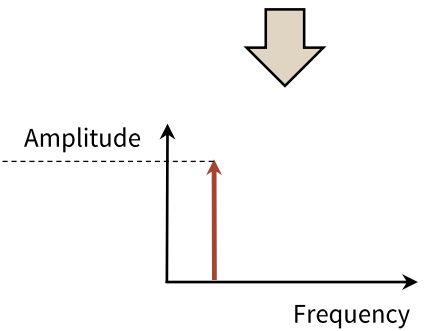
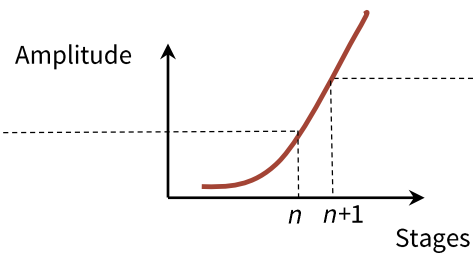
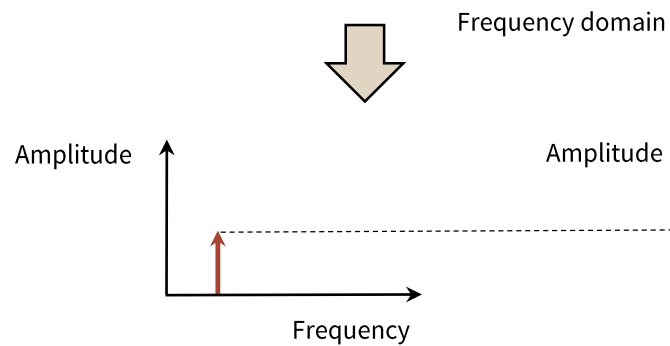
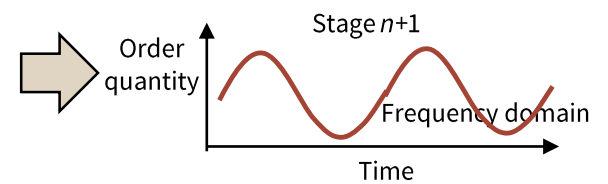
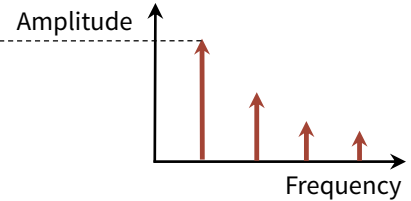
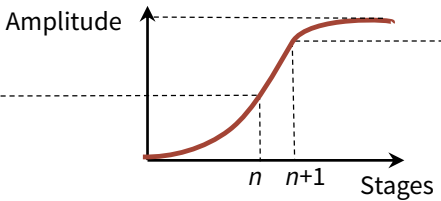
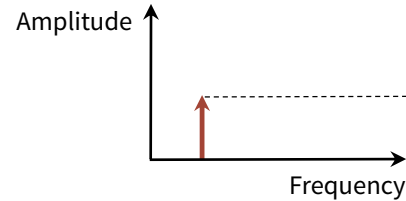
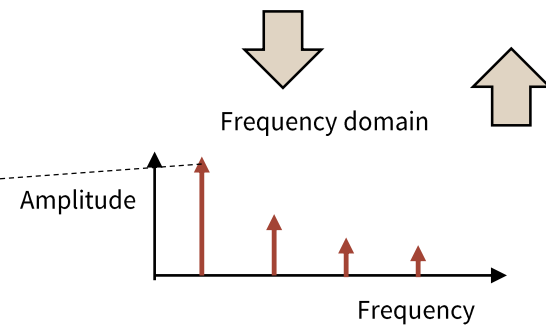
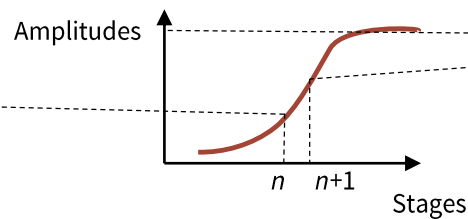
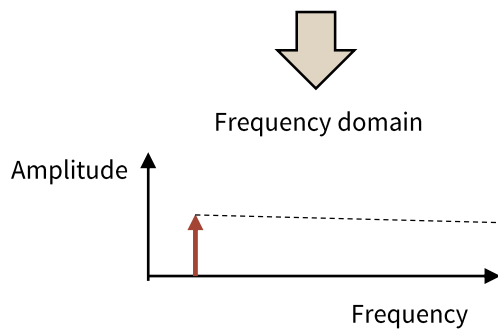
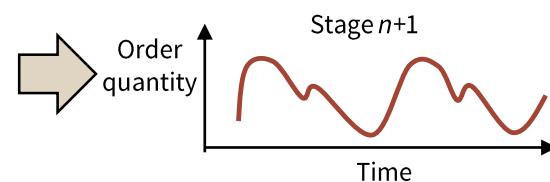
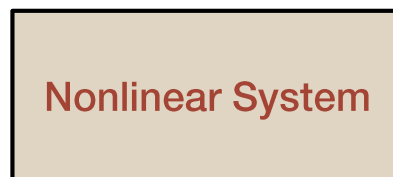
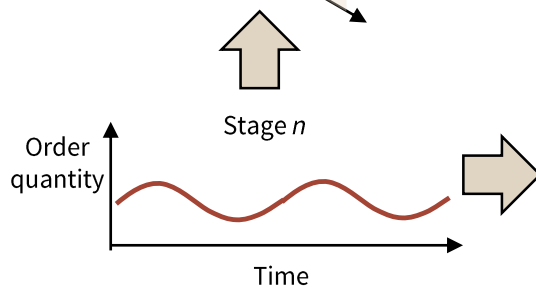
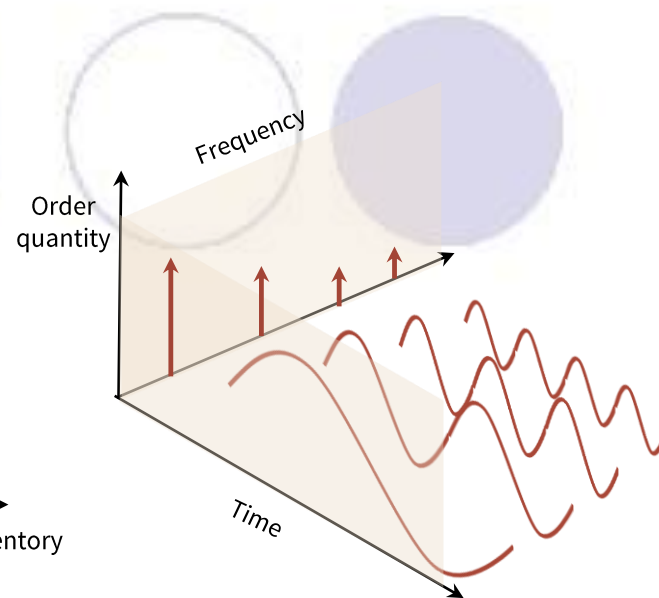
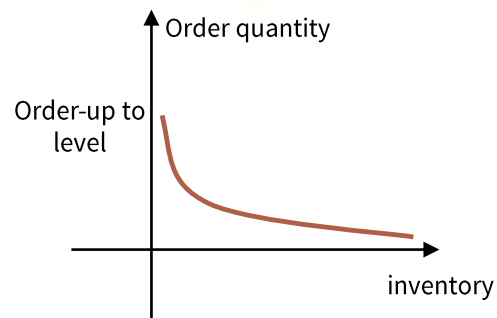
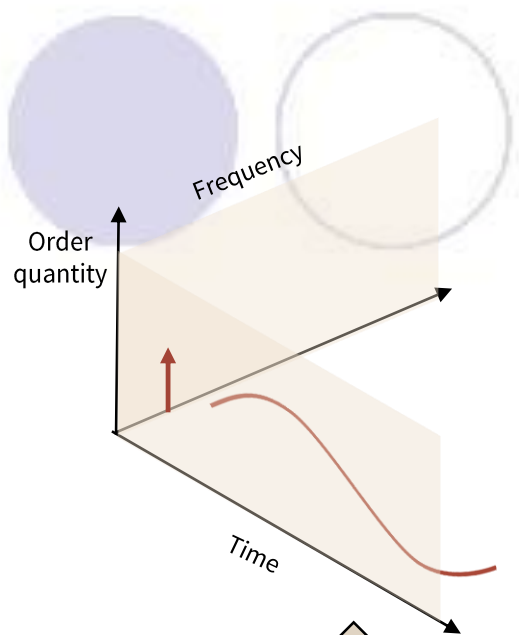
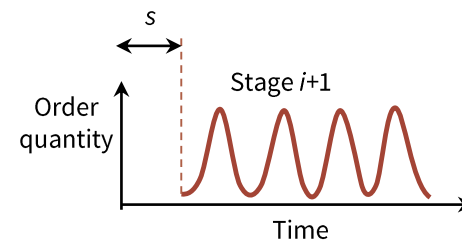
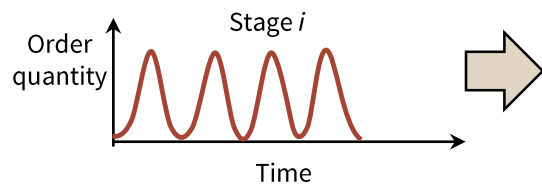
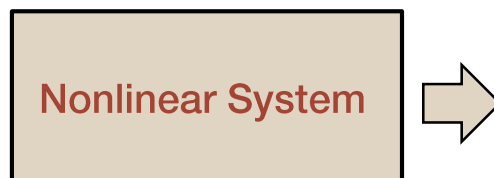
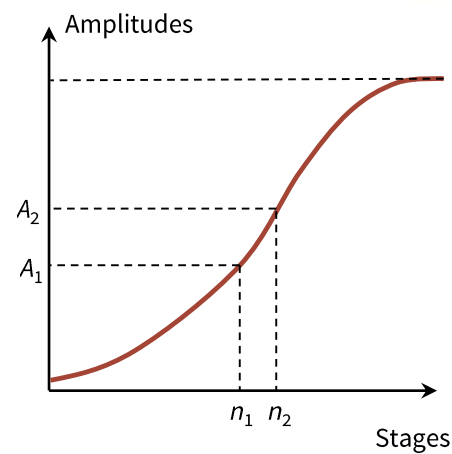
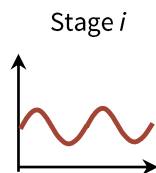
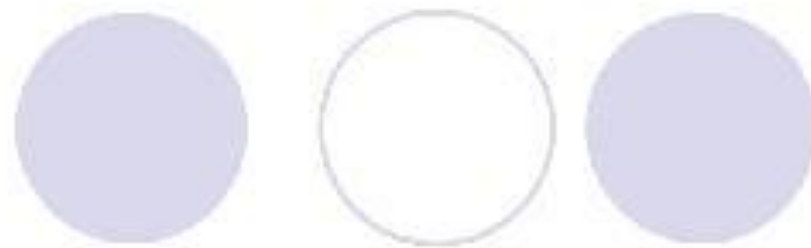
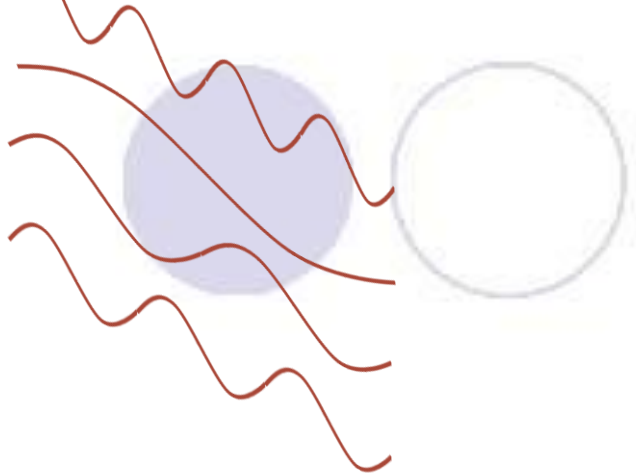


