

Stationary distribution of the tandem fluid queue and its application* to the accumulated priority queue

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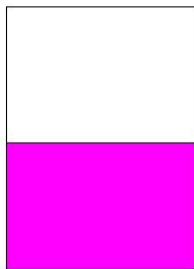
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Outline

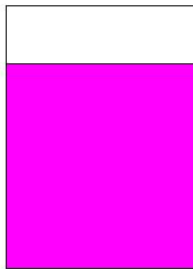
- 1 Tandem Fluid Queue (TFQ) Model
- 2 Analysis and numerical scheme
- 3 Application to Accumulated Priority Queue*
- 4 References

TFQ Model: two fluid queues driven by $\varphi(t)$

- CTMC $\{\varphi(t) : t \geq 0\}$ with finite state space S , generator \mathbf{T}
- Two fluid queues, contents $X(t)$ and $Y(t)$, both $\in [0, \infty)$



Buffer X



Buffer Y

First queue $X(t)$ driven by $\varphi(t)$

- $(\varphi(t), X(t))$ is standard fluid queue
- Fluid rates in $\mathbf{R} = \text{diag}(r_i)_{i \in \mathcal{S}}$

$$\begin{aligned} \frac{d}{dt}X(t) &= r_{\varphi(t)} && \text{when } X(t) > 0, \\ \frac{d}{dt}X(t) &= \max(0, r_{\varphi(t)}) && \text{when } X(t) = 0. \end{aligned}$$

- $\mathcal{S} = \mathcal{S}_+ \cup \mathcal{S}_- \cup \mathcal{S}_0$, e.g. $\mathcal{S}_+ = \{i \in \mathcal{S} : r_i > 0\}$
(upstates, downstates, zero-states)
- also: $\mathcal{S}_\ominus = \mathcal{S}_- \cup \mathcal{S}_0$ (“zero-states at $X(t) = 0$ ”)
- after ordering,

$$\mathbf{T} = \begin{bmatrix} \mathbf{T}_{++} & \mathbf{T}_{+-} & \mathbf{T}_{+0} \\ \mathbf{T}_{-+} & \mathbf{T}_{--} & \mathbf{T}_{-0} \\ \mathbf{T}_{0+} & \mathbf{T}_{0-} & \mathbf{T}_{00} \end{bmatrix}.$$

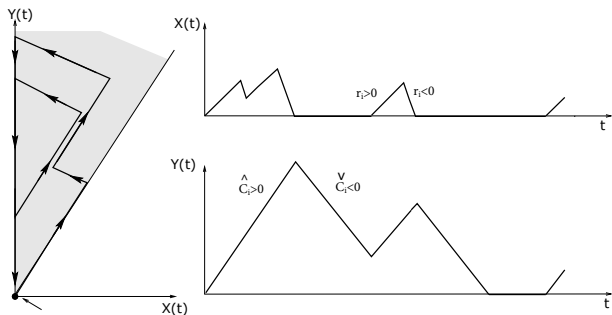
Second queue $Y(t)$ driven by $(\varphi(t), X(t))$

- $\widehat{\mathbf{C}} = \text{diag}(\widehat{c}_i)_{i \in \mathcal{S}}$, $\widehat{c}_i > 0$, and
- $\check{\mathbf{C}} = \text{diag}(\check{c}_i)_{i \in \mathcal{S}_\Theta}$, $\check{c}_i < 0$.
- $Y(t)$ increases when $X(t) > 0$, at rate $\widehat{c}_{\varphi(t)}$
- $Y(t)$ decreases when $X(t) = 0$, at rate $\check{c}_{\varphi(t)}$ (unless $Y(t) = 0$).

That is,

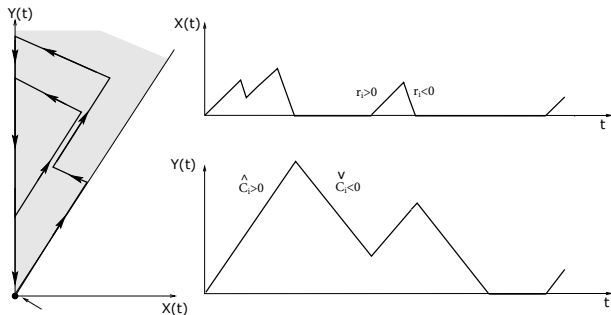
$$\begin{aligned} \frac{d}{dt} Y(t) &= \widehat{c}_{\varphi(t)} > 0 && \text{when } X(t) > 0, \\ \frac{d}{dt} Y(t) &= \check{c}_{\varphi(t)} < 0 && \text{when } X(t) = 0, Y(t) > 0, \\ \frac{d}{dt} Y(t) &= \widehat{c}_{\varphi(t)} \cdot \mathbf{1}\{\varphi(t) \in \mathcal{S}_+\} && \text{when } X(t) = 0, Y(t) = 0. \end{aligned}$$

