

New Stability and Diffusion Results for Multi-Class Queueing Networks

**Yongjiang Guo¹, Erjen Lefeber², Yoni Nazarathy³,
Gideon Weiss⁴, Hanqin Zhang⁵.**

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1 Beijing University of Post and Telecommunications

2 Eindhoven University of Technology

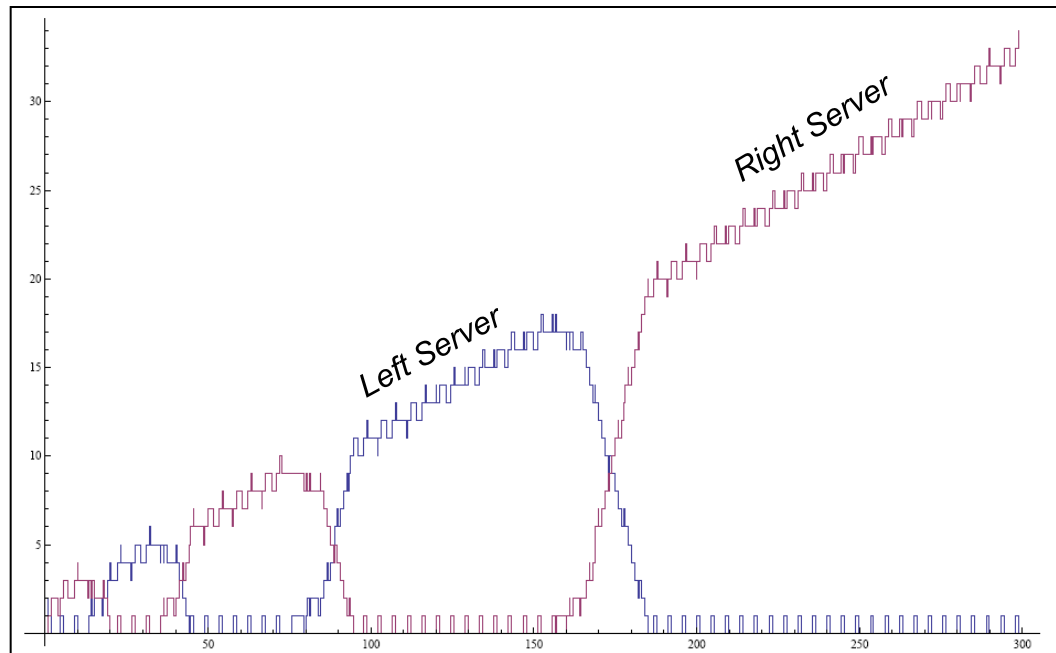
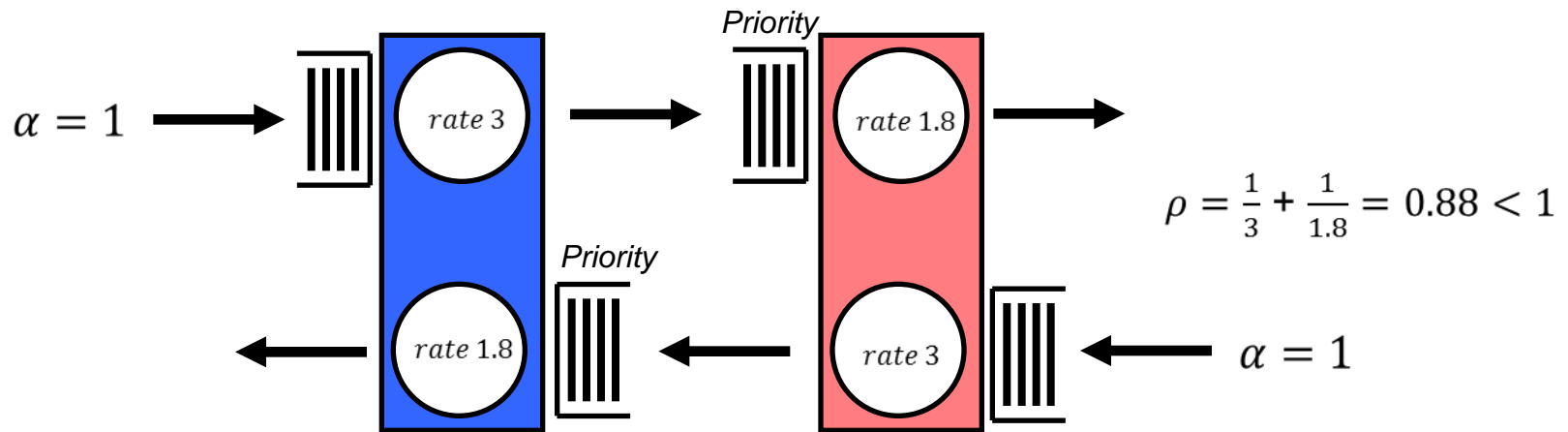
3 Swinburne University of Technology