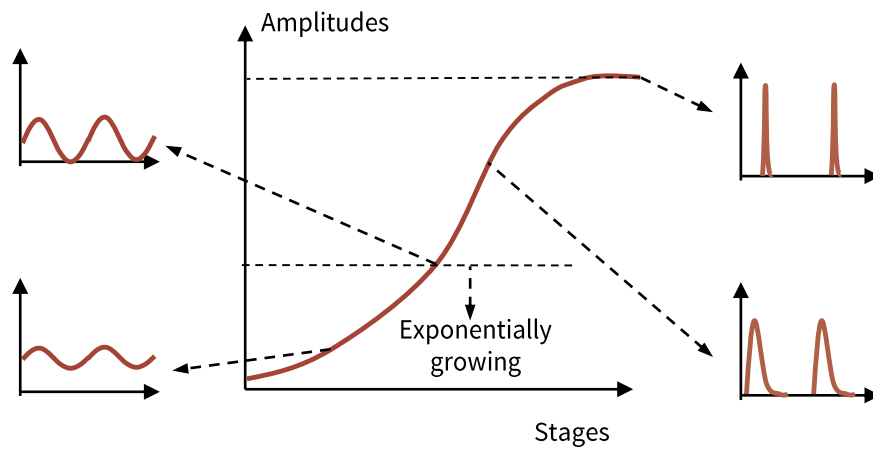
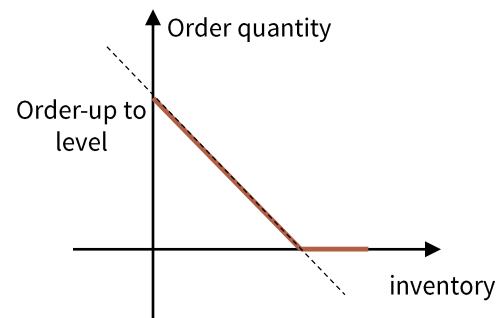
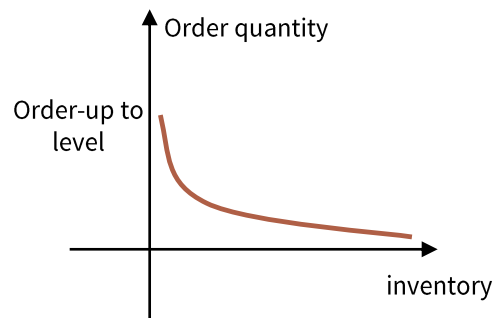
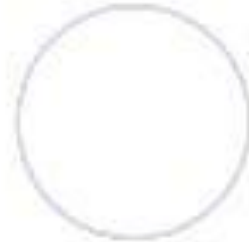
Linear System