Qualitative behaviour

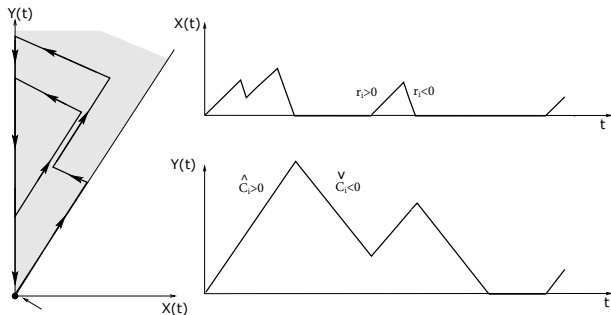


Assuming stability (see paper) process $(\varphi(t), X(t), Y(t))$ alternates between:

- (i) periods on $x = 0$
- (ii) periods on $x > 0$

Qualitative behaviour (i) on $x = 0$ (i) periods on $x = 0$

- ▶ $Y(t)$ decreasing, unless at $x = 0, y = 0$
- ▶ $\varphi(t)$ in \mathcal{S}_\ominus
- ▶ starts at $x = 0, y > 0$, with $\varphi(t)$ in \mathcal{S}_-
- ▶ ends at $x = 0, y \geq 0$, with $\varphi(t)$ jumping from \mathcal{S}_\ominus to \mathcal{S}_+

Qualitative behaviour (ii) on $x > 0$ (ii) periods on $x > 0$

- ▶ $Y(t)$ increasing (while $X(t)$ can either increase or decrease)
- ▶ $\varphi(t)$ in S (any phase)
- ▶ starts at $x = 0, y \geq 0$, with $\varphi(t) \in S_+$
- ▶ ends at $x = 0, y > 0$, with $\varphi(t) \in S_-$

Stationary distribution

has following form (all *vectors* with $|\mathcal{S}|$ components):

- (i)
 - ▶ 1-dimensional densities $\pi(0, y)$
at $x = 0, y > 0$
 - ▶ point masses $\mathbf{p}(0, 0)$
at $(0, 0)$

- (ii)
 - ▶ 2-dimensional densities $\pi(x, y)$
on $\{(x, y) : x > 0, y > x \cdot \min_{i \in \mathcal{S}_+} \{\widehat{c}_i/r_i\}\}$
 - ▶ 1-dimensional density $\pi^i(x, x\widehat{c}_i/r_i)$
on line $y = x\widehat{c}_i/r_i, i \in \mathcal{S}_+$

Approach

- Introduce embedded discrete-time process J_k
- Find its stationary distribution ξ_y
- Express $\pi(0, y)$ and $\mathbf{p}(0, 0)$ in ξ_y , using **down-shift** in Y
- Normalise based on knowledge of $(\varphi(t), X(t))$
- Express $\pi(x, y)$ in $\pi(0, y)$ and $\mathbf{p}(0, 0)$, using **up-shift** in Y
- Express $\pi^i(x, x\hat{c}_i/r_i)$ in $\mathbf{p}(0, 0)$

Mostly as LST's (but not always)

(i) down-shift: $\check{\mathbf{Q}}_{\ominus\ominus}$ and $\check{\mathbf{Q}}_{\ominus+}$

Let $D(t) = \int_{u=0}^t |\check{\mathbf{c}}_{\varphi(u)}| du$ and $t_z = \inf\{t > 0 : D(t) = z\}$.

Define

$$\check{\mathbf{Q}}_{\ominus\ominus} = (|\check{\mathbf{C}}_{\ominus}|)^{-1} \mathbf{T}_{\ominus\ominus}, \quad \check{\mathbf{Q}}_{\ominus+} = (|\check{\mathbf{C}}_{\ominus}|)^{-1} \mathbf{T}_{\ominus+}.$$

Then for $i, j \in \mathcal{S}_{\ominus}$, and $z > 0$,

$$[e^{\check{\mathbf{Q}}_{\ominus\ominus} z}]_{ij} = P(\varphi(t_z) = j, \varphi(u) \in \mathcal{S}_{\ominus}, 0 \leq u \leq t_z \mid \varphi(0) = i, X(0) = 0)$$

and $\check{\mathbf{Q}}_{\ominus+}$ is a matrix of transition rates (w.r.t. level) to phases in \mathcal{S}_+ (for times at which X and Y start increasing).

[Bean, O'Reilly and Taylor. Hitting probabilities and hitting times for stochastic fluid flows, *Stochastic Processes and their Applications*, 2005]

(ii) up-shift: $\widehat{\mathbf{Q}}(s)$ and $\widehat{\Psi}(s)$

Let $\theta = \inf\{t > 0 : X(t) = 0\}$ and $U(t) = \int_{u=0}^t \widehat{c}_{\varphi(u)} du$.

Then $U(\theta)$ is total up-shift in Y during Busy Period of X .