4 The University of Haifa

5 National University of Singapore

INTRODUCTION: STABILITY OF MULTI-CLASS QUEUEING NETWORKS

The $K_{umar} S_{eidman} R_{ybko} S_{toylar}$ Queueing Network (90's)



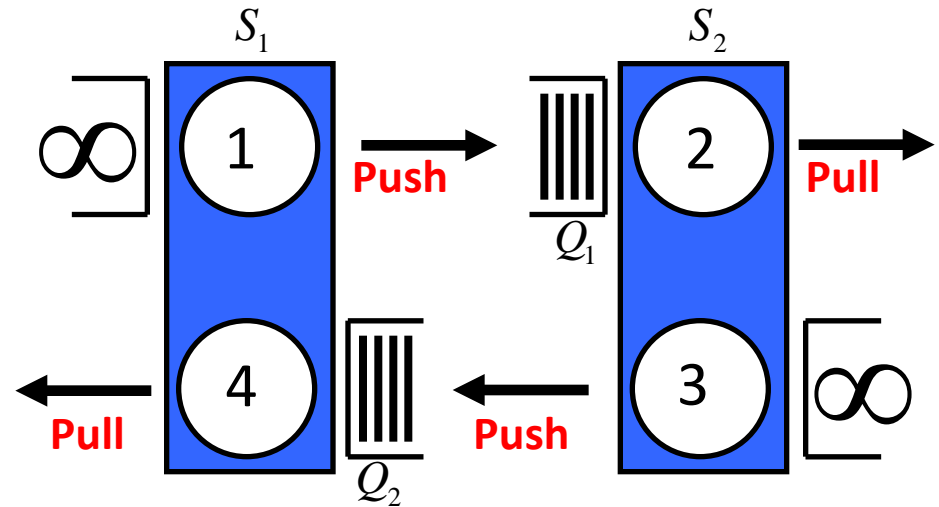
In this talk

- Outline past research on Multi-Class Queueing Networks with **Infinite Supplies**
 - *N., Weiss, 2008-2009*
- Overview of New Results
 - *Stability of certain examples*
 - *Diffusion Limits*
- Outlook
 - *Stability of Queueing Networks is “hard”*
 - *Need methods to construct general policies*

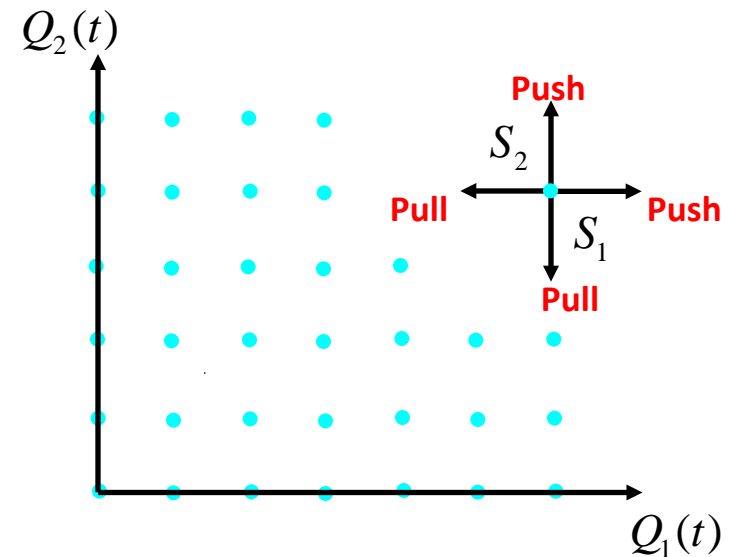
MULTI-CLASS QUEUEING NETWORKS WITH INFINITE SUPPLIES

Example: The Push-Pull Network

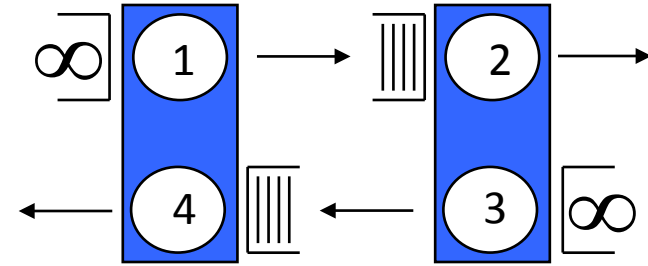
- 2 job streams, 4 steps
- Queues at pull operations
- Infinite job supply at 1 and 3
- 2 servers



- Control choice based on $Q_1(t), Q_2(t)$
- No idling, FULL UTILIZATION
- Preemptive resume



