

# Controller design for flow networks of switched servers with setup times

The Kumar-Seidman case as an illustrative example

Erjen Lefeber

Eindhoven University of Technology

Mathematical modeling of transport and production logistics  
January 11, 2008, Bremen

# Motivation



# Motivation



# Motivation



# Motivation



# Problem

## Problem

How to control these networks?

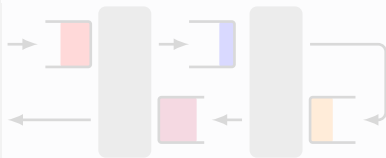
Decisions: **When** to switch, and **to which** job-type

Goals: Maximal throughput, minimal flow time

## Current approach

**Start from policy**, analyze resulting dynamics

## Kumar, Seidman (1990)



Clearing



# Problem

## Problem

How to control these networks?

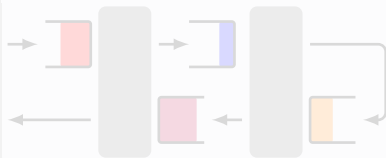
Decisions: **When** to switch, and **to which** job-type

Goals: Maximal throughput, minimal flow time

## Current approach

**Start from policy**, analyze resulting dynamics

Kumar, Seidman (1990)



Clearing



# Problem

## Problem

How to control these networks?

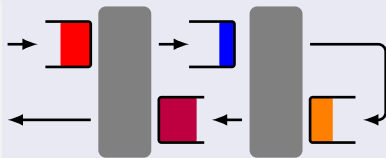
Decisions: **When** to switch, and **to which** job-type

Goals: Maximal throughput, minimal flow time

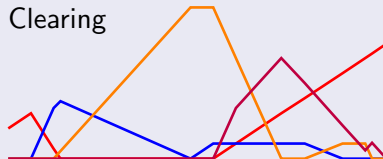
## Current approach

**Start from policy**, analyze resulting dynamics

## Kumar, Seidman (1990)



Clearing



# Problem

## Current status (after two decades)

Several policies exist that guarantee **stability** of the network

## Remark

Stability is **only a prerequisite** for a good policy

## Open issues

- Do existing policies yield satisfactory network performance?
- How to obtain pre-specified network behavior?

## Main subject of study (modest)

Fixed, deterministic flow networks (not evolving, constant inflow)

# Problem

## Current status (after two decades)

Several policies exist that guarantee **stability** of the network

## Remark

Stability is **only a prerequisite** for a good policy

## Open issues

- Do existing policies yield satisfactory network performance?
- How to obtain pre-specified network behavior?

## Main subject of study (modest)

Fixed, deterministic flow networks (not evolving, constant inflow)

# Problem

## Current status (after two decades)

Several policies exist that guarantee **stability** of the network

## Remark

Stability is **only a prerequisite** for a good policy

## Open issues

- Do existing policies yield satisfactory network performance?
- How to obtain pre-specified network behavior?

## Main subject of study (modest)

Fixed, deterministic flow networks (not evolving, constant inflow)

# Problem

## Current status (after two decades)

Several policies exist that guarantee **stability** of the network

## Remark

Stability is **only a prerequisite** for a good policy

## Open issues

- Do existing policies yield satisfactory network performance?
- How to obtain pre-specified network behavior?

## Main subject of study (modest)

Fixed, deterministic flow networks (not evolving, constant inflow)

# Main idea

## Important observation

“The main interest is in the **resulting behavior**. So why not use that as a **starting point**?”

## Approach

**Start from desired behavior** and *design* policy, instead of **start from policy** and analyze resulting dynamics

## Consequence

Separation of concern: **desired behavior** and **controller** can be designed **separately**.

# Main idea

## Important observation

“The main interest is in the **resulting behavior**. So why not use that as a **starting point**?”

## Approach

**Start from desired behavior** and *design* policy, instead of **start from policy** and analyze resulting dynamics

## Consequence

Separation of concern: **desired behavior** and **controller** can be designed **separately**.