Its $|\mathcal{S}_+| \times |\mathcal{S}_-|$ density matrix $\widehat{\psi}(z)$ is given via LST

$$\widehat{\Psi}(s) = \int_{z=0}^{\infty} e^{-sz} \widehat{\psi}(z) dz$$

with

$$[\widehat{\Psi}(s)]_{ij} = E(e^{-sU(\theta)} \mathbf{1}\{\varphi(\theta) = j\} \mid \varphi(0) = i, X(0) = 0).$$

[Bean and O'Reilly. A stochastic two-dimensional fluid model, *Stochastic Models*, 2013]

(ii) up-shift: $\hat{\mathbf{Q}}(s)$ and $\hat{\Psi}(s)$

To find $\hat{\Psi}(s)$ define Key generator matrix

$$\hat{\mathbf{Q}}(s) = \begin{bmatrix} \hat{\mathbf{Q}}(s)_{++} & \hat{\mathbf{Q}}(s)_{+-} \\ \hat{\mathbf{Q}}(s)_{-+} & \hat{\mathbf{Q}}(s)_{--} \end{bmatrix}$$

$$\hat{\mathbf{Q}}(s)_{++} = (\mathbf{R}_+)^{-1} \left(\mathbf{T}_{++} - s\hat{\mathbf{C}}_+ - \mathbf{T}_{+0}(\mathbf{T}_{00} - s\hat{\mathbf{C}}_0)^{-1}\mathbf{T}_{0+} \right)$$

$$\hat{\mathbf{Q}}(s)_{+-} = (\mathbf{R}_+)^{-1} \left(\mathbf{T}_{+-} - \mathbf{T}_{+0}(\mathbf{T}_{00} - s\hat{\mathbf{C}}_0)^{-1}\mathbf{T}_{0-} \right)$$

$$\hat{\mathbf{Q}}(s)_{-+} = (|\mathbf{R}_-|)^{-1} \left(\mathbf{T}_{-+} - \mathbf{T}_{-0}(\mathbf{T}_{00} - s\hat{\mathbf{C}}_0)^{-1}\mathbf{T}_{0+} \right)$$

$$\hat{\mathbf{Q}}(s)_{--} = (|\mathbf{R}_-|)^{-1} \left(\mathbf{T}_{--} - s\hat{\mathbf{C}}_- - \mathbf{T}_{-0}(\mathbf{T}_{00} - s\hat{\mathbf{C}}_0)^{-1}\mathbf{T}_{0-} \right).$$

Then $\hat{\Psi}(s)$ is minimum nonnegative solution of Riccati eq.

$$\hat{\mathbf{Q}}(s)_{+-} + \hat{\mathbf{Q}}(s)_{++}\hat{\Psi}(s) + \hat{\Psi}(s)\hat{\mathbf{Q}}(s)_{--} + \hat{\Psi}(s)\hat{\mathbf{Q}}(s)_{-+}\hat{\Psi}(s) = \mathbf{O}.$$

[Bean and O'Reilly. A stochastic two-dimensional fluid model, *Stochastic Models*, 2013]

Embedded process J_k

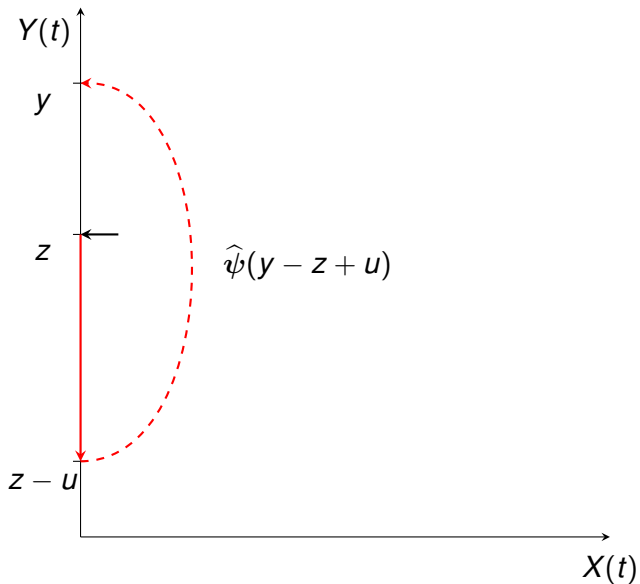
Let $J_k = (\varphi(\theta_k), Y(\theta_k))$ be with state space $\mathcal{S}_- \times (0, \infty)$, where θ_k is k -th time that $(\varphi(t), X(t), Y(t))$ hits $x = 0$.