Processing Times



$$\xi_k = \{\xi_k^j, j = 1, 2, \dots\}, \quad k = 1, 2, 3, 4$$

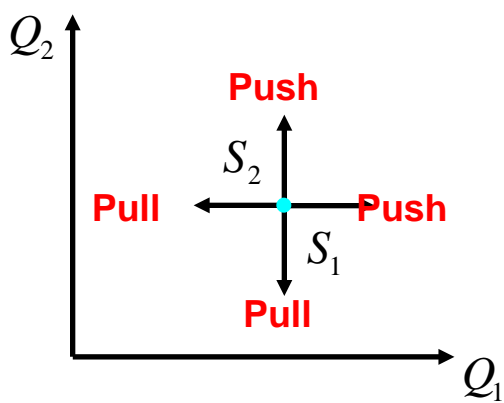
ξ_k i.i.d.

$$E[\xi_1] = 1, \quad E[\xi_3] = 1 \quad (\text{for simplicity})$$

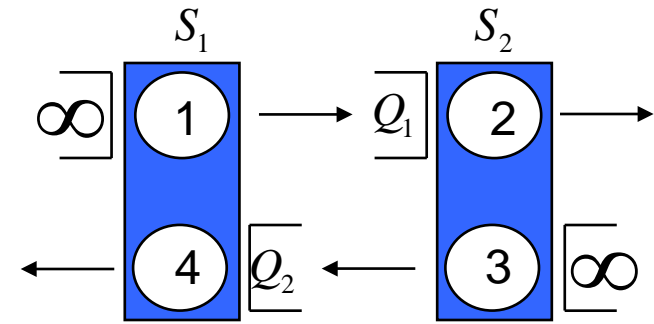
$$E[\xi_2] = \rho_2, \quad E[\xi_4] = \rho_2$$

“interesting” Configurations:

$$\rho_1, \rho_2 < 1 \quad \text{or} \quad \rho_1, \rho_2 > 1$$



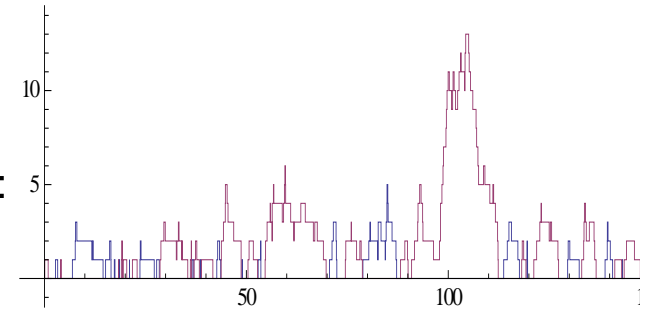
Policies



$$\rho_i < 1$$

Policy: Pull priority (LBFS)

Typical Behavior:

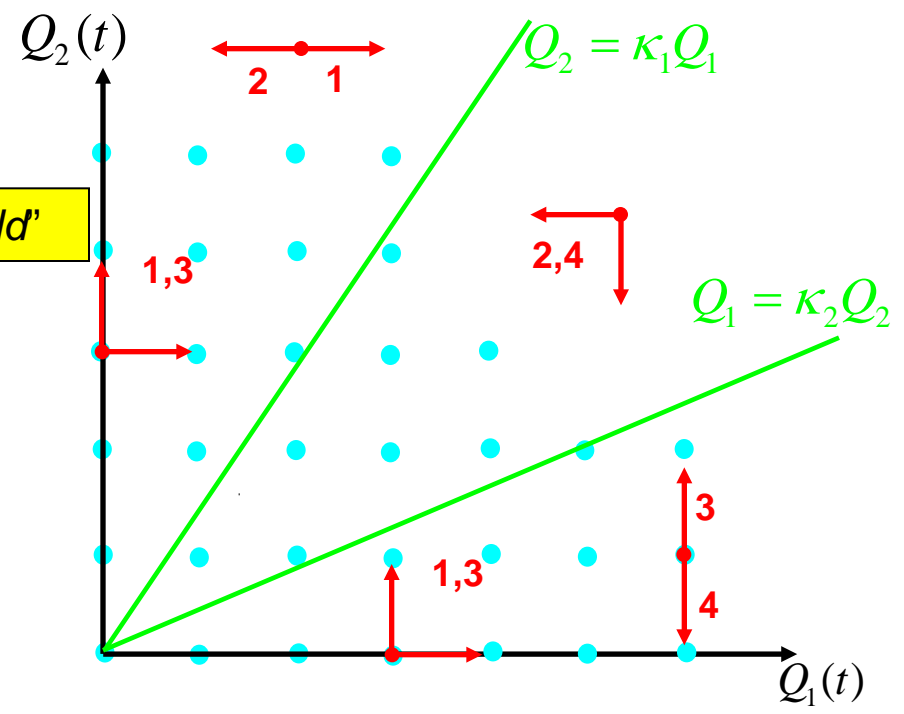
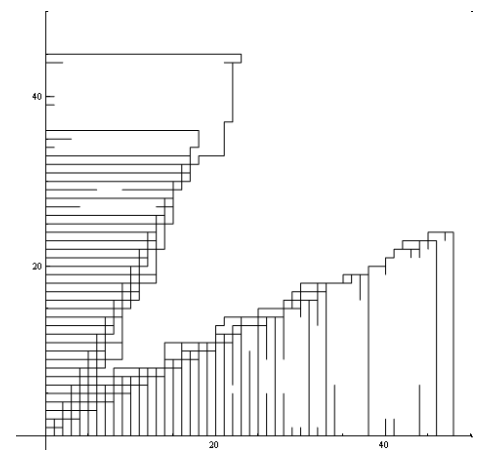