# Main idea

## Important observation

“The main interest is in the **resulting behavior**. So why not use that as a **starting point**?”

## Approach

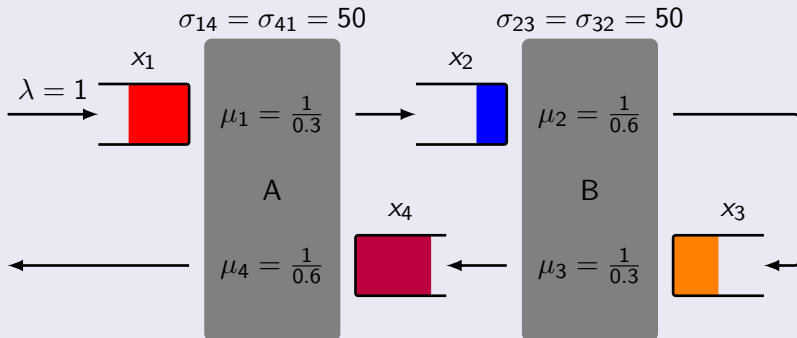
**Start from desired behavior** and *design* policy, instead of **start from policy** and analyze resulting dynamics

## Consequence

Separation of concern: **desired behavior** and **controller** can be designed **separately**.

# Kumar-Seidman case

Transactions on Automatic Control, Vol 35, No 3, March 1990



## Observation

Sufficient capacity (consider period of at least 1000).

# Model (hybrid)

## State

$$x_0^A, x_0^B$$

remaining setup time machine A,B

$$x_i$$

buffer contents ( $i \in \{1, 2, 3, 4\}$ )

$$m = (m^A, m^B)$$

mode  $\in \{(1, 2), (1, 3), (4, 2), (4, 3)\}$

## Input

$$u_0^A, u_0^B$$

activity  $\in \{\textcircled{1}, \textcircled{2}, \textcircled{3}, \textcircled{4}, \textcircled{1}, \textcircled{2}, \textcircled{3}, \textcircled{4}\}$

$$u_i$$

service rate step  $i \in \{1, 2, 3, 4\}$

# Model (hybrid)

## State

$$x_0^A, x_0^B$$

remaining setup time machine A,B

$$x_i$$

buffer contents ( $i \in \{1, 2, 3, 4\}$ )

$$m = (m^A, m^B)$$

mode  $\in \{(1, 2), (1, 3), (4, 2), (4, 3)\}$

## Input

$$u_0^A, u_0^B$$

activity  $\in \{\textcircled{1}, \textcircled{2}, \textcircled{3}, \textcircled{4}, \textcircled{1}, \textcircled{2}, \textcircled{3}, \textcircled{4}\}$

$$u_i$$

service rate step  $i \in \{1, 2, 3, 4\}$

# Model (hybrid)

## Continuous dynamics

$$\dot{x}_0^A(t) = \begin{cases} -1 & \text{if } u_0^A \in \{\textcircled{1}, \textcircled{4}\} \\ 0 & \text{if } u_0^A \in \{\textcircled{1}, \textcircled{4}\} \end{cases} \quad \dot{x}_0^B(t) = \begin{cases} -1 & \text{if } u_0^B \in \{\textcircled{2}, \textcircled{3}\} \\ 0 & \text{if } u_0^B \in \{\textcircled{2}, \textcircled{3}\} \end{cases}$$

$$\dot{x}_1(t) = \lambda - u_1(t)$$

$$\dot{x}_2(t) = u_1(t) - u_2(t)$$

$$\dot{x}_4(t) = u_3(t) - u_4(t)$$

$$\dot{x}_3(t) = u_2(t) - u_3(t).$$

## Discrete event dynamics

$$x_0^A := \sigma_{14}; \quad m^A := 4 \quad \text{if } u_0^A = \textcircled{4} \text{ and } m^A = 1$$

$$x_0^A := \sigma_{41}; \quad m^A := 1 \quad \text{if } u_0^A = \textcircled{1} \text{ and } m^A = 4$$

$$x_0^B := \sigma_{23}; \quad m^B := 3 \quad \text{if } u_0^B = \textcircled{3} \text{ and } m^B = 2$$

$$x_0^B := \sigma_{32}; \quad m^B := 2 \quad \text{if } u_0^B = \textcircled{2} \text{ and } m^B = 3$$

