customers. *Advances in Applied Probability*, 41(3):695–730, 2009.
- A. Busic, V. Gupta, and J. Mairesse. Stability of the bipartite matching model. *Advances in Applied Probability*, 45(2):351–378, 2013.

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Performance paradox in general matching models:

- A. Cadas, J. Doncel, J.-M. Fourneau, and A. Busic. Flexibility can hurt dynamic matching system performance. *ACM SIGMETRICS Performance Evaluation Review*, 49(3):37-42, 2022.
- A. Busic, A. Cadas, J. Doncel, and J.-M. Fourneau. Performance paradox of dynamic matching models under greedy policies. *Queueing Systems* 107, 257–293 (2024).

Thanks and questions?

THANKS FOR YOUR ATTENTION

QUESTIONS?