$$\rho_i > 1$$

Policy: Linear thresholds

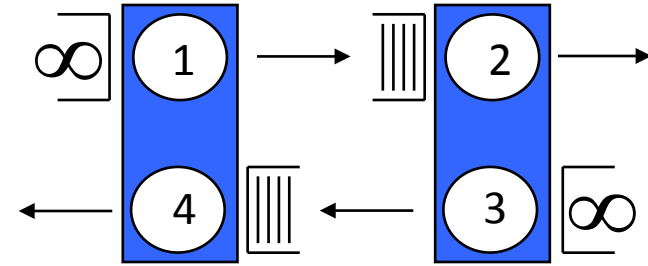
Server: "don't let opposite queue go below threshold"

Typical Behavior:



A Markov Process

$$X(t) = \begin{pmatrix} \text{Queue} & \text{Residual} \\ Q(t) & U(t) \end{pmatrix}$$



$X(t)$ is strong Markov with state space

$$\mathbb{S} = \mathbb{Z}_+^2 \times \mathbb{R}_+^2 \times \mathbb{R}_+^2$$

- For $x \in \mathbb{S}$, $B \in \mathcal{B}(\mathbb{S})$

$$P^t(x, B) = P_x(X(t) \in B) = P\{X(t) \in B | X(0) = x\}.$$

- π σ -finite on $(\mathbb{S}, \mathcal{B}(\mathbb{S}))$ is **invariant** if for all t

$$\pi(B) = \int_{\mathbb{S}} P^t(x, B) \pi(dx), \quad B \in \mathcal{B}(\mathbb{S}).$$

- $\tau_A = \inf\{t \geq 0 : X(t) \in A\}$. X is **Harris recurrent** if for measure ν , $A \in \mathcal{B}(\mathbb{S})$ with $\nu(A) > 0$ implies $P_x(\tau_A < \infty) = 1$ for all $x \in \mathbb{S}$.
- Harris Recurrent $\implies \exists$ invariant measure π .
- When π is a probability X is **positive Harris recurrent**.

Push-Pull Results (earlier work)

Stability

Theorem (N., Weiss): Pull-priority, $\rho_i < 1$, $X(t)$ is PHR

Theorem (N., Weiss): Linear thresholds, $\rho_i > 1$, $X(t)$ is PHR

Performance Analysis

Theorem (Kopzon, N., Weiss): Closed form for stationary distribution in specific cases and with memory-less assumptions

Diffusion (CLT) Limits of Outputs

Theorem (N., Weiss): Denote by $D(t)$ the number of outputs during $[0,t]$.

$$D^{(n)}(t) \Rightarrow BM(\Sigma, 0)$$

$$D^{(n)}(t) = \frac{D(nt) - vnt}{\sqrt{n}}$$

Note: This is weak convergence of the sequence $D^{(n)}(t)$ $n=1,2,\dots$

Explicit expression for covariance matrix Σ

Main Tool For Stability Results

Establish that an “associated” deterministic system $\bar{X}(t)$ is “stable”

The “framework” then implies that $X(t)$ is “stable”

Nice, since stability of $\bar{X}(t)$ is sometimes easier to establish

This “fluid framework” was pioneered and exploited in 90’s by Dai, Meyn, Stoylar, Bramson, Weiss, Chen

Stochastic and Fluid Equations

Fluid Network process

$$\bar{Y}(t) = (\bar{Q}_1(t), \bar{Q}_2(t), \bar{T}_1(t), \bar{T}_2(t), \bar{T}_3(t), \bar{T}_4(t))$$

Fluid Dynamics

~~$$\bar{S}_k(t) = \sup \left\{ n \cdot \sum_{j=1}^n \xi_{S_k}^j \leq t \right\} = \mu_k t$$~~