# Model (hybrid)

## Continuous dynamics

$$\dot{x}_0^A(t) = \begin{cases} -1 & \text{if } u_0^A \in \{\textcircled{1}, \textcircled{4}\} \\ 0 & \text{if } u_0^A \in \{\textcircled{1}, \textcircled{4}\} \end{cases} \quad \dot{x}_0^B(t) = \begin{cases} -1 & \text{if } u_0^B \in \{\textcircled{2}, \textcircled{3}\} \\ 0 & \text{if } u_0^B \in \{\textcircled{2}, \textcircled{3}\} \end{cases}$$

$$\dot{x}_1(t) = \lambda - u_1(t)$$

$$\dot{x}_2(t) = u_1(t) - u_2(t)$$

$$\dot{x}_4(t) = u_3(t) - u_4(t)$$

$$\dot{x}_3(t) = u_2(t) - u_3(t).$$

## Discrete event dynamics

$$x_0^A := \sigma_{14}; \quad m^A := 4 \quad \text{if } u_0^A = \textcircled{4} \text{ and } m^A = 1$$

$$x_0^A := \sigma_{41}; \quad m^A := 1 \quad \text{if } u_0^A = \textcircled{1} \text{ and } m^A = 4$$

$$x_0^B := \sigma_{23}; \quad m^B := 3 \quad \text{if } u_0^B = \textcircled{3} \text{ and } m^B = 2$$

$$x_0^B := \sigma_{32}; \quad m^B := 2 \quad \text{if } u_0^B = \textcircled{2} \text{ and } m^B = 3$$

# Model (hybrid)

## Input constraints

$$u_0^A \in \{\textcircled{1}, \textcircled{4}\}, u_1 = 0, u_4 = 0$$

$$\text{if } x_0^A > 0$$

$$u_0^A \in \{\textcircled{1}, \textcircled{4}\}, u_1 \leq \mu_1, u_4 = 0$$

$$\text{if } x_0^A = 0, x_1 > 0, m^A = 1$$

$$u_0^A \in \{\textcircled{1}, \textcircled{4}\}, u_1 \leq \lambda, u_4 = 0$$

$$\text{if } x_0^A = 0, x_1 = 0, m^A = 1$$

$$u_0^A \in \{\textcircled{1}, \textcircled{4}\}, u_1 = 0, u_4 \leq \mu_4$$

$$\text{if } x_0^A = 0, x_4 > 0, m^A = 4$$

$$u_0^A \in \{\textcircled{1}, \textcircled{4}\}, u_1 = 0, u_4 \leq \min(u_3, \mu_4) \quad \text{if } x_0^A = 0, x_4 = 0, m^A = 4$$

$$u_0^B \in \{\textcircled{2}, \textcircled{3}\}, u_2 = 0, u_3 = 0$$

$$\text{if } x_0^B > 0$$

$$u_0^B \in \{\textcircled{2}, \textcircled{3}\}, u_2 \leq \mu_2, u_3 = 0$$

$$\text{if } x_0^B = 0, x_2 > 0, m^B = 2$$

$$u_0^B \in \{\textcircled{2}, \textcircled{3}\}, u_2 \leq \min(u_1, \mu_2), u_3 = 0 \quad \text{if } x_0^B = 0, x_2 = 0, m^B = 2$$