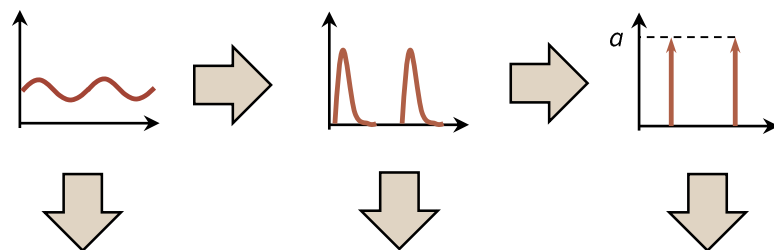
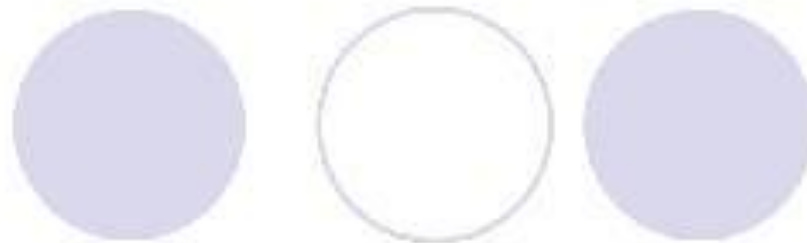
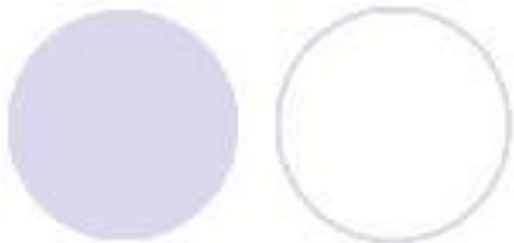


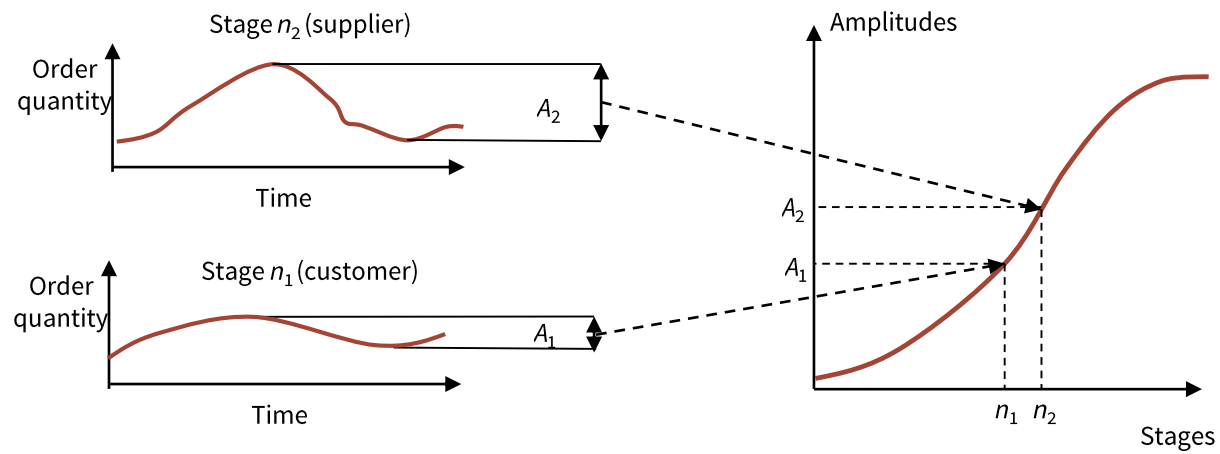
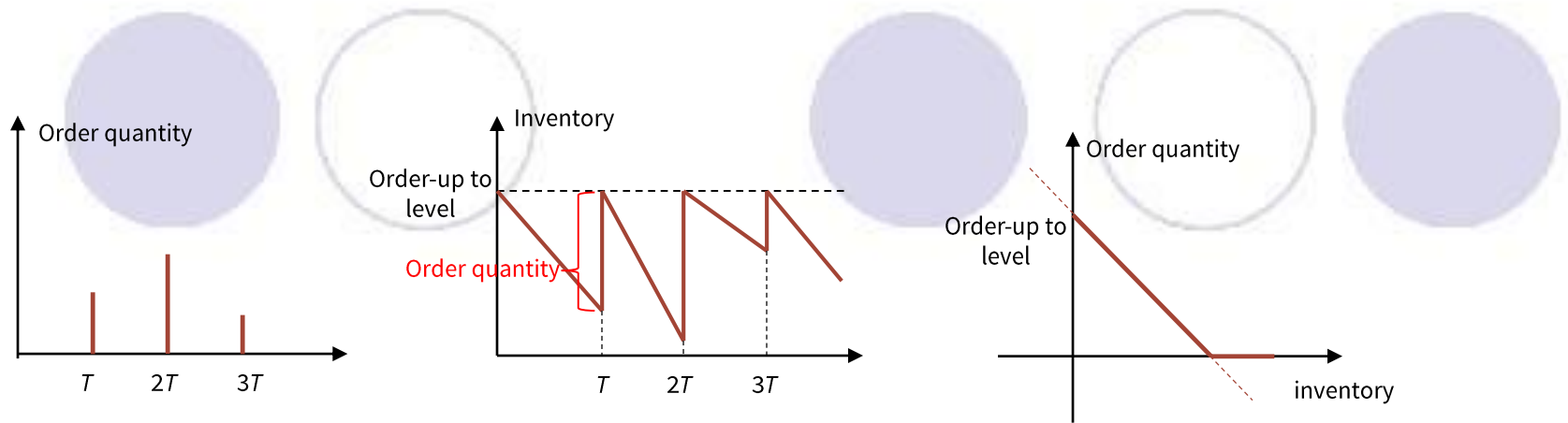