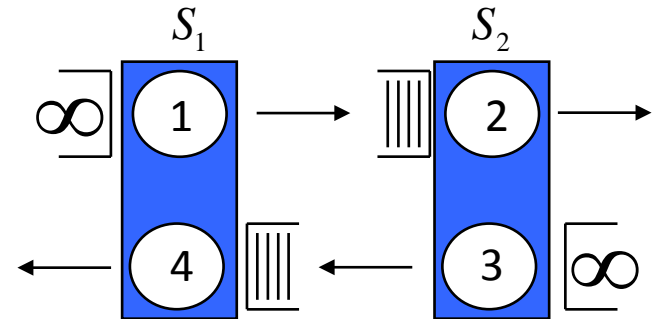
$$\bar{T}_k(0) = 0, \quad \bar{T}_k(t) \nearrow$$

$$\bar{T}_1(t) + \bar{T}_4(t) = t, \quad \bar{T}_2(t) + \bar{T}_3(t) = t$$

~~$$\bar{D}_k(t) = \bar{S}_k(\bar{T}_k(t)) = \mu_k \bar{T}_k(t)$$~~

$$\bar{Q}_k(0) = q_k, \quad \bar{Q}_k(t) \geq 0$$

$$\bar{Q}_k(t) = \bar{Q}_k(0) + \bar{D}_{k-1}(t) - \bar{D}_k(t)$$



Pull priority policy

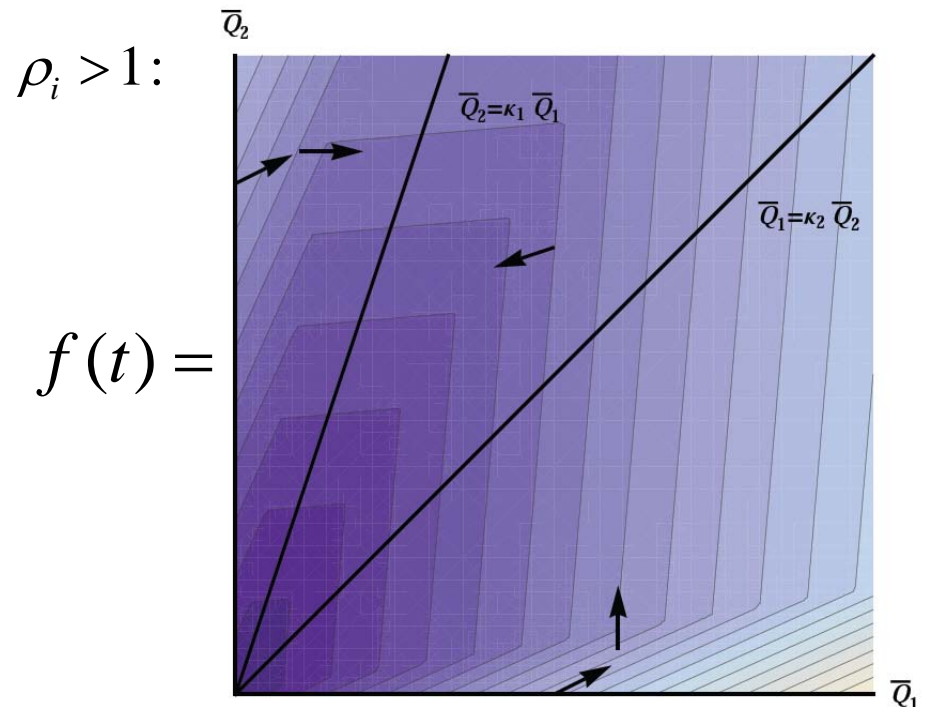
$$\int_0^t \bar{Q}_2(s) d\bar{T}_4(s) = 0 \quad \int_0^t \bar{Q}_1(s) d\bar{T}_1(s) = 0$$

E.g. Lyapounov Proofs for Fluid Stability

Need: for every solution of fluid model:

- When $f(t) = 0$, it stays at 0.
- When $f(t) > 0$, at regular points of t , $\dot{f}(t) \leq -\varepsilon$.

$$\rho_i < 1: f(t) = \bar{Q}_2(t) + \bar{Q}_4(t)$$



OUTLINE OF “NEW RESULTS”

For a ring of M servers with Pull-Priority (generalizing Push-Pull)

Theorem (Guo, Lefebvre, N., Weiss, Zhang): $X(t)$ is PHR (stable) if

- (i) $\rho_i < 1 \quad i = 1, \dots, M$
- (ii) When M is odd and $1 < \rho_i \quad i = 1, \dots, M$ and