$$u_0^B \in \{\textcircled{2}, \textcircled{3}\}, u_2 = 0, u_3 \leq \mu_3$$

$$\text{if } x_0^B = 0, x_3 > 0, m^B = 3$$

$$u_0^B \in \{\textcircled{2}, \textcircled{3}\}, u_2 = 0, u_3 \leq u_2$$

$$\text{if } x_0^B = 0, x_3 = 0, m^B = 3$$

# Model (hybrid)

## Input constraints

$$u_0^A \in \{\mathbf{1}, \mathbf{4}\}, u_1 = 0, u_4 = 0$$

$$\text{if } x_0^A > 0$$

$$u_0^A \in \{\mathbf{1}, \mathbf{4}\}, u_1 \leq \mu_1, u_4 = 0$$

$$\text{if } x_0^A = 0, x_1 > 0, m^A = 1$$

$$u_0^A \in \{\mathbf{1}, \mathbf{4}\}, u_1 \leq \lambda, u_4 = 0$$

$$\text{if } x_0^A = 0, x_1 = 0, m^A = 1$$

$$u_0^A \in \{\mathbf{1}, \mathbf{4}\}, u_1 = 0, u_4 \leq \mu_4$$

$$\text{if } x_0^A = 0, x_4 > 0, m^A = 4$$

$$u_0^A \in \{\mathbf{1}, \mathbf{4}\}, u_1 = 0, u_4 \leq \min(u_3, \mu_4) \quad \text{if } x_0^A = 0, x_4 = 0, m^A = 4$$

$$u_0^B \in \{\mathbf{2}, \mathbf{3}\}, u_2 = 0, u_3 = 0$$

$$\text{if } x_0^B > 0$$

$$u_0^B \in \{\mathbf{2}, \mathbf{3}\}, u_2 \leq \mu_2, u_3 = 0$$

$$\text{if } x_0^B = 0, x_2 > 0, m^B = 2$$

$$u_0^B \in \{\mathbf{2}, \mathbf{3}\}, u_2 \leq \min(u_1, \mu_2), u_3 = 0 \quad \text{if } x_0^B = 0, x_2 = 0, m^B = 2$$

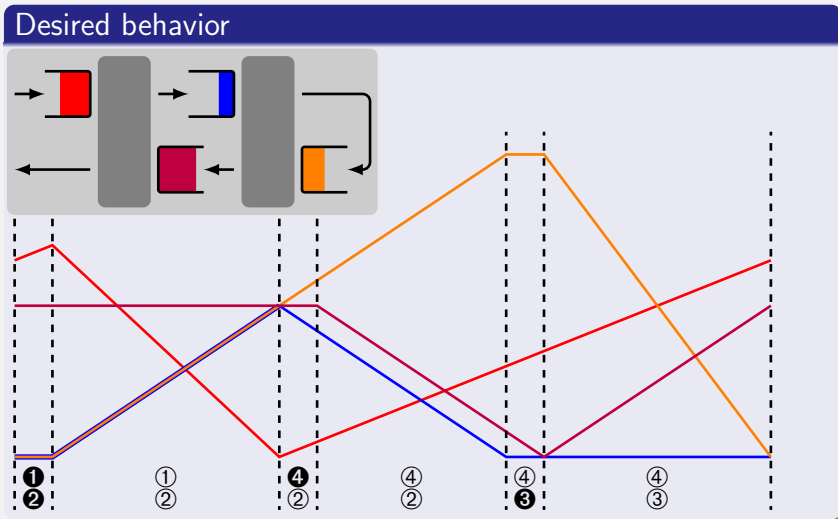
$$u_0^B \in \{\mathbf{2}, \mathbf{3}\}, u_2 = 0, u_3 \leq \mu_3$$

$$\text{if } x_0^B = 0, x_3 > 0, m^B = 3$$

$$u_0^B \in \{\mathbf{2}, \mathbf{3}\}, u_2 = 0, u_3 \leq u_2$$

$$\text{if } x_0^B = 0, x_3 = 0, m^B = 3$$

# Desired behavior



# Controller design

## Main idea

Lyapunov: If energy is decreasing all the time  $\Rightarrow$  system should settle down at constant energy level

## Challenge

Determine energy-function (based on desired periodic orbit)

## Observation I

Desired periodic orbit provides a fixed sequence of modes, with a given duration.

