

Job migrations in queueing networks: some non-conventional product-forms

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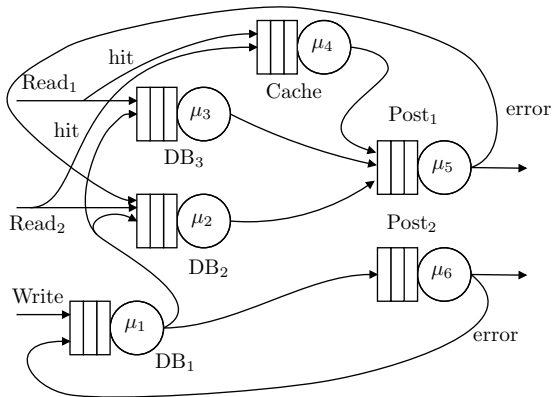
Two approaches from the literature

- The problem of dynamic load-balancing is widely studied in the literature
- Two major approaches
 - ▶ **Sender initiated**: a station with many jobs tries to send some of them to another station
 - ▶ **Receiver initiated**: a station with few jobs tries to serve some of the jobs of another station
- Receiver initiated strategies have been shown to work better in most practical cases
- Our approaches belong to the class of **receiver initiated strategies**

Section 2

LB-networks

Example: Stations with same functionalities



Job migration policy

- When station i is empty for a random amount of time α_{ij} , it polls station j
- Let t be the time epoch of polling and T_{ji} be the number of jobs moved from station j to i
- The distribution of $T_{ji}(t)$ is a truncated geometric with parameter b_{ij} :

$$Pr\{T_{ji}(t) = k\} = \begin{cases} (1 - b_{ij})b_{ij}^{n_j(t)} & \text{if } k < n_j(t) \\ b_{ij}^{n_j(t)} & \text{if } k = n_j(t) \\ 0 & \text{otherwise} \end{cases}$$

- We assume the job transfer time to be negligible w.r.t. the other delays
- When $b_{ij} = 1$ all the jobs present at time t in j are moved to i

The load factor of a queue

- Let ρ_i be the load factor of queue i

$$\rho_i = \frac{\sum_{\substack{k=1 \\ k \neq i}}^N x_{ki}^z + \sum_{\substack{k=1 \\ k \neq i}}^N x_{ki}^w + \lambda_i}{\mu_i + \sum_{\substack{k=1 \\ k \neq i}}^N x_{ki}^v b_{ki}}$$

- arrivals due to ordinary routing

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- arrivals due to ordinary routing
- arrivals due to migrations into station i
- exogenous arrivals

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- arrivals due to ordinary routing
- arrivals due to migrations into station i
- exogenous arrivals
- departures due to internal service
- departures due to job migration

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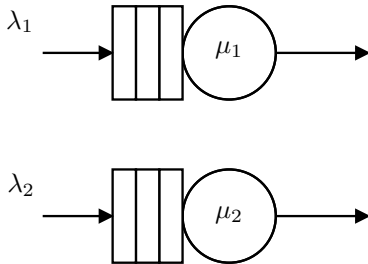
Properties of the solutions

Let $\chi_{ij}^v, \chi_{ij}^w, \chi_{ij}^z, q_i$ be a feasible solution of the rate equation system for the unknowns $x_{ij}^v, x_{ij}^w, x_{ij}^z, \rho_i$, respectively. Then, if $x_{ji}^v > 0$ we have:

$$q_j/q_i = b_{ji}$$

In other words, if $b_{ji} = 1$ is a feasible solution in which a station i polls a station j with a positive rate (when empty) the load factor of the two stations are the same

Two parallel queues



- We analyse the solutions of the system of rate equations

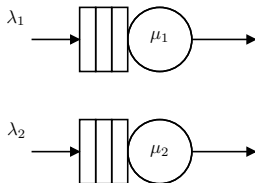
Solution for $b_{12} = b_{21} = 1$

- The system admits infinite non-trivial solutions that all give perfect load balance
- Without loss of generality, assume the unbalanced load factors $\bar{\rho}_2 > \bar{\rho}_1$
- We have that the perfect balance is obtained if the polling rates satisfy the following equation:

$$x_{12}^v = \frac{\lambda_2 \mu_1 - \lambda_1 \mu_2}{\lambda_1 + \lambda_2} + x_{21}^v$$