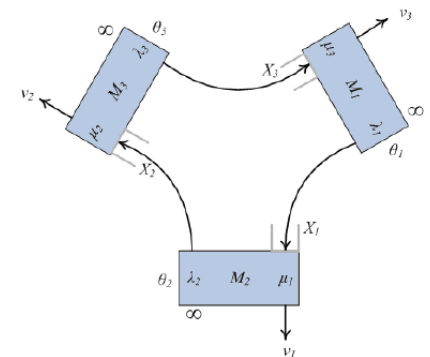
$$\Delta = \sum_{i=1}^M c_i \left(\frac{M-1}{2} (\rho_i - 1) - 1 \right) < 0$$

$$c_i = (\dots (\rho_{i-1} - 1) \rho_{i-2} + 1) \rho_{i-3} - 1) \rho_{i-4} + \dots + (-1)^{1+(M-1)/2} \rho_{i+1} + (-1)^{(M-1)/2}$$

Specifically when $\rho_i = \rho$ then $\Delta < 0$ iff $\rho < 1 + \frac{2}{M-1}$

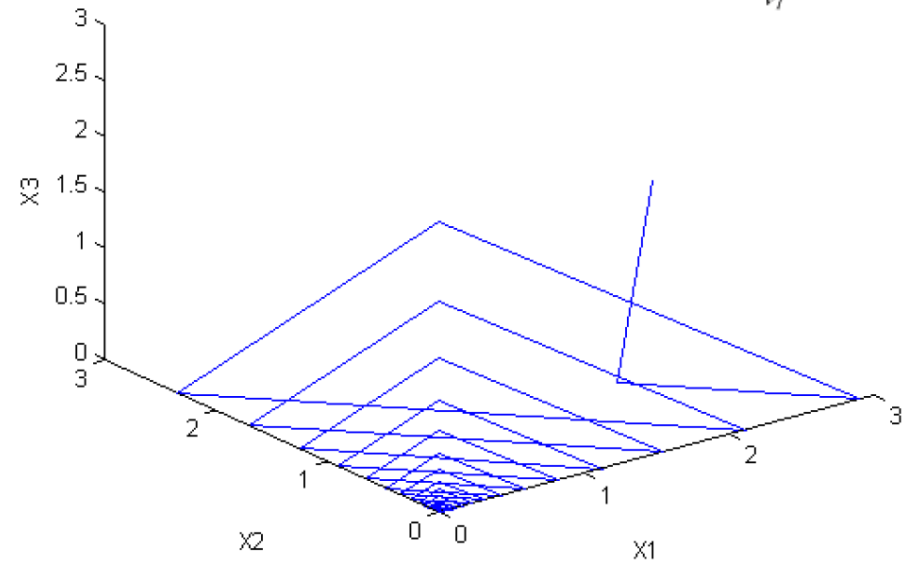
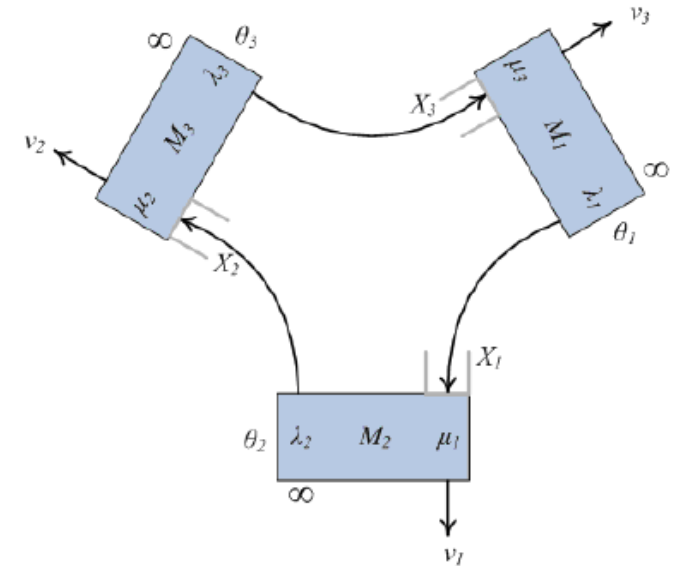
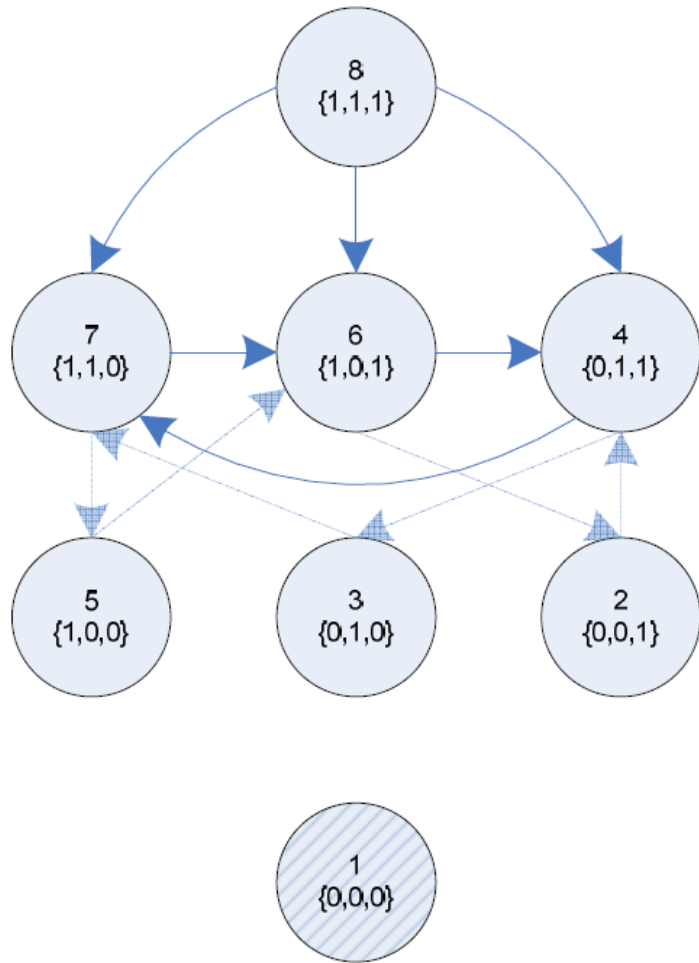
And when $M=3$,