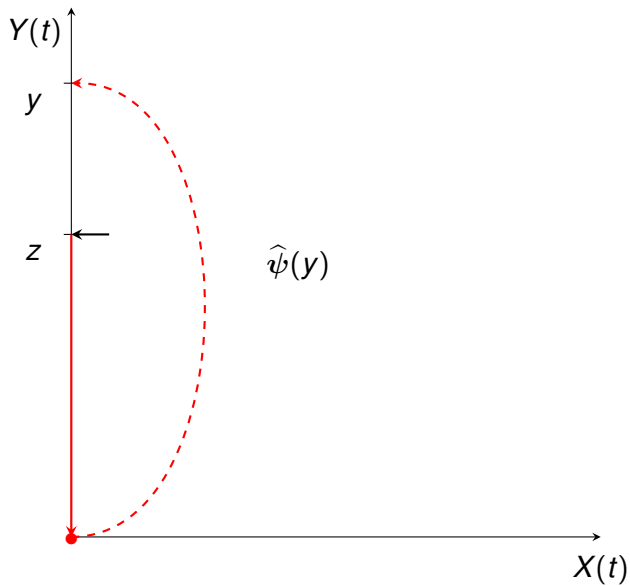
Lemma

The transition kernel of J_k is given by

$$\begin{aligned} \mathbf{P}_{z,y} = & \int_{u=[z-y]^+}^z [\mathbf{I} \quad \mathbf{0}] e^{\check{\mathbf{Q}}_{\ominus\ominus} u} \check{\mathbf{Q}}_{\ominus+}^{-1} \hat{\psi}(y - z + u) du \\ & + [\mathbf{I} \quad \mathbf{0}] e^{\check{\mathbf{Q}}_{\ominus\ominus} z} (-\check{\mathbf{Q}}_{\ominus\ominus})^{-1} \check{\mathbf{Q}}_{\ominus+} \hat{\psi}(y) \end{aligned}$$

where $[x]^+$ denotes $\max(0, x)$, and $[\mathbf{I} \quad \mathbf{0}]$ is a $|\mathcal{S}_-| \times |\mathcal{S}_\ominus|$ matrix.

Embedded process J_k 

Embedded process J_k 

Embedded process J_k

Corollary

The Laplace-Stieltjes transform of $\mathbf{P}_{z,y}$ w.r.t. y is given by

$$\begin{aligned} \mathbf{P}_{z,\cdot}(s) &= \begin{bmatrix} \mathbf{I} & \mathbf{O} \end{bmatrix} e^{-sz} \left(\check{\mathbf{Q}}_{\ominus\ominus} + s\mathbf{I} \right)^{-1} \left(e^{(\check{\mathbf{Q}}_{\ominus\ominus} + s\mathbf{I})z} - \mathbf{I} \right) \\ &\quad \times \check{\mathbf{Q}}_{\ominus+} \hat{\Psi}(s) \\ &+ \begin{bmatrix} \mathbf{I} & \mathbf{O} \end{bmatrix} e^{\check{\mathbf{Q}}_{\ominus\ominus}z} \left(-\check{\mathbf{Q}}_{\ominus\ominus} \right)^{-1} \check{\mathbf{Q}}_{\ominus+} \hat{\Psi}(s). \end{aligned}$$

Embedded process J_k

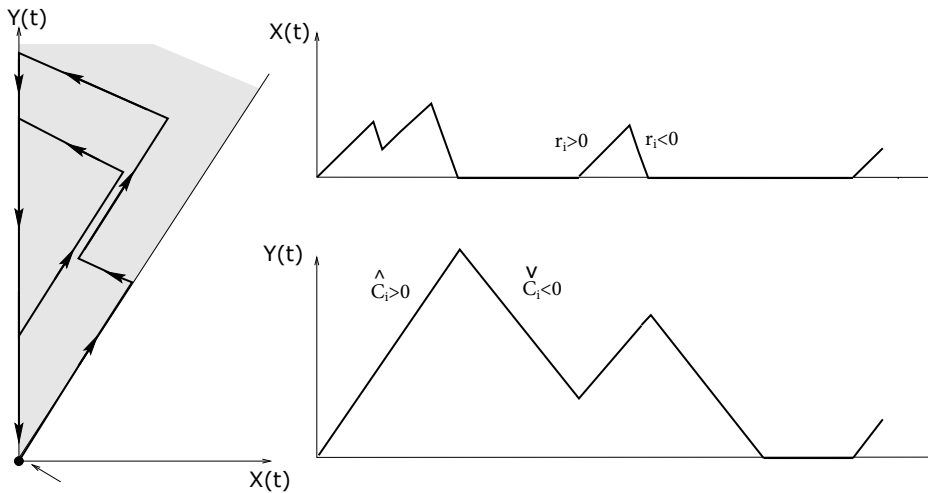
Stationary distribution of J_k is given by row vector $\xi_z = [\xi_{i,z}]_{i \in \mathcal{S}_-}$ of densities, satisfying

$$\begin{cases} \int_{z=0}^{\infty} \xi_z \mathbf{P}_{z,y} dz & = \xi_y \\ \int_{y=0}^{\infty} \xi_y dy \mathbf{1} & = 1 \end{cases}$$

will be solved numerically.

Next step:

Express stationary distribution of $(\varphi(t), X(t), Y(t))$ at level $x = 0$ in terms of ξ_z .