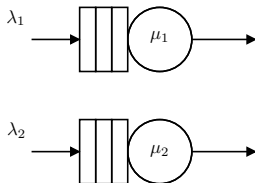
- As expected $\alpha_{12} = x_{12}^v > \alpha_{21} = x_{21}^v$
- Among the infinite solutions, in order to reduce the migration of jobs, one would choose the solution where $x_{21}^v = 0$

A numerical example



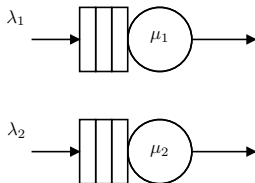
- $\lambda_1 = 0.8s^{-1}$, $\lambda_2 = 1.2s^{-1}$, $\mu_1 = 1s^{-1}$, $\mu_2 = 2s^{-1}$
- $\bar{\rho}_1 = 0.8$, $\bar{\rho}_2 = 0.6 \Rightarrow S_2$ polls S_1
- $\alpha_{21} = 0.2s^{-1}$ (polling rate)
- After load balancing $\rho_1 = \rho_2 = 2/3$
- No LB: Expected occupancy $\bar{N} = 5.5$, expected response time $\bar{R} = 2.75s$
- With LB: Expected occupancy $N = 4$, expected response time $R = 2s$

A numerical example



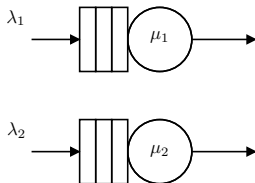
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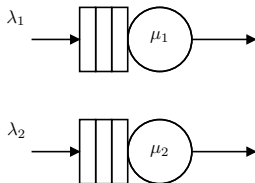
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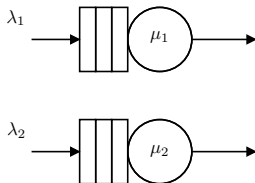
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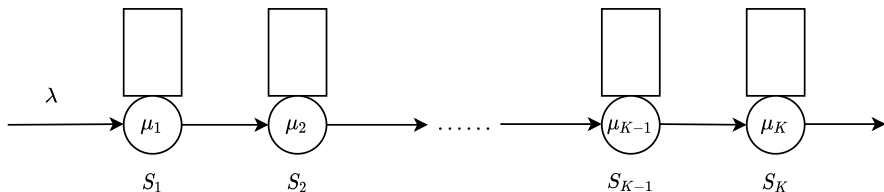
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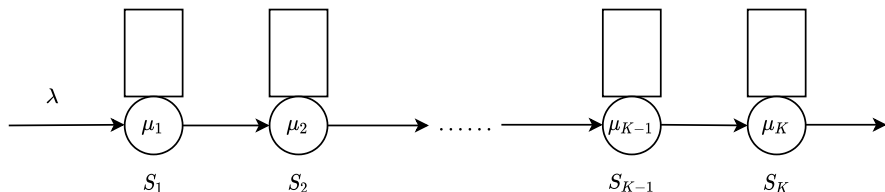
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Introduction



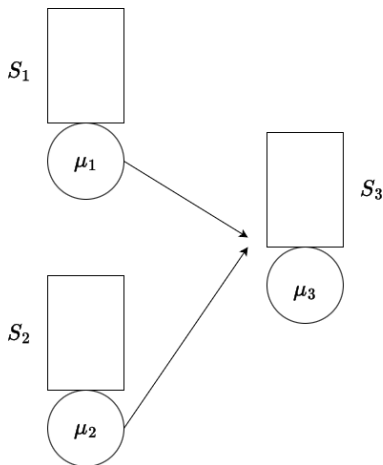
- If station S_i gets empty, it sends a signal to steal a job from S_{i-1} , $i > 1$
- This policy propagates backward until a non-empty station is found
- If the system is open, we can always fetch a job from the outside
- If the system is closed, thanks to the ergodicity of the routing process, we can always find a non-empty queue

Advantages



- In Jackson networks, the stability condition would be $\lambda < \min\{\mu_i\}$
- In this case, the stability condition is $\lambda < \mu_K$
- If we can order the stations $\mu_1 < \mu_2 < \dots < \mu_K$, the stability condition is $\lambda < \max\{\mu_i\}$ and the expected number of stages experienced by a job is maximized

Who do I choose?