$$\Delta = \prod_{i=1}^3 (\rho_i - 1) - 1$$

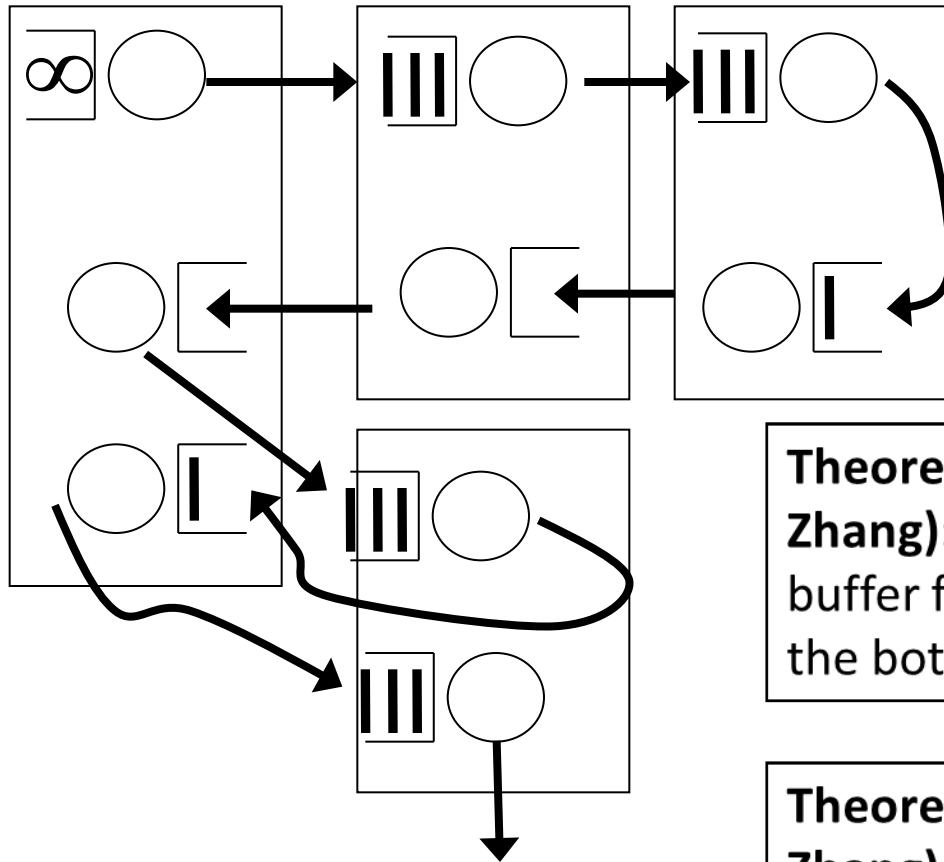


Stable Fluid Trajectory of $M=3$ Pull-Priority

$\rho_i > 1$



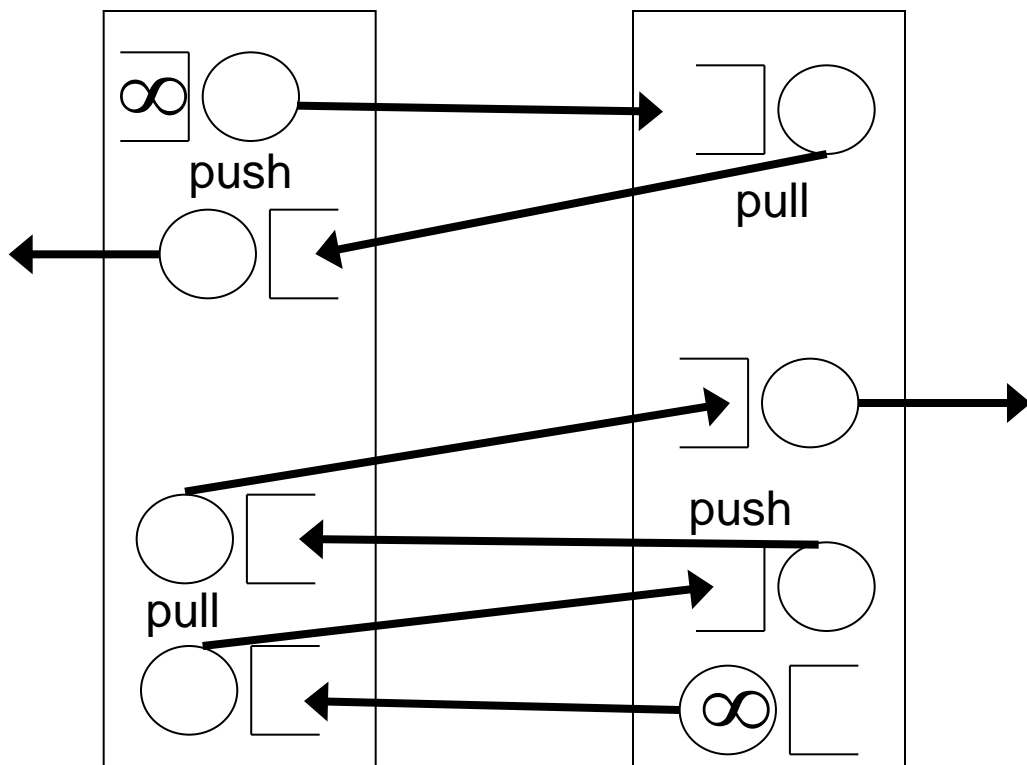
A re-entrant line with infinite supply



Theorem (Guo, Lefebvre, N., Weiss, Zhang): $X(t)$ is PHR (stable) under the last-buffer first served (LBFS) policy if first server is the bottleneck.

Theorem (Guo, Lefebvre, N., Weiss, Zhang): $X(t)$ is PHR (stable) under the **FBFS** policy under further conditions.

Two re-entrant lines in a push-pull



Theorem (Guo, Lefebvre, N., Weiss, Zhang):

$X(t)$ is PHR (stable) under a “pull-LBFS” priority policy if

push activities mean > pull activities mean
in each of the re-entrant lines

Diffusion Limit Results

Theorem (Guo, Lefebvre, N., Weiss, Zhang):

For general Multi-Class Queueing Networks with Infinite Supplies and deterministic routes (as in this talk), if the network is PHR,

$$D^{(n)}(t) \Rightarrow BM(\Sigma, 0)$$

With explicit “matrix expressions” for Σ .

Note that the diffusion limit does not depend on the specific policy (as long as it is a stabilizing policy).