# Controller design

## Main idea

Lyapunov: If energy is decreasing all the time  $\Rightarrow$  system should settle down at constant energy level

## Challenge

Determine energy-function (based on desired periodic orbit)

## Observation I

Desired periodic orbit provides a fixed sequence of modes, with a given duration.

# Controller design

## Main idea

Lyapunov: If energy is decreasing all the time  $\Rightarrow$  system should settle down at constant energy level

## Challenge

Determine energy-function (based on desired periodic orbit)

## Observation I

Desired periodic orbit provides a fixed sequence of modes, with a given duration.

# Controller design

## Observation II

Blindly applying fixed sequence of modes for corresponding duration makes system converge to translated desired periodic orbit, i.e. with additional lots in buffers (A.V. Savkin, 1998)

## Observation III

Remaining duration of current mode can still be chosen

## Final ingredient

Amount of work:  $1.8x_1 + 1.5x_2 + 0.9x_3 + 0.6x_4$

# Controller design

## Observation II

Blindly applying fixed sequence of modes for corresponding duration makes system converge to translated desired periodic orbit, i.e. with additional lots in buffers (A.V. Savkin, 1998)

## Observation III

Remaining duration of current mode can still be chosen

## Final ingredient

Amount of work:  $1.8x_1 + 1.5x_2 + 0.9x_3 + 0.6x_4$

# Controller design

## Observation II

Blindly applying fixed sequence of modes for corresponding duration makes system converge to translated desired periodic orbit, i.e. with additional lots in buffers (A.V. Savkin, 1998)

## Observation III

Remaining duration of current mode can still be chosen

## Final ingredient

Amount of work:  $1.8x_1 + 1.5x_2 + 0.9x_3 + 0.6x_4$

# Controller design

## Notice

- Current state
- Remaining duration of current mode



Additional amount of work

## Lyapunov function candidate

For given state: the lowest possible additional amount of work

## Controller design

Over all possible inputs: pick one which makes Lyapunov function candidate decrease the most.

# Controller design

## Notice

- Current state
- Remaining duration of current mode



Additional amount of work

## Lyapunov function candidate

For given state: the lowest possible additional amount of work

## Controller design

Over all possible inputs: pick one which makes Lyapunov function candidate decrease the most.

# Controller design

## Notice

- Current state
- Remaining duration of current mode



Additional amount of work

## Lyapunov function candidate

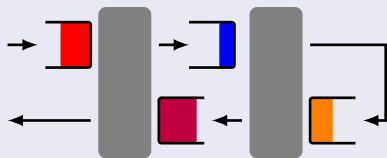
For given state: the lowest possible additional amount of work

## Controller design

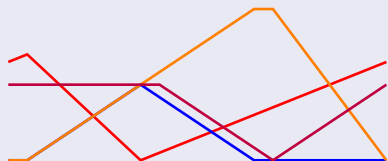
Over all possible inputs: pick one which makes Lyapunov function candidate decrease the most.

# Resulting controller

## Network



## Desired behavior



## Resulting controller

Mode (1,2): to (4,2) when both  $x_1 = 0$  and  $x_2 + x_3 \geq 1000$

Mode (4,2): to (4,3) when both  $x_2 = 0$  and  $x_4 \leq 83\frac{1}{3}$

Mode (4,3): to (1,2) when  $x_3 = 0$

# Resulting controller

## Proof

Buffer amounts as A starts serving 1 for  $k^{\text{th}}$  time:  $(x_1^k, x_2^k, x_3^k, x_4^k)$

Then for  $k > 2$ :

$$x_1^{k+1} = 100 + \frac{3}{7}x_1^k + \max(\frac{3}{7}x_1^k, \frac{3}{5}x_4^k) \quad x_2^{k+1} = 0$$

$$x_4^{k+1} = \max(500, \frac{5}{7}x_1^k) \quad x_3^{k+1} = 0$$

Since  $x_1^{k+1} \leq 100 + \frac{3}{7}x_1^k + \frac{3}{7} \max(x_1^k, x_1^{k-1})$ :

Contraction with fixed point (700, 0, 0, 500).