- The provider is chosen with a probability that is proportional to the arrival stream from that station
- If S_3 gets empty, it chooses S_1 as a provider with probability $\mu_1/(\mu_1 + \mu_2)$

The product-form result

Let $X(t)$ be the irreducible CTMC underlying a fetching queueing network, then its invariant measure has product-form:

$$\pi(\mathbf{n}) \propto \prod_{i=1}^K \rho_i^{n_i},$$

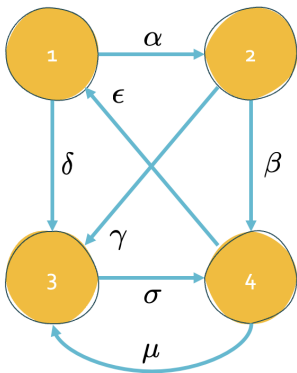
where:

$$\rho_i = \frac{\sum_{k=0}^K \mu_k p_{ki}}{\mu_i p_{i0} + \sum_{k=1}^K x_{ik}},$$

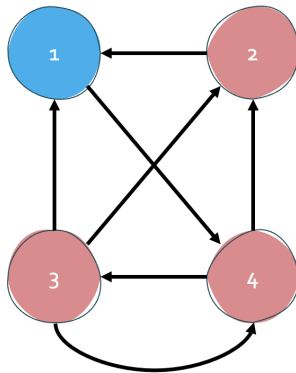
and x_{ik} is the solution the linear system of rate equations:

$$x_{ik} = \left(\sum_{j=1}^K x_{kj} + \mu_k p_{k0} \right) \frac{\mu_i p_{ik}}{\sum_{j=0}^K \mu_j p_{jk}}.$$

Reversing a process

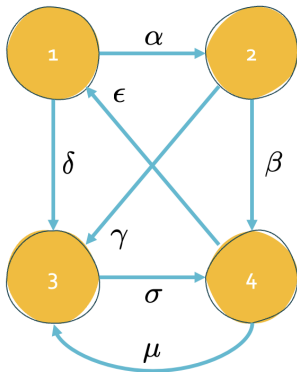


Forward

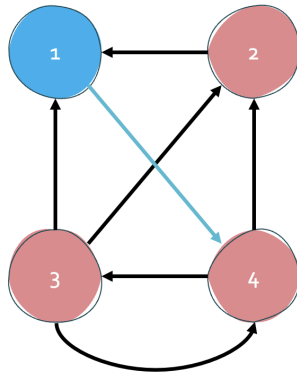


Reversed

Reversing a process



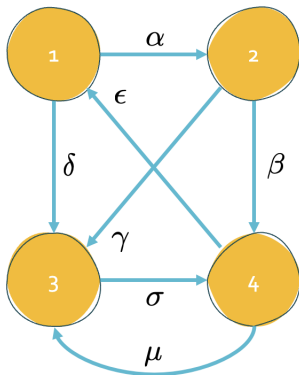
Forward



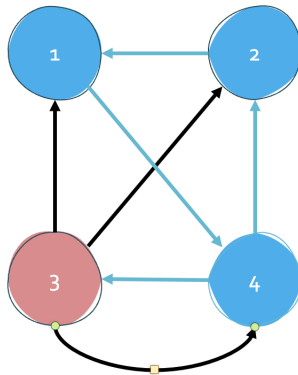
Reversed



Reversing a process

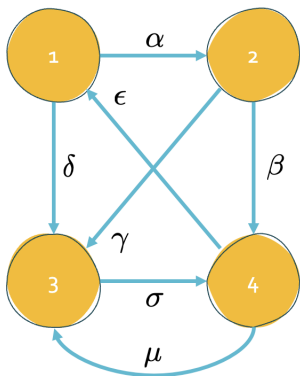


Forward

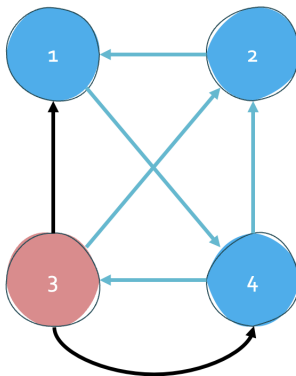


Reversed

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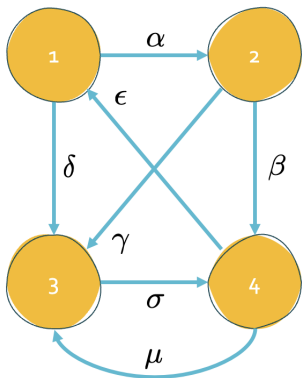


