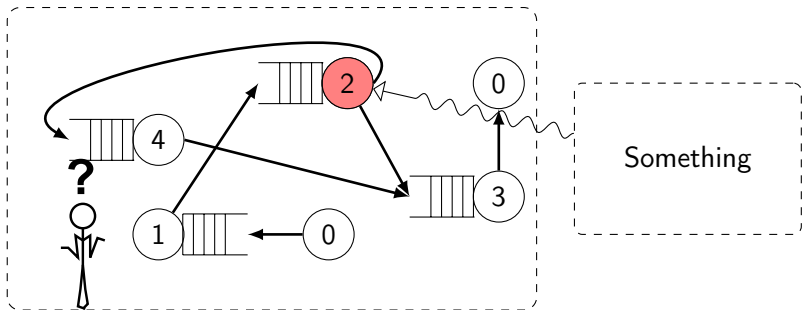


Jackson Network in a random environment

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Question

What does a served customer do when next node is degraded?

Answer

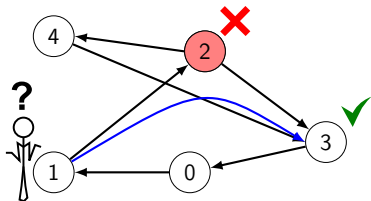
Change the route.

Random walk with random skipping

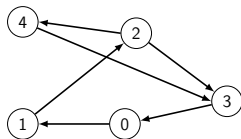
- Given a routing matrix r on a node set $\{0, 1, \dots, J\}$
- Given an acceptance probability vector $\alpha := (\alpha_0, \alpha_1, \dots, \alpha_J)$

Algorithm how to move from i

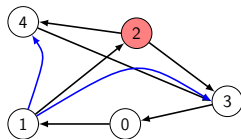
- 1 set $k = i$
- 2 move to j with probability $r(k, j)$
- 3 check if new node j will accept the customer with probability α_j
 - 3.1 if accepted, stay in j , END.
 - 3.2 if not accepted, skip: set $k = j$ go to 2



$$\alpha := (1, 1, 0.5, 1, 1)$$



(a) original routing r



(b) rerouting $r^{(\alpha)}$ by skipping

Figure: Skipping with $\alpha := (\alpha_0, \alpha_1, \alpha_2, \alpha_3, \alpha_4) = (1, 1, 0.5, 1, 1)$

rerouting matrix $r^{(\alpha)}$

$$r^{(\alpha)} = (I - r \cdot \text{diag}(1 - \alpha))^{-1} r \cdot \text{diag}(\alpha)$$

Given:

- Original routing matrix r
- Acceptance probability vector $\alpha := (\alpha_0, \alpha_1, \dots, \alpha_J)$
- Rerouting matrix $r^{(\alpha)}$

Nice Property of $r^{(\alpha)}$

If

$$\eta = (\eta_0, \dots, \eta_J)$$

solves

$$\eta r = \eta,$$

then

$$\eta^{(\alpha)} = (\alpha_0 \eta_0, \alpha_1 \eta_1, \dots, \alpha_J \eta_J)$$

solves

$$\eta^{(\alpha)} r^{(\alpha)} = \eta^{(\alpha)}.$$

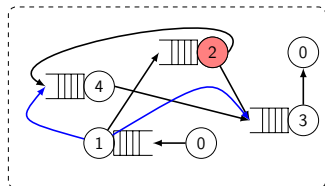
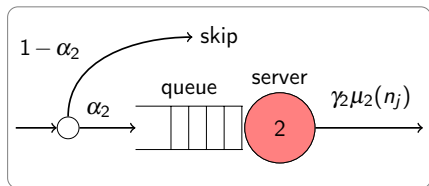
Modified Jackson network

Set of nodes $\bar{J} = \{1, \dots, J\}$, routing matrix r , service rates $\mu_1(n_1), \dots, \mu_J(n_J)$

Process: $\mathbf{X}(t) := (X(t) : t \in [0, \infty))$

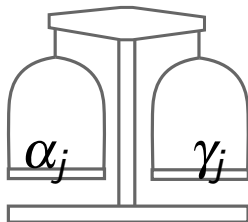
Modification

- + action: service rate changes by factor $\gamma_j \in [0, 1]$, $\gamma = (\gamma_1, \dots, \gamma_J)$
 \implies new service rates $\gamma_j \cdot \mu_j(n_j)$
- + reaction: rerouting $r^{(\alpha)}$



acceptance probability $\alpha_j = \gamma_j$

$$\alpha_j = \gamma_j$$



New service rate $\gamma_j \mu_j(n_j)$

	service rate factor γ_j	acceptance α_j
= 0	service down	everyone skips over
= 1	service up	everyone accepted
$0 < \cdot < 1$	degraded	sometimes accepted