Decentralized Systems

- System-level instability
 - disturbances amplified across system
 - compromise system performance
- Can we achieve centralized performance with decentralized control?
 - stabilize
 - optimize



Outline

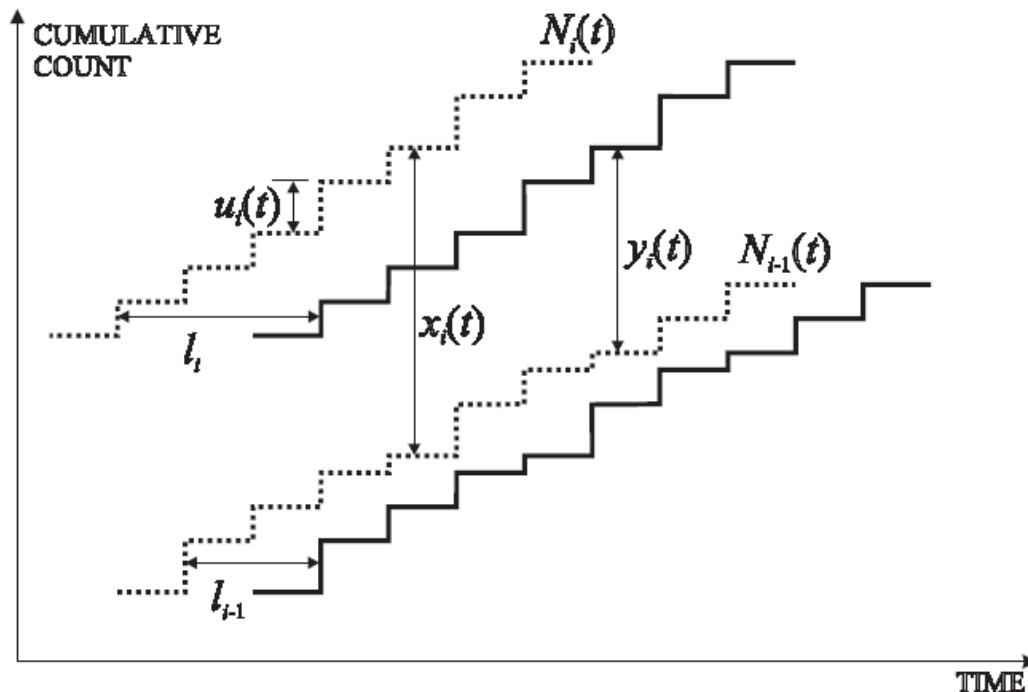
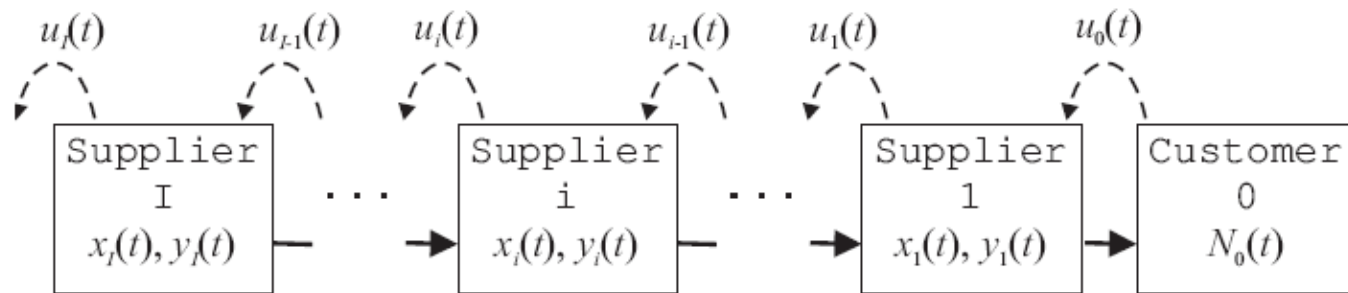
- Basics
 - Background
 - The bullwhip effect
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Goal

- Establish a system-control framework
 - Generalize past work on deterministic and homogeneous system (Daganzo, 2001, 2003a, b; Dejonckheere *et al.*, 2003a, b)
 - Further understand the bullwhip effect
- Develop analytical methods to examine the existence of the bullwhip effect

Supply Chain Representation



Definitions

$N_i(t)$ = cumulative orders placed by supplier i by period t

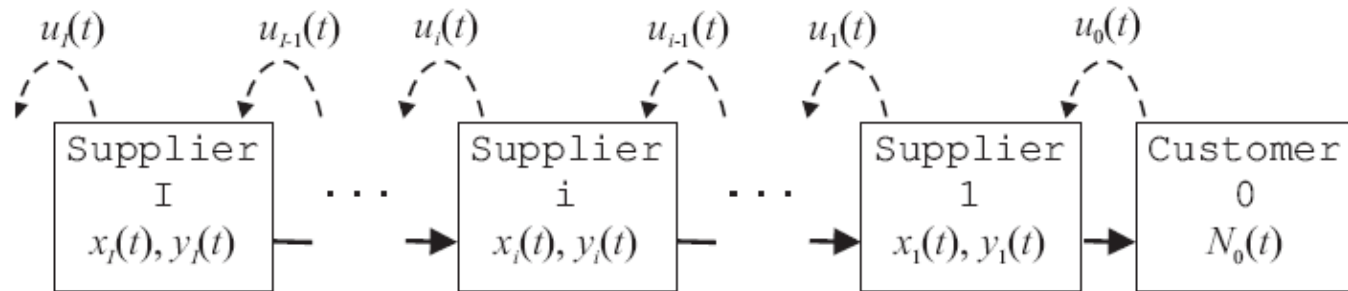
$u_i(t)$ = order placed by supplier i in period t