Forward

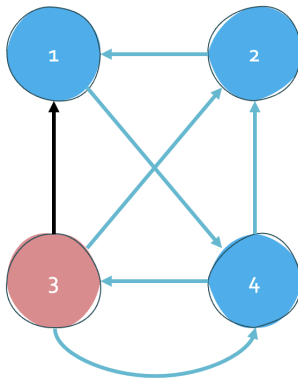


Reversed

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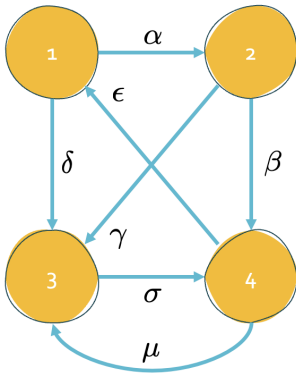


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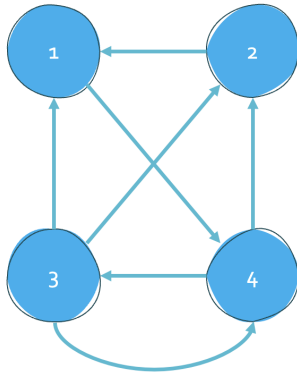


Reversed

Reversing a process

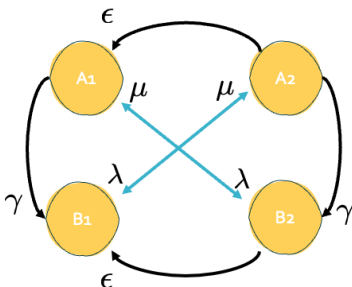
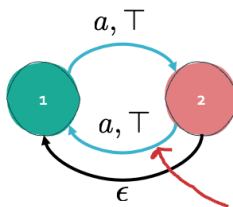
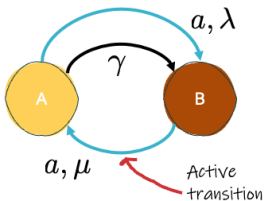


Forward

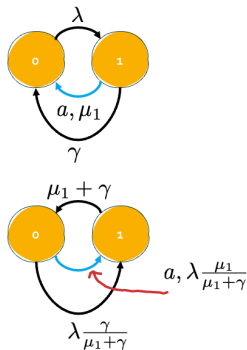


Reversed

PEPA-like Synchronization (labels)



Reversing labelled agents



- This is a reversible process
- The transition rate from 0 to 1 is λ
- The transition rate from 1 to 0 is $\gamma + \mu_1$
- The reversed rate splits among the parallel transition proportionally to the forward rate

RCAT methodology (Harrison, 2002)

Conditions

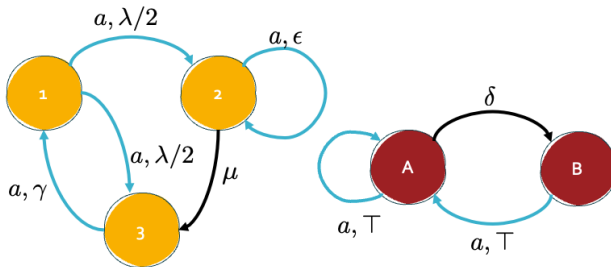
- 1 For each label a , if an agent is passive w.r.t. a , every state has an outgoing transition labelled a , named x_a
- 2 For each label a , if an agent is active w.r.t. a , every state has the same sum of reversed rates of the incoming transitions

Product-form: if the conditions are met, solve each process in isolation using the reversed rates x_a for each passive transitions a