Expressing $\pi(0, y)$ and $\mathbf{p}(0, 0)$ in ξ_y 

Expressing $\pi(0, y)$ and $\mathbf{p}(0, 0)$ in ξ_y

Lemma

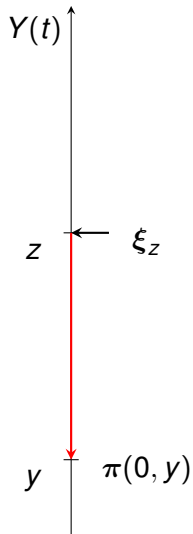
We have $\pi(0, y) = [\mathbf{0} \quad \pi(0, y)_\ominus]$, where

$$\pi(0, y)_\ominus = \alpha \int_{z=y}^{\infty} [\xi_z \quad \mathbf{0}] e^{\check{\mathbf{Q}}_{\ominus\ominus}(z-y)} (|\check{\mathbf{C}}_{\ominus}|)^{-1} dz,$$

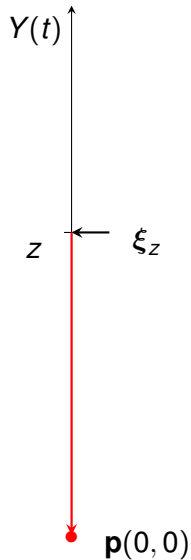
and $\mathbf{p}(0, 0) = [\mathbf{0} \quad \mathbf{p}(0, 0)_\ominus]$, where

$$\mathbf{p}(0, 0)_\ominus = \alpha \int_{z=0}^{\infty} [\xi_z \quad \mathbf{0}] e^{\check{\mathbf{Q}}_{\ominus\ominus}z} dz (-\mathbf{T}_{\ominus\ominus})^{-1}.$$

Here, α is a normalizing constant and the total rate of hitting $x = 0$.

Expressing $\pi(0, y)$ and $\mathbf{p}(0, 0)$ in ξ_y 

or



Expressing $\pi(0, y)$ and $\mathbf{p}(0, 0)$ in ξ_y

Define LST of density part

$$\pi(0, \cdot)(s) = \int_{z=0}^{\infty} e^{-sy} \pi(0, y) dy.$$

Corollary

We have $\pi(0, \cdot)(s) = [\mathbf{0} \quad \pi(0, \cdot)(s)_{\ominus}]$, where

$$\begin{aligned} \pi(0, \cdot)(s)_{\ominus} &= \alpha \int_{z=0}^{\infty} [\xi_z \quad \mathbf{0}] e^{\check{\mathbf{Q}}_{\ominus\ominus} z} (\check{\mathbf{Q}}_{\ominus\ominus} + s\mathbf{I})^{-1} \\ &\quad \times \left(\mathbf{I} - e^{-(\check{\mathbf{Q}}_{\ominus\ominus} + s\mathbf{I})z} \right) (|\check{\mathbf{C}}_{\ominus}|)^{-1} dz. \end{aligned}$$

Normalise, based on 1-dim fluid queue $(\varphi(t), X(t))$

Lemma

The normalising constant α is given by

$$\alpha = \left\{ \begin{aligned} & [\boldsymbol{\xi} \quad \mathbf{0}] (-\mathbf{T}_{\ominus\ominus})^{-1} \left(\mathbf{1} \right. \\ & \left. + \mathbf{T}_{\ominus+} \mathbf{K}^{-1} [(\mathbf{R}_+)^{-1} \quad \boldsymbol{\Psi} (|\mathbf{R}_-|)^{-1}] \right. \\ & \left. \times \left(\mathbf{1} + \mathbf{T}_{\pm\circ} (-\mathbf{T}_{\circ\circ})^{-1} \mathbf{1} \right) \right\}^{-1}, \end{aligned}$$

where, $\boldsymbol{\xi} = \int_{z=0}^{\infty} \boldsymbol{\xi}_z dz$, $\boldsymbol{\Psi} = \widehat{\boldsymbol{\Psi}}(s)|_{s=0}$ and $\mathbf{K} = \widehat{\mathbf{K}}(s)|_{s=0}$ with

$$\widehat{\mathbf{K}}(s) = \widehat{\mathbf{Q}}(s)_{++} + \widehat{\boldsymbol{\Psi}}(s) \widehat{\mathbf{Q}}(s)_{-+}.$$

Normalise, based on 1-dim fluid queue $(\varphi(t), X(t))$

Proof. Integrating $\pi(0, y)$ and adding $\mathbf{p}(0, 0)$ yields the probability mass vector of $\varphi(t)$ at $x = 0$,

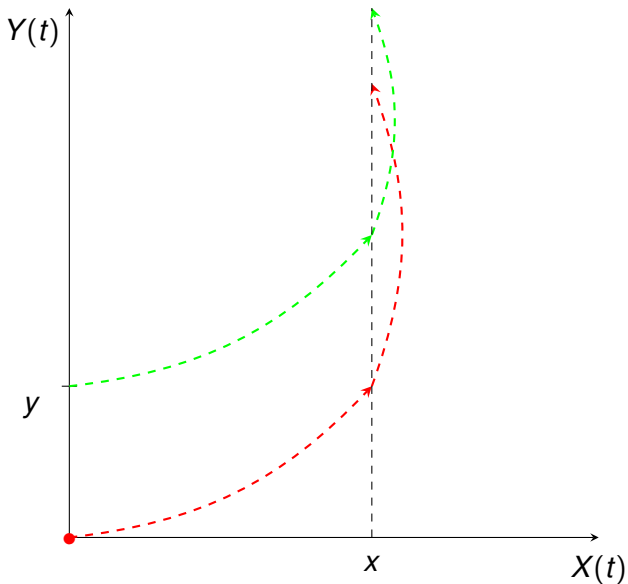
$$\begin{bmatrix} \mathbf{p}_- & \mathbf{p}_0 \end{bmatrix} = \alpha \begin{bmatrix} \boldsymbol{\xi} & \mathbf{0} \end{bmatrix} (-\mathbf{T}_{\ominus\ominus})^{-1}.$$