Modified Jackson network

- J nodes, input rate λ , routing matrix r , service rates $\mu_j(n_j)$
- n_j is number of customers at the node j

Notation: $\mathbf{n} := (n_1, \dots, n_J)$, $\mathbf{e}_j := (0, \dots, 0, \overset{\text{on } j\text{-th pos.}}{\underbrace{1}}, 0, \dots, 0)$
Classical, with generator $Q^{\mathbf{X}} = (q^{\mathbf{X}}(\mathbf{n}, \mathbf{n}') : \mathbf{n}, \mathbf{n}' \in \mathbb{N}_0^J)$

$$q^{\mathbf{X}}(\mathbf{n}, \mathbf{n} + \mathbf{e}_i) = \lambda r(0, i)$$

$$q^{\mathbf{X}}(\mathbf{n}, \mathbf{n} - \mathbf{e}_j + \mathbf{e}_i) = 1_{[n_j > 0]} \mu_j(n_j) r(j, i)$$

$$q^{\mathbf{X}}(\mathbf{n}, \mathbf{n} - \mathbf{e}_j) = 1_{[n_j > 0]} \mu_j(n_j) r(j, 0)$$

Modification

- + action: service rate changes by $\gamma = (\gamma_1, \dots, \gamma_2)$
- + reaction: rerouting $r^{(\alpha)}$

New system

$$\begin{aligned} \dots &= \lambda r^{(\alpha)}(0, i) \\ \dots &= 1_{[n_j > 0]} \gamma_j \mu_j(n_j) r^{(\alpha)}(j, i) \\ \dots &= 1_{[n_j > 0]} \gamma_j \mu_j(n_j) r^{(\alpha)}(j, 0) \end{aligned}$$

Theorem (classical)

For the steady state distribution $\xi(\mathbf{n}) := \xi(n_1, \dots, n_J)$ holds

$$\xi(n_1, \dots, n_J) = \prod_{j=1}^J \xi(n_j) \quad \text{with} \quad \xi_j(n_j) := \prod_{k=1}^{n_j} \frac{\eta_j}{\mu_j(k)} C(j)^{-1}$$

with η solution of $\eta r = \eta$, $\eta_0 = \lambda$.

Blocked nodes $B(\gamma) := \{j \in \bar{J}, \gamma_j = 0\}$

Theorem (modified)

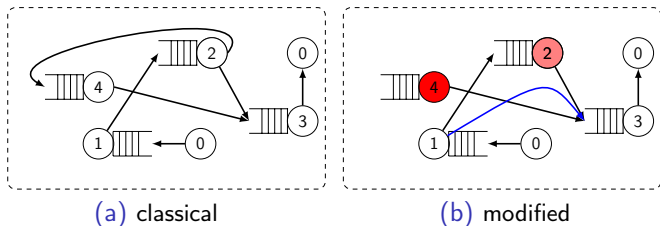
If acceptance probabilities $\equiv \alpha_j = \gamma_j \equiv$ service rate factors, then

$$\xi^{(\alpha)}(n_1, \dots, n_J) = \prod_{j \in \bar{J} \setminus B(\gamma)} \xi(n_j) \cdot \prod_{i \in B(\gamma)} \varphi_i(n_i)$$

with arbitrary distribution $\varphi_i(n_i)$.

Jackson network with degrading nodes

Example: $\gamma = (1, 0.5, 1, 0) \implies B(\gamma) = \{4\}$



If acceptance probabilities $\equiv \alpha_j = \gamma_j \equiv$ service rate factors, then steady state distribution ξ is

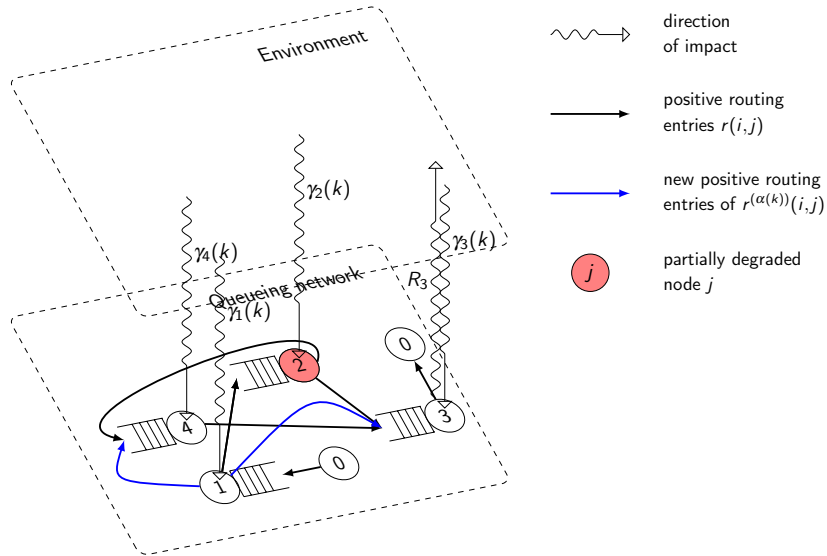
- the same on nodes 1, 2 and 3,
- arbitrary on 4 in the degraded case.