$y_i(t)$ = in-stock inventory

$x_i(t)$ = inventory position

l_i = lead time

Assumptions

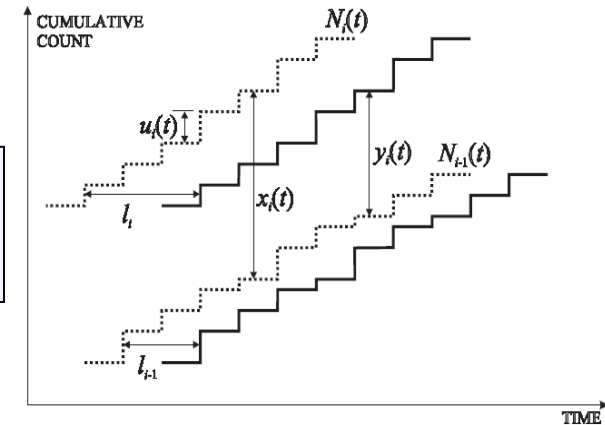


- Decentralized decision-making
- Rational ordering decisions
- Linear ordering policy
- Deterministic system operations

System Dynamics

Inventory
dynamics:

$$\begin{aligned} x_i(t+1) &= x_i(t) + u_i(t) - u_{i-1}(t), \forall i = 0, 1, \dots, \\ y_i(t+1) &= y_i(t) + u_i(t - l_i) - u_{i-1}(t), \forall i = 1, 2, \dots \end{aligned}$$



Ordering
policy:

$$u_i(t) = \gamma_i + A_i(P)x_i(t) + B_i(P)y_i(t) + C_i(P)u_{i-1}(t-1), i = 1, 2, \dots$$

P : shift operator; i.e. $P^k x(t) = x(t-k)$, \forall integer $k \geq 0$

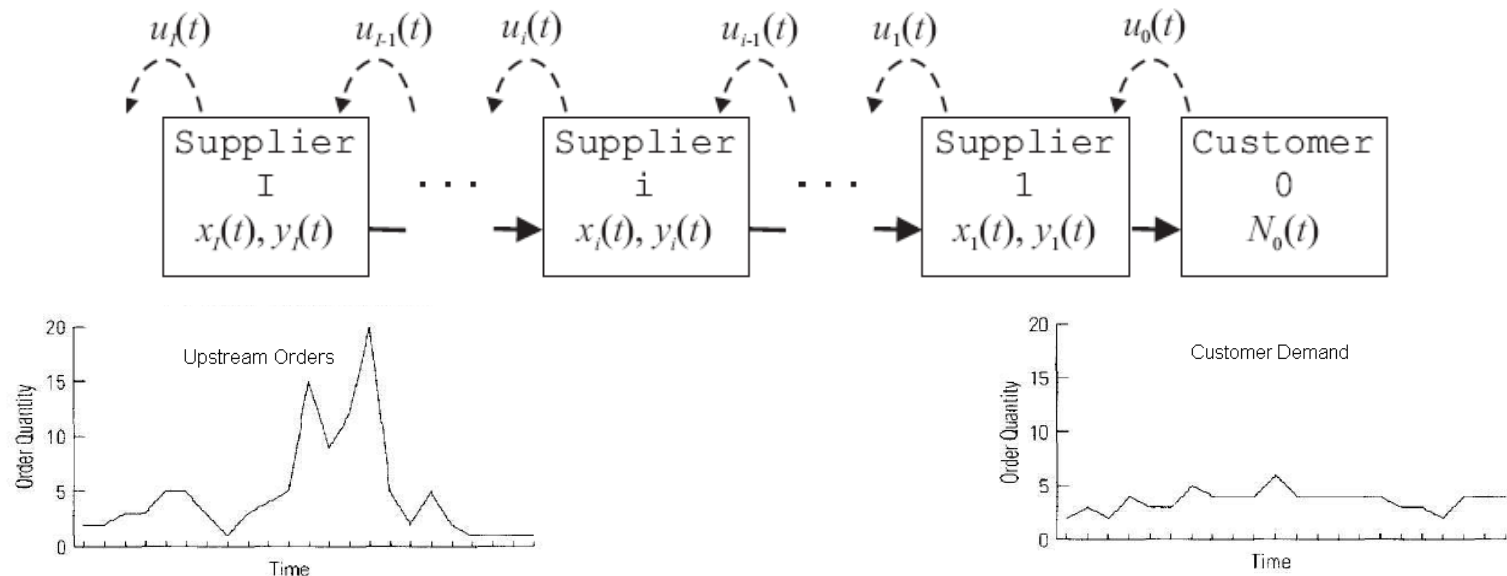
$A_i(\cdot)$, $B_i(\cdot)$, $C_i(\cdot)$: polynomials with real coefficients;

γ_i : constant

Equilibrium
state:

$$u^\infty = \gamma_i + A_i(1)x_i^\infty + B_i(1)y_i^\infty + C_i(1)u^\infty,$$

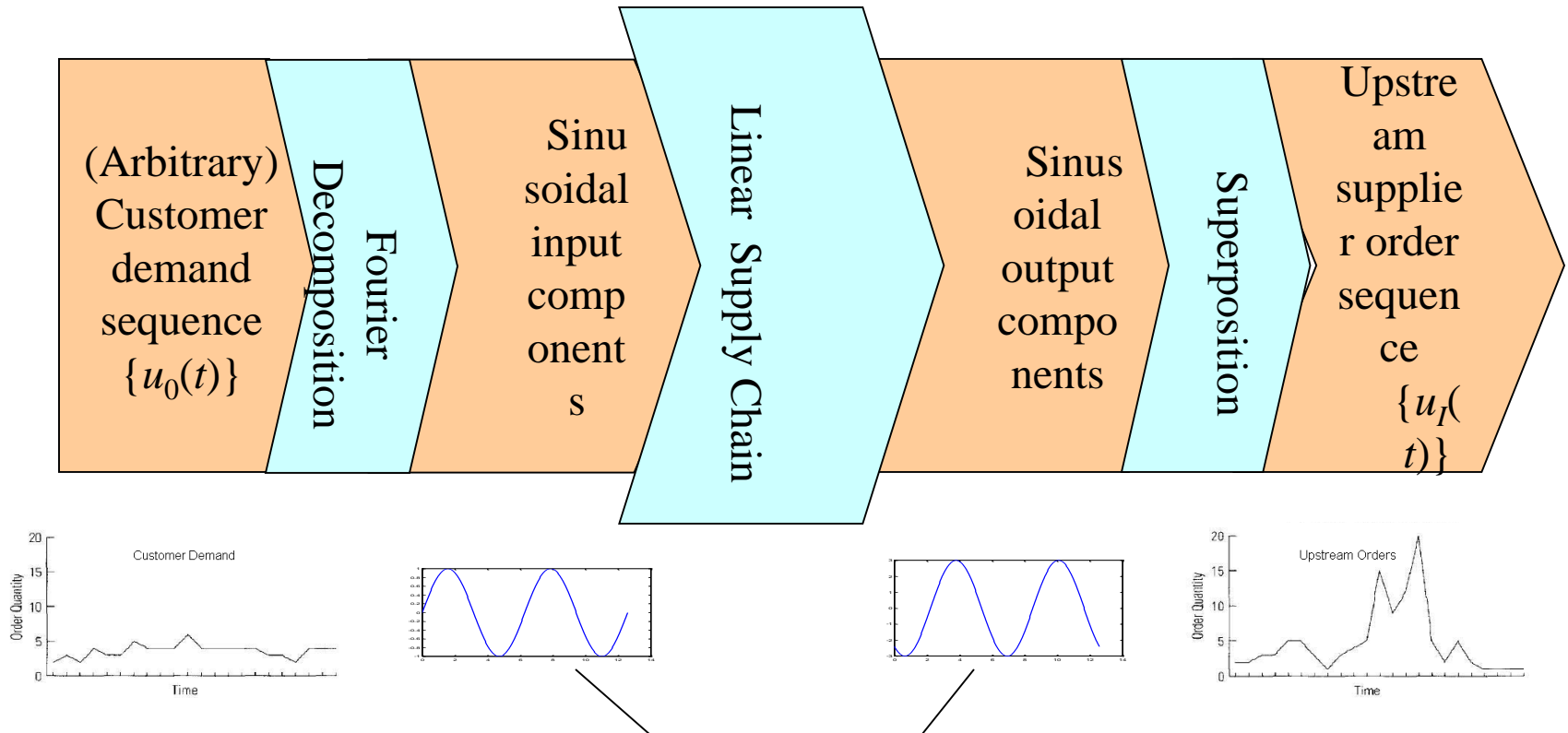
The Bullwhip Effect Metric



Worst-case RMSE amplification (L_2 norm gain)

$$\max_{\forall \{\bar{u}_0\} \neq 0} \frac{\left(\sum_{t=0}^{\infty} \bar{u}_I^2(t) \right)^{\frac{1}{2}}}{\left(\sum_{t=0}^{\infty} \bar{u}_0^2(t) \right)^{\frac{1}{2}}}, \text{ where } \bar{u}_i(t) := u_i(t) - u^\infty, i = 0, I.$$

Frequency Response Analysis



Frequency preserves; the amplitude and phase angle changes according to the transfer function.

The Bullwhip Effect Metrics

Time domain

- ✓ Worst-case RMSE amplification
(L_2 norm gain)

$$\max_{\forall \{\bar{u}_0\} \neq 0} \frac{\left(\sum_{t=0}^{\infty} \bar{u}_I^2(t) \right)^{\frac{1}{2}}}{\left(\sum_{t=0}^{\infty} \bar{u}_0^2(t) \right)^{\frac{1}{2}}}$$

Frequency domain

- ✓ The peak value on the Bode plot
(H_{∞} norm)

$$\max_{\forall w \in [0, 2\pi)} |T_I(e^{jw})|$$

where T_I is the transfer function from customer demand to supplier I orders.

System Dynamics

- For each variable, take its difference from the equilibrium-state value

$\forall i, t > 0$

$$\bar{x}_i(t+1) = \bar{x}_i(t) + \bar{u}_i(t) - \bar{u}_{i-1}(t)$$

$$\bar{y}_i(t+1) = \bar{y}_i(t) + \bar{u}_i(t - l_i) - \bar{u}_{i-1}(t)$$

$$\bar{u}_i(t) = A_i(P)\bar{x}_i(t) + B_i(P)\bar{y}_i(t) + C_i(P)\bar{u}_{i-1}(t-1)$$

- In the frequency domain (z -transform):

$$(z-1)X_i(z) = U_i(z) - U_{i-1}(z)$$

$$(z-1)Y_i(z) = z^{-l_i}U_i(z) - U_{i-1}(z)$$

$$U_i(z) = A_i(z^{-1})X_i(z) + B_i(z^{-1})Y_i(z) + z^{-1}C_i(z^{-1})U_{i-1}(z)$$

Eliminating $X_i(z)$ and $Y_i(z)$,

$$U_i(z) = \frac{z^{-1}C_i(z^{-1}) - (z-1)^{-1}[A_i(z^{-1}) + B_i(z^{-1})]}{1 - (z-1)^{-1}[A_i(z^{-1}) + z^{-l_i}B_i(z^{-1})]}U_{i-1}(z), \quad i = 1, 2, \dots$$

The Transfer Function

- The transfer function from the customer demand to the upstream orders of supplier I is

$$T_I(z) := \prod_{i=1}^I T_{i-1,i}(z)$$

where

$$T_{i-1,i}(z) := \frac{z^{-1}C_i(z^{-1}) - (z-1)^{-1}[A_i(z^{-1}) + B_i(z^{-1})]}{1 - (z-1)^{-1}[A_i(z^{-1}) + z^{-l_i}B_i(z^{-1})]}$$

- To examine the bullwhip effect, check the H_∞ norm:
whether $\exists w \in [0, 2\pi)$, such that

$$|T_I(e^{jw})| = \prod_{i=1}^I |T_{i-1,i}(e^{jw})| > 1.$$



Deterministic Chain Results

Theorem 1 (Sufficient condition for instability)

Supplier $I+1$ in the deterministic (LTI) supply chain experiences the bullwhip effect if

$$\sum_{i=1}^I \frac{1 + B_i(1)l_i - C_i(1)}{A_i(1) + B_i(1)} > 0.$$

Corollary 1 (Homogeneous chain)

When $A_i = A$, $B_i = B$, $C_i = C$, and $l_i = l$, $\forall i$, all upstream suppliers experience the bullwhip effect if

$$\frac{1 + B(1)l - C(1)}{A(1) + B(1)} > 0.$$

Example (Homogeneous Chain)

- Order-up-to policy
 - Variable “order-up-to level”: two-period moving-average demand forecasting
 - Lead time $l = 2$

$$A(P) = -1, B(P) = 0, C(P) = \frac{1}{2} (1 + P)l = 1 + P$$

- By Corollary 1:

$$\frac{1 + B(1)l - C(1)}{A(1) + B(1)} = \frac{1 + 0 - 2}{-1 + 0} = 1 > 0.$$

The bullwhip effect exists for sure!