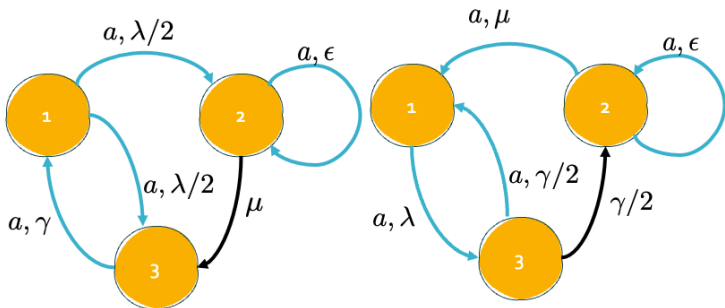
Tips:

- When verifying the second condition, assume the passive transitions have all the same rate
- Use Kolmogorov's criteria to simplify the reasoning

Find the product-form condition



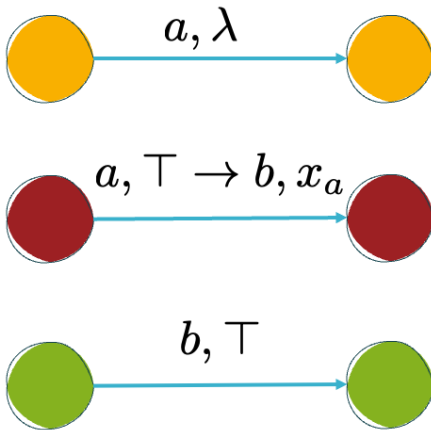
Find the product-form condition



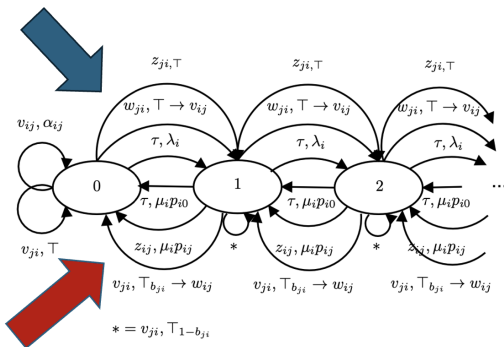
$$\lambda = \gamma/2 = \mu + \epsilon$$

Multiway synchronisations

Proposed in: *Peter G. Harrison, Andrea Marin: Product-Forms in Multi-Way Synchronizations. Comput. J. 57(11): 1693-1710 (2014)*



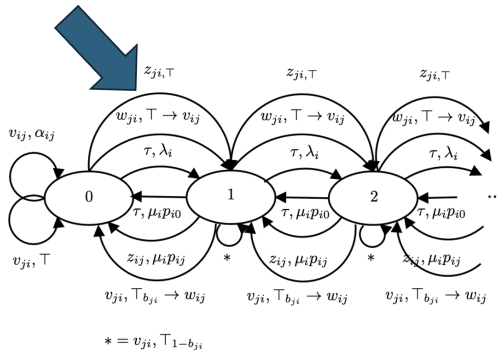
Proof of product-form for LB-networks



- for simplicity assume only station i and j
- Load factor:

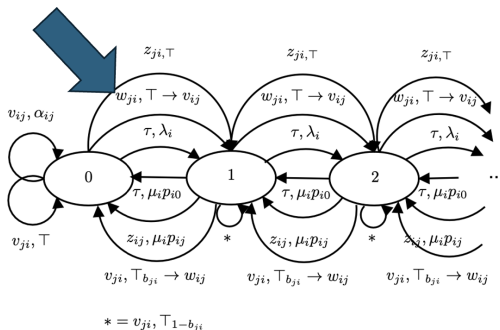
$$\rho_i = \frac{x_{ji}^z + x_{ji}^w + \lambda_i}{\mu_i + x_{ji}^v b_{ji}}$$

Proof of product-form for LB-networks



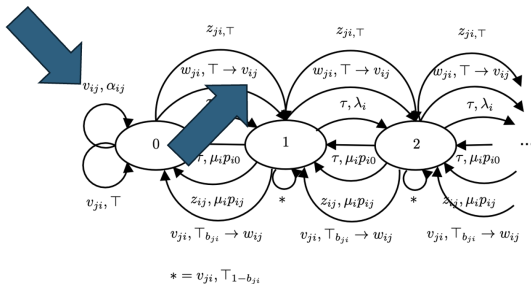
- z_{ij} : arrivals at station i from station j (passive)
- always enabled