Environment: additional parameter of the system

- $\mathbf{Y} = (Y(t) : t \in [0, \infty))$
- Lives on finite state space K
- Can change its state with transition rates $V = (v(k, m) : k, m \in K)$, independent from queueing system
- Can be changed by queueing system with probabilities $R_j = (R(k, m) : k, m \in K)$, each time a customer from j leaves the network

$Y(t)$ is **not** necessarily a Markov process!

Jackson network in a random environment



Jackson network in a random environment

Given:

- \mathbf{X} -Jackson network with J nodes and parameters λ, μ, r
- \mathbf{Y} -environment on a state space K and rates $V = (v(k, m) : k, m \in K)$
- Interactions if environment is in state k :
 - Environment \rightarrow Queues: by changing of service rates by $\gamma(k)$, and therefore changing of rerouting $r^{(\alpha)}$, $\alpha = \gamma$
 - Queues \rightarrow Environment: by changing environment states with $R_j = (R(k, m) : k, m \in K)$, when a customer leaves the network from j

Steady state distribution $\pi(\mathbf{n}, k)$

$$\pi(\mathbf{n}, k) = \xi(\mathbf{n}) \cdot \theta(k), \quad \mathbf{n} \in \mathbb{N}_0^{\bar{J}}, k \in K,$$

$\xi(\mathbf{n})$ Solution for the original Jackson network,

$\theta(k)$ Stochastic solution of $\theta \cdot Q_{red} = 0$,

$$Q_{red} := \left[V + \sum_{j \in \bar{J}} \eta_j \text{diag} \left(r^{(\alpha(\cdot))}(j, 0) \right) (R_j - I) \right].$$

General service rate factor $\gamma = (\gamma_1, \dots, \gamma_J)$ with $\gamma_j \in [0, \infty)$

+ action: change of service rates by $\gamma = (\gamma_1, \dots, \gamma_2)$
 \implies new service rates $\gamma_j \cdot \mu_j(n_j)$

+ reaction: increase overall input $\beta := \begin{cases} 1 & \text{if } \|\gamma\|_\infty \leq 1 \\ \|\gamma\|_\infty & \text{if } \|\gamma\|_\infty > 1 \end{cases}$,

+ reaction: rerouting $r^{(\alpha)}$ with $\alpha_j := \begin{cases} \gamma_j & \text{if } \|\gamma\|_\infty \leq 1 \\ \frac{\gamma_j}{\|\gamma\|_\infty} & \text{if } \|\gamma\|_\infty > 1 \end{cases}$.

$$q^{\mathbf{X}}(\mathbf{n}, \mathbf{n} + \mathbf{e}_i) = \beta \lambda r^{(\alpha)}(0, i)$$

$$q^{\mathbf{X}}(\mathbf{n}, \mathbf{n} - \mathbf{e}_j + \mathbf{e}_i) = 1_{[n_j > 0]} \gamma_j \mu_j(n_j) r^{(\alpha)}(j, i)$$

$$q^{\mathbf{X}}(\mathbf{n}, \mathbf{n} - \mathbf{e}_j) = 1_{[n_j > 0]} \gamma_j \mu_j(n_j) r^{(\alpha)}(j, 0)$$

The same results for the steady state distribution

Other rerouting strategy $r^{(\alpha)}$, for example randomized reflection.
Or any rerouting $r^{(\alpha)}$ for which the **Nice Property** holds:

If $\eta = (\eta_0, \dots, \eta_J)$ solves

$$\eta r = \eta,$$

then $\eta^{(\alpha)} = (\alpha_0 \eta_0, \alpha_1 \eta_1, \dots, \alpha_J \eta_J)$ solves

$$\eta^{(\alpha)} r^{(\alpha)} = \eta^{(\alpha)}.$$

The same results for the steady state distribution

Nice Property of skipping

$$r^{(\alpha)} := (I - r \cdot \text{diag}(1 - \alpha))^{-1} r \cdot \text{diag}(\alpha)$$

If $\eta = (\eta_0, \dots, \eta_J)$ solves

$$\eta r = \eta,$$

then $\eta^{(\alpha)} = (\alpha_0 \eta_0, \alpha_1 \eta_1, \dots, \alpha_J \eta_J)$ solves

$$\eta^{(\alpha)} r^{(\alpha)} = \eta^{(\alpha)}.$$

- How can we use this property above in importance sampling?
- Applications for the new product form results for Jackson network + Environment.

Thank you for your attention!



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Randomization for Markov chains with applications to networks in a random environment.

July 2014.

<http://arxiv.org/abs/1407.8378>