Similarly, we have expression for density $\pi(x)$ at $x > 0$,

$$\begin{aligned} \begin{bmatrix} \pi(x)_+ & \pi(x)_- \end{bmatrix} &= \begin{bmatrix} \mathbf{p}_- & \mathbf{p}_0 \end{bmatrix} \mathbf{T}_{\ominus+} e^{\mathbf{K}x} \begin{bmatrix} (\mathbf{R}_+)^{-1} & \boldsymbol{\Psi}(|\mathbf{R}_-|)^{-1} \end{bmatrix}, \\ \pi(x)_0 &= \begin{bmatrix} \pi(x)_+ & \pi(x)_- \end{bmatrix} \mathbf{T}_{\pm 0} (-\mathbf{T}_{\circ\circ})^{-1}. \end{aligned}$$

Now solve α from

$$\mathbf{p}\mathbf{1} + \int_{x=0}^{\infty} \pi(x) dx \mathbf{1} = 1.$$

Expressing $\pi(x, y)$ in $\pi(0, y)$ and $\mathbf{p}(0, 0)$ 

Expressing $\pi(x, y)$ in $\pi(0, y)$ and $\mathbf{p}(0, 0)$

Lemma

We have

$$\pi(x, \cdot)(s) = [\pi(x, \cdot)(s)_+ \quad \pi(x, \cdot)(s)_- \quad \pi(x, \cdot)(s)_0]$$

with

$$\begin{aligned} [\pi(x, \cdot)(s)_+ \quad \pi(x, \cdot)(s)_-] &= (\pi(0, \cdot)(s)_\ominus + \mathbf{p}(0, 0)_\ominus) \\ &\times \mathbf{T}_{\ominus+} e^{\hat{\mathbf{K}}(s)x} \times [(\mathbf{R}_+)^{-1} \quad \hat{\Psi}(s)(|\mathbf{R}_-|)^{-1}], \end{aligned}$$

and

$$\begin{aligned} \pi(x, \cdot)(s)_0 &= [\pi(x, \cdot)(s)_+ \quad \pi(x, \cdot)(s)_-] \\ &\times \mathbf{T}_{\pm 0} (s\hat{\mathbf{C}}_0 - \mathbf{T}_{00})^{-1}. \end{aligned}$$

Expressing $\pi(x, y)$ in $\pi(0, y)$ and $\mathbf{p}(0, 0)$

Let $\pi(\cdot, \cdot)(v, s) = \int_{x=0}^{\infty} e^{-vx} \pi(x, \cdot)(s) dx$.

Corollary

We have

$$\pi(\cdot, \cdot)(v, s) = \left[\pi(\cdot, \cdot)(v, s)_+ \quad \pi(\cdot, \cdot)(v, s)_- \quad \pi(\cdot, \cdot)(s)_o \right]$$

with

$$\begin{aligned} \left[\pi(\cdot, \cdot)(v, s)_+ \quad \pi(\cdot, \cdot)(v, s)_- \right] &= (\pi(0, \cdot)(s)_\ominus + \mathbf{p}(0, 0)_\ominus) \\ &\times \begin{bmatrix} \mathbf{T}_{-+} \\ \mathbf{T}_{o+} \end{bmatrix} (-\hat{\mathbf{K}}(s) + v\mathbf{I})^{-1} \left[(\mathbf{R}_+)^{-1} \quad \hat{\Psi}(s)(|\mathbf{R}_-|)^{-1} \right] \end{aligned}$$

and

$$\begin{aligned} \pi(\cdot, \cdot)(s)_o &= \left[\pi(\cdot, \cdot)(s)_+ \quad \pi(\cdot, \cdot)(s)_- \right] \mathbf{T}_{\pm o} \\ &\times (s\hat{\mathbf{C}}_o - \mathbf{T}_{oo})^{-1}. \end{aligned}$$

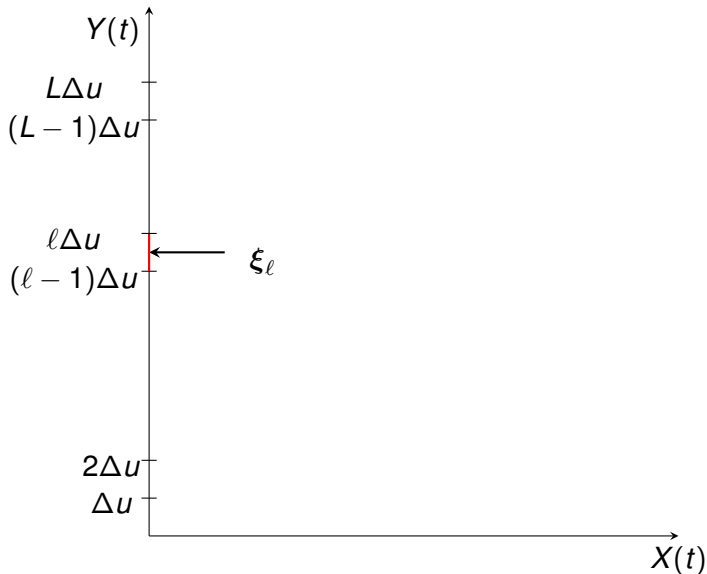
Expressing $\pi^i(x, x\hat{c}_i/r_i)$ in $\mathbf{p}(0, 0)$