Proof of product-form for LB-networks



- w_{ji} : arrivals at station i from migration from station j (passive)
- always enabled
- it propagates as v_{ij} , i.e., it requires another job if available

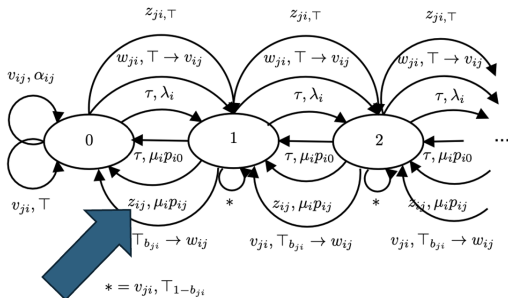
Proof of product-form for LB-networks



- v_{ij} : request for one job from i to j (active)
- reversed rate:

$$x_{ij}^v = \alpha_{ij} = \frac{x_{ji}^w}{\rho_i}$$

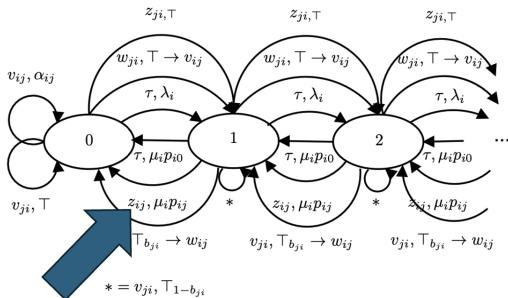
Proof of product-form for LB-networks



- z_{ij} : job completion at i moving to j (active)
- reversed rate:

$$x_{ij}^z = \mu_i p_{ij} \rho_i$$

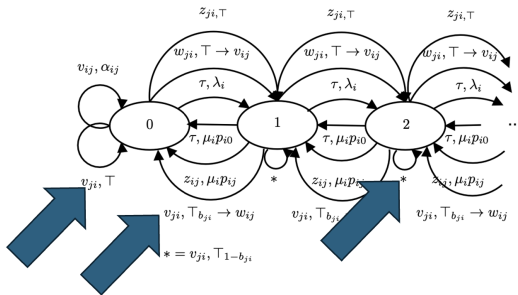
Proof of product-form for LB-networks



- z_{ij} : job completion at i moving to j (active)
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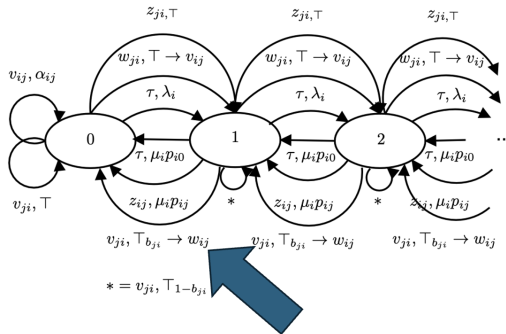
$$x_{ij}^z = \mu_i p_{ij} \rho_i$$

Proof of product-form for LB-networks



- v_{ji} : i receives a request of a job from j and decides to send it with probability b_{ji} (triggers w_{ij}) if any is present. If empty, it does not send anything. (passive)
- always enabled

Proof of product-form for LB-networks



- w_{ij} : i sends a job to j because of a request. (active)
- reversed rate:

$$x_{ij}^w = x_{ji}^v b_{ji} \rho_i$$

Final remarks

- Both queueing network models can be studied with Mean-Value approaches or convolution
- The MVA algorithms are obtained algebraically of thanks to the duality relation for fetching QNs
- RCAT proofs are very modular once you can model appropriately the interaction among the components
- **Open problems:** give formulations and proofs of the arrival theorems for LB-networks or fetching networks

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A caching system

- N objects
- The object is live when the state is $0, \dots, T - 1$, it is evicted when state is T
- Object i state n_i represents the age of the object
- The object ages from state n_i to $n_i + 1$ with rate γ_i , for $n_i < T$
- When an object receives a request:
 - ▶ If there is a **hit** then the state (age) is reset to 0
 - ▶ If there is a **miss**, the request is forwarded to another object j with probability $P(i, j)$ (in another cache) and the age of this newly retrieved object is set to the steady-state distribution of object i

Process algebraic representation