## Remark

Centralized controller, i.e. non-distributed

# Resulting controller

## Proof

Buffer amounts as A starts serving 1 for  $k^{\text{th}}$  time:  $(x_1^k, x_2^k, x_3^k, x_4^k)$

Then for  $k > 2$ :

$$x_1^{k+1} = 100 + \frac{3}{7}x_1^k + \max(\frac{3}{7}x_1^k, \frac{3}{5}x_4^k) \quad x_2^{k+1} = 0$$

$$x_4^{k+1} = \max(500, \frac{5}{7}x_1^k) \quad x_3^{k+1} = 0$$

Since  $x_1^{k+1} \leq 100 + \frac{3}{7}x_1^k + \frac{3}{7} \max(x_1^k, x_1^{k-1})$ :

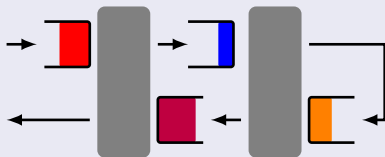
Contraction with fixed point (700, 0, 0, 500).

## Remark

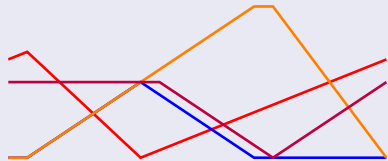
Centralized controller, i.e. non-distributed

# Distributed controller

## Network



## Desired behavior



## Distributed controller

**Serving 1:** Serve at least 1000 jobs until  $x_1 = 0$ , then switch.  
Let  $\bar{x}_1$  be nr of jobs served.

**Serving 4:** Let  $\bar{x}_4$  be nr of jobs in Buffer 4 after setup. Serve  $\bar{x}_4 + \frac{1}{2}\bar{x}_1$  jobs, then switch.

**Serving 2:** Serve at least 1000 jobs until  $x_2 = 0$ , then switch.

**Serving 3:** Empty buffer, then switch.

# Conclusions

## Non-distributed/centralized control

- Given a feasible periodic orbit, a controller can be derived.
- Approach can deal with
  - General networks
  - Finite buffers
  - Transportation delays

## Distributed control

- For case was shown that distributed implementation exists
- Relates to **observability**

# Conclusions

## Non-distributed/centralized control

- Given a feasible periodic orbit, a controller can be derived.
- Approach can deal with
  - General networks
  - Finite buffers
  - Transportation delays

## Distributed control

- For case was shown that distributed implementation exists
- Relates to **observability**

# Concluding remarks

## Ideas from control theory can be useful

- 1 Determine **optimal behavior** (trajectory generation)
- 2 Derive **centralized controller** (state feedback control)
- 3 Derive **decentralized controllers** (dyn. output feedback)

## Many questions remaining

- How to find good (or even optimal) network behavior?
- How to design decentralized controllers (observability)?
- Robustness against parameter changes?
- What if network is modified?
- What if arrival rate not constant?
- What if routing is not fixed?

# Concluding remarks

## Ideas from control theory can be useful

- 1 Determine **optimal behavior** (trajectory generation)
- 2 Derive **centralized controller** (state feedback control)
- 3 Derive **decentralized controllers** (dyn. output feedback)

## Many questions remaining

- How to find good (or even optimal) network behavior?
- How to design decentralized controllers (observability)?
- Robustness against parameter changes?
- What if network is modified?
- What if arrival rate not constant?
- What if routing is not fixed?