Outline

- Basics
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 - Advance order commitment



Motivation

- Stochastic Environment
 - Unreliable shipments, variable lead times, price fluctuations, etc.
 - Randomness may affect system stability (bullwhip effect)
- Develop analytical conditions to examine the existence of the bullwhip effect

System Dynamics

- Return to the time-domain formulation

$$\forall i, t > 0$$

$$\bar{x}_i(t+1) = \bar{x}_i(t) + \bar{u}_i(t) - \bar{u}_{i-1}(t)$$

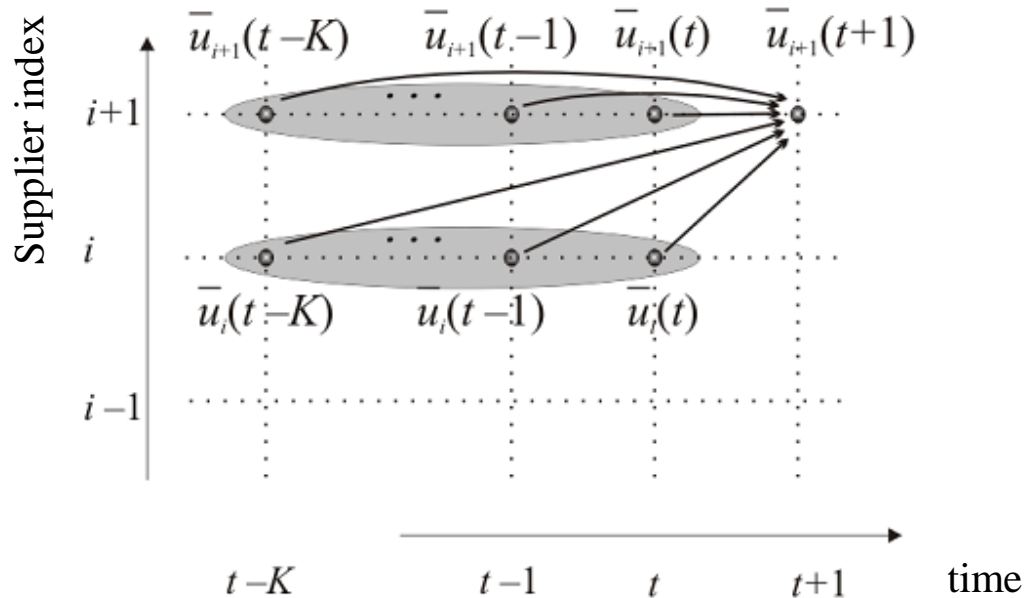
$$\bar{y}_i(t+1) = \bar{y}_i(t) + \bar{u}_i(t - l_i) - \bar{u}_{i-1}(t)$$

$$\bar{u}_i(t) = A_i(P)\bar{x}_i(t) + B_i(P)\bar{y}_i(t) + C_i(P)\bar{u}_{i-1}(t-1)$$

- Simple algebra gives

$$\bar{u}_i(t+1) = [1 + A_i(P) + P^{l_i} B(P)]\bar{u}_i(t) + [(1-P)C_i(P) - B_i(P) - A_i(P)]\bar{u}_{i-1}(t)$$

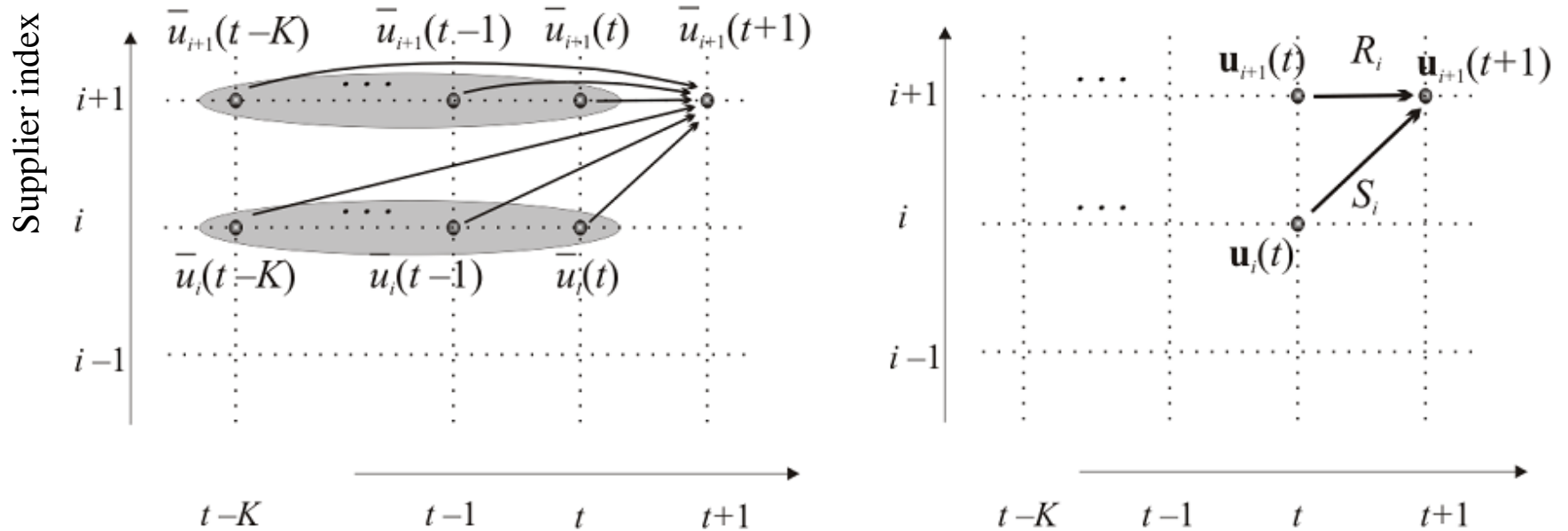
System Dynamics



- Simple algebra gives

$$\bar{u}_i(t+1) = [1 + A_i(P) + P^{l_i} B(P)]\bar{u}_i(t) + [(1-P)C_i(P) - B_i(P) - A_i(P)]\bar{u}_{i-1}(t)$$

System Dynamics



- Let $\mathbf{u}_i(t) := [\bar{u}_i(t), \bar{u}_i(t-1), \dots, \bar{u}_i(t-K)]^T$, then

$$\mathbf{u}_i(t+1) = R_i \cdot \mathbf{u}_i(t) + S_i \cdot \mathbf{u}_{i-1}(t), \forall i, t > 0$$

Markovian Jump Linear System (MJLS)