$$D^{(n)}(t) = \frac{D(nt) - vnt}{\sqrt{n}}$$

Summary

- Specific cases of networks with infinite supplies (and full utilization) can be analysed with “some effort”
- General policies for stabilizing general networks remains an “open problem”

Recommended book on the subject: Bramson, 2009, Stability of queueing networks.

Association of Fluid Model and Stochastic System

fluid scalings

$$\bar{Y}^n(t, \omega) = \frac{Y^n(nt, \omega)}{n}$$

$\bar{Y}(t) = (\bar{Q}(t) \quad \bar{T}(t))$ is fluid limit

if exists $r \rightarrow \infty$ and $\omega: \bar{Y}^r(\cdot, \omega) \rightarrow \bar{Y}(\cdot)$, u.o.c.

\bar{Y} is **associated** with Y

if w.p.1 every fluid limit is a fluid model solution

Stability of Fluid Model

Definition: A fluid model is **stable**, if when ever, $q_1 + q_2 = 1$ there exists T, such that for all solutions,

$$Q_1(t) + Q_2(t) = 0 \quad \forall t \geq T$$

Definition: A fluid model is **weakly stable**, if when ever $q_1 + q_2 = 0$

$$Q_1(t) + Q_2(t) = 0 \quad \forall t \geq 0$$

Main Results of “Fluid Limit Method”