Lemma

For all $i \in \mathcal{S}_+$,

$$\pi^i(x, x\hat{c}_i/r_i) = \sum_{j \in \mathcal{S}_\ominus} \mathbf{p}_j(0, 0) T_{ji} \exp(-(T_{ii}/r_i)x)/r_i.$$

Numerical scheme: $(l - 1)\Delta u \leq z \leq l\Delta u$



Numerical scheme

- Truncate and discretize the state space of J_k to get the DTMC $\{\bar{J}_k : k = 0, 1, 2, \dots\}$ with state space $\{(i, \ell) : i \in \mathcal{S}_-, \ell = 1, 2, \dots, L\}$

and matrix $\bar{\mathbf{P}} = [\bar{\mathbf{P}}_{\ell m}]_{\ell, m=0, 1, 2, \dots, L}$ made up of block matrices $\bar{\mathbf{P}}_{\ell m} = [\bar{P}_{i, \ell; j, m}]_{i, j \in \mathcal{S}_-}$, where

$$\bar{P}_{i, \ell; j, m} = P(\bar{J}_{k+1} = (j, m) \mid \bar{J}_k = (i, \ell)).$$

- Get $\bar{\mathbf{P}}$ by approximating

$$\bar{\mathbf{P}}_{\ell m} = \int_{y=(m-1)\Delta u}^{m\Delta u} \mathbf{P}_{\ell \Delta u, y} dy \approx \Delta u \mathbf{P}_{\ell \Delta u, m \Delta u},$$

and normalizing so that $\sum_{m=0}^L \bar{\mathbf{P}}_{\ell m} \mathbf{1} = \mathbf{1}$.

- Find $\bar{\xi}_\ell = [\bar{\xi}_{j; \ell}]_{j \in \mathcal{S}_-}$ by solving $\bar{\xi} \bar{\mathbf{P}} = \bar{\xi}$, $\bar{\xi} \mathbf{1} = \mathbf{1}$.

Numerical scheme

- Use this this to approximate

$$\mathbf{p}(0,0)_{\ominus} \approx \alpha \sum_{\ell=1}^L [\bar{\xi}_{\ell} \quad \mathbf{0}] e^{\check{\mathbf{Q}}_{\ominus\ominus} \ell \Delta u} (-\mathbf{T}_{\ominus\ominus})^{-1}$$

and

$$\begin{aligned} \pi(0, \cdot)(s)_{\ominus} &\approx \alpha \sum_{\ell=1}^L [\bar{\xi}_{\ell} \quad \mathbf{0}] e^{\check{\mathbf{Q}}_{\ominus\ominus} \ell \Delta u} (\check{\mathbf{Q}}_{\ominus\ominus} + s\mathbf{I})^{-1} \\ &\quad \times \left(\mathbf{I} - e^{-(\check{\mathbf{Q}}_{\ominus\ominus} + s\mathbf{I}) \ell \Delta u} \right) (|\check{\mathbf{C}}_{\ominus}|)^{-1}. \end{aligned}$$

- Evaluate $\pi(x, \cdot)(s)$ and invert using Abate and Whitt.

Numerical Example

We consider a process with the following parameters: $\mathcal{S} = \{1, 2\}$, $r_1 = 2$, $r_2 = -6$, $\hat{c}_1 = \hat{c}_2 = 2$, $\check{c}_1 = \check{c}_2 = -3$, and

$$\mathbf{T} = \begin{bmatrix} -2 & 2 \\ 1 & -1 \end{bmatrix}.$$

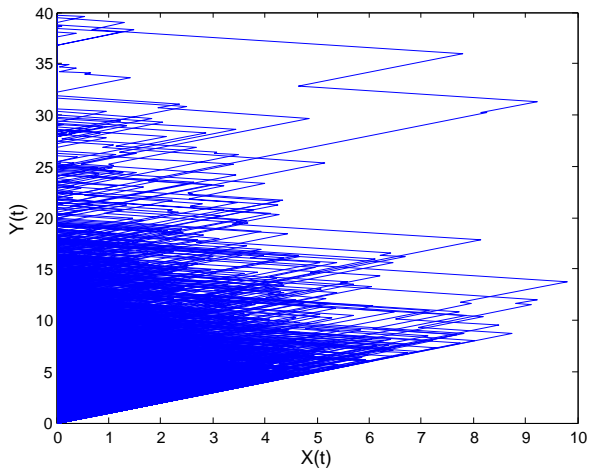
This simple process is similar to the model studied in Kroese and Scheinhardt (2011) and Werner (1998)

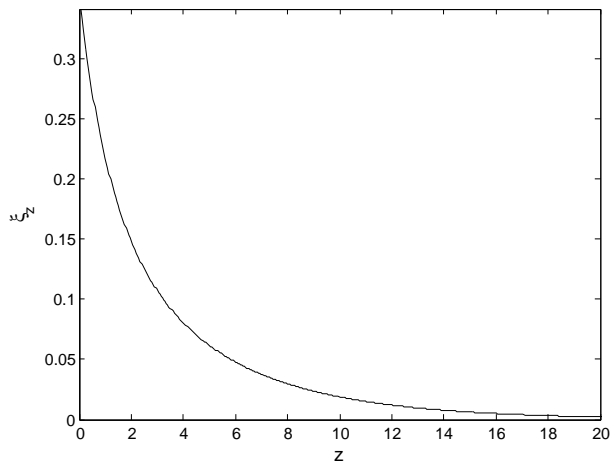
(but different from the numerical example analysed there).

[D.P. Kroese and W.R.W. Scheinhardt. Joint Distributions for Interacting Fluid Queues. *Queueing Systems*, 2001.]

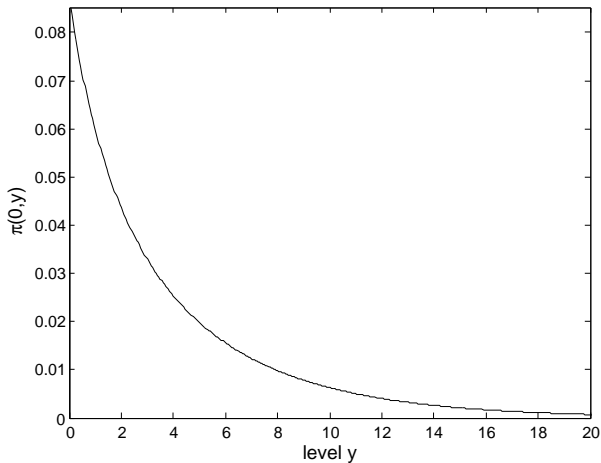
[W.R.W. Scheinhardt, PhD Thesis, 1998.]

Simulated values $(X(t), Y(t)), 0 \leq t \leq 10^5$

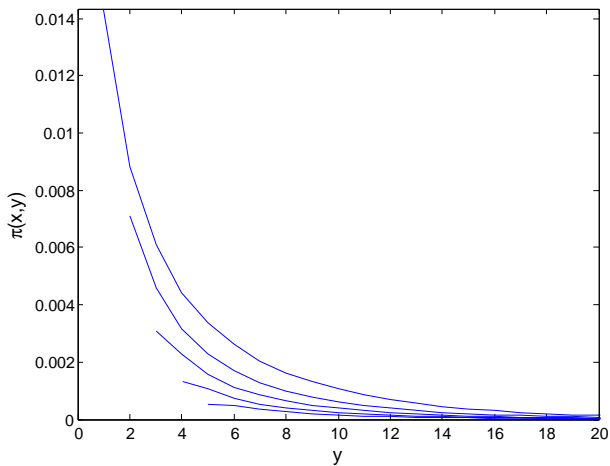


The estimated values $[\xi_z]_j$ for $j = 2$ 

The estimated values $[\pi(0, y)]_j$ for $j = 2$



The estimated values $[\pi(x, y)]_j$ for $j = 1, x = 1, \dots, 5$



Two-class Accumulating Priority Queue

- Single-server queue with PH service time
- Customer classes $i = 1, 2$
- Poisson arrivals with rates λ_i for $i = 1, 2$
- Class i customer accumulates priority at rate b_i
(Upon arrival) with $b_1 > b_2$
- Customer with the **highest accumulated priority** commences service (After completion of service).

Maximum Priority Process \mathbf{M}

- Let $M_i(t)$ be the **least upper bound** for all class i customers present in the queue at time t .
- Maximum Priority Process $\mathbf{M} = \{(M_1(t), M_2(t)); t \geq 0\}$.
- We are interested in the stationary distribution of \mathbf{M} **embedded** at the moments of commencement of service.

Result

- We map \mathbf{M} to a certain TFQ $\{(\varphi(t), \tilde{Z}(t), \tilde{M}_2(t)); t \geq 0\}$.
- The stationary distribution of $\{(\varphi(t), \tilde{Z}(t), \tilde{M}_2(t)); t \geq 0\}$ can be obtained using our results discussed above.
- The stationary distribution of \mathbf{M} embedded at the moments of commencement of service

is equivalent to the part of the stationary distribution of $\{(\varphi(t), \tilde{Z}(t), \tilde{M}_2(t)); t \geq 0\}$ corresponding to down-phase —.

[M. M. O'Reilly and W. R. W. Scheinhardt. Stationary distributions for a class of Markov-modulated tandem fluid queues. *Submitted to Stochastic Models*, 2016.]

Future work

- Numerical analysis for the accumulating priority queue.
- Analysis of the dual model.

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Thank you for listening!