- Allow stochastic model parameters
- Consider

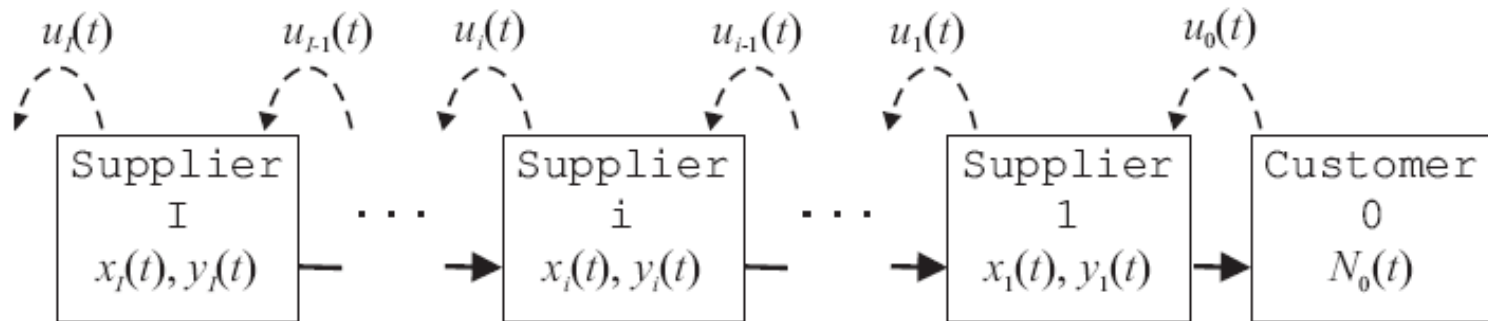
$$\mathbf{u}_i(t+1) = R_{\theta_i(t)} \cdot \mathbf{u}_i(t) + S_{\theta_i(t)} \cdot \mathbf{u}_{i-1}(t), \forall i, t > 0$$

where matrix pair $\{R_{\theta_i(t)}, S_{\theta_i(t)}\}$ takes value from a finite set according to an exogenous Markov chain $\{\theta_i(t)\}$.

$\theta_i(t) \in \mathcal{M}_i = \{1, 2, \dots, M_i\}$, with transition probability matrix

$$\mathcal{P}_i = [p_{mn}^i]_{M_i \times M_i}.$$

The Bullwhip Effect Metric



(Expected L_2 norm gain)

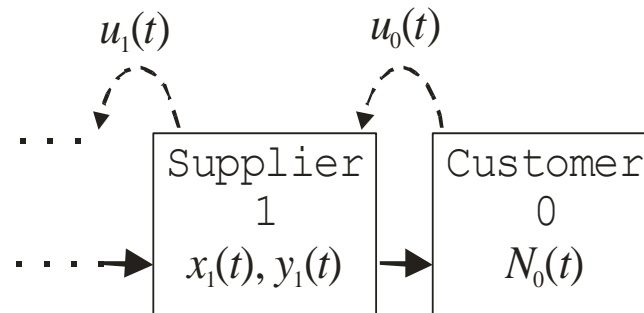
There is no bullwhip effect if

$$\max_{\forall \{\bar{u}_0\} \neq 0} \frac{E\left(\sum_{t=0}^{\infty} \bar{u}_I^2(t)\right)^{\frac{1}{2}}}{\left(\sum_{t=0}^{\infty} \bar{u}_0^2(t)\right)^{\frac{1}{2}}} \leq 1,$$

where the expectation is taken across Markov chain realizations.

Simplifying Assumption

- Homogeneous supply chain
 - Only need to consider one supplier, i.e., $i = 1$



- Drop subscript i ; e.g., $\theta(t) := \theta_i(t)$
- When $\theta(t) = m$, let $R_m := R_{\theta(t)}$, $S_m := S_{\theta(t)}$



Stability Results

Theorem 2 (Sufficient condition for stability)

The bullwhip effect is avoided if there exists non-zero, positive semi-definite matrices $G \geq 0$ and $H := \text{diag}(h_0, h_1, \dots, h_K) \geq 0$ such that

$$\begin{bmatrix} G & 0 \\ 0 & H \end{bmatrix} - \sum_{m=1}^M p_{nm} \begin{bmatrix} R_m & S_m \\ E & 0 \end{bmatrix}^T \begin{bmatrix} G & 0 \\ 0 & H \end{bmatrix} \begin{bmatrix} R_m & S_m \\ E & 0 \end{bmatrix} \geq 0, \forall n \in \mathcal{M},$$

where E is the identity matrix.

Theorem 3 (Necessary condition for stability)

The condition in Theorem 2 is also necessary if:

- a) the system is “weakly controllable” (Ji and Chizeck, 1988)
- b) the transition probabilities satisfy $p_{nm} \equiv p_m, \forall n \in \mathcal{M}$.



Stability Results

Corollary 2 (Deterministic chains)

The bullwhip effect is avoided in deterministic LTI chains **if and only if** there exists non-zero matrices $G \geq 0$ and $H := \text{diag}(h_0, h_1, \dots, h_K) \geq 0$ such that

$$\begin{bmatrix} G & 0 \\ 0 & H \end{bmatrix} - \begin{bmatrix} R & S \\ E & 0 \end{bmatrix}^T \begin{bmatrix} G & 0 \\ 0 & H \end{bmatrix} \begin{bmatrix} R & S \\ E & 0 \end{bmatrix} \geq 0,$$

where E is the identity matrix.

Example (Deterministic Chain)

A family of “order-based” policies with advance demand information

$$\bar{u}_1(t+1) = \alpha \cdot \bar{u}_1(t) + [\beta_0 + \beta_1 P + \cdots + \beta_K P^K] \bar{u}_0(t)$$

where

$$|\alpha| < 1, \alpha + \beta_0 + \beta_1 + \cdots + \beta_K = 1 \quad (\text{properness})$$

$$\alpha, \beta_0, \beta_1, \cdots, \beta_K \geq 0 \quad (\text{with advance demand information})$$

Note:

$$R = \begin{bmatrix} \alpha & 0 & \cdots & 0 & 0 \\ 1 & 0 & \cdots & 0 & 0 \\ 0 & 1 & \cdots & 0 & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & \cdots & 1 & 0 \end{bmatrix}, S = \begin{bmatrix} \beta_0 & \beta_1 & \cdots & \beta_K \\ 0 & 0 & \cdots & 0 \\ \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \cdots & 0 \end{bmatrix}.$$

When G and H are as follows:

$$G = \begin{bmatrix} 1 & 0 & \cdots & 0 & 0 \\ 0 & \sum_{k=1}^K \beta_k & \cdots & 0 & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & \cdots & \beta_{K-1} + \beta_K & 0 \\ 0 & 0 & \cdots & 0 & \beta_K \end{bmatrix}, H = \begin{bmatrix} \beta_0 & 0 & \cdots & 0 \\ 0 & \beta_1 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \beta_K \end{bmatrix},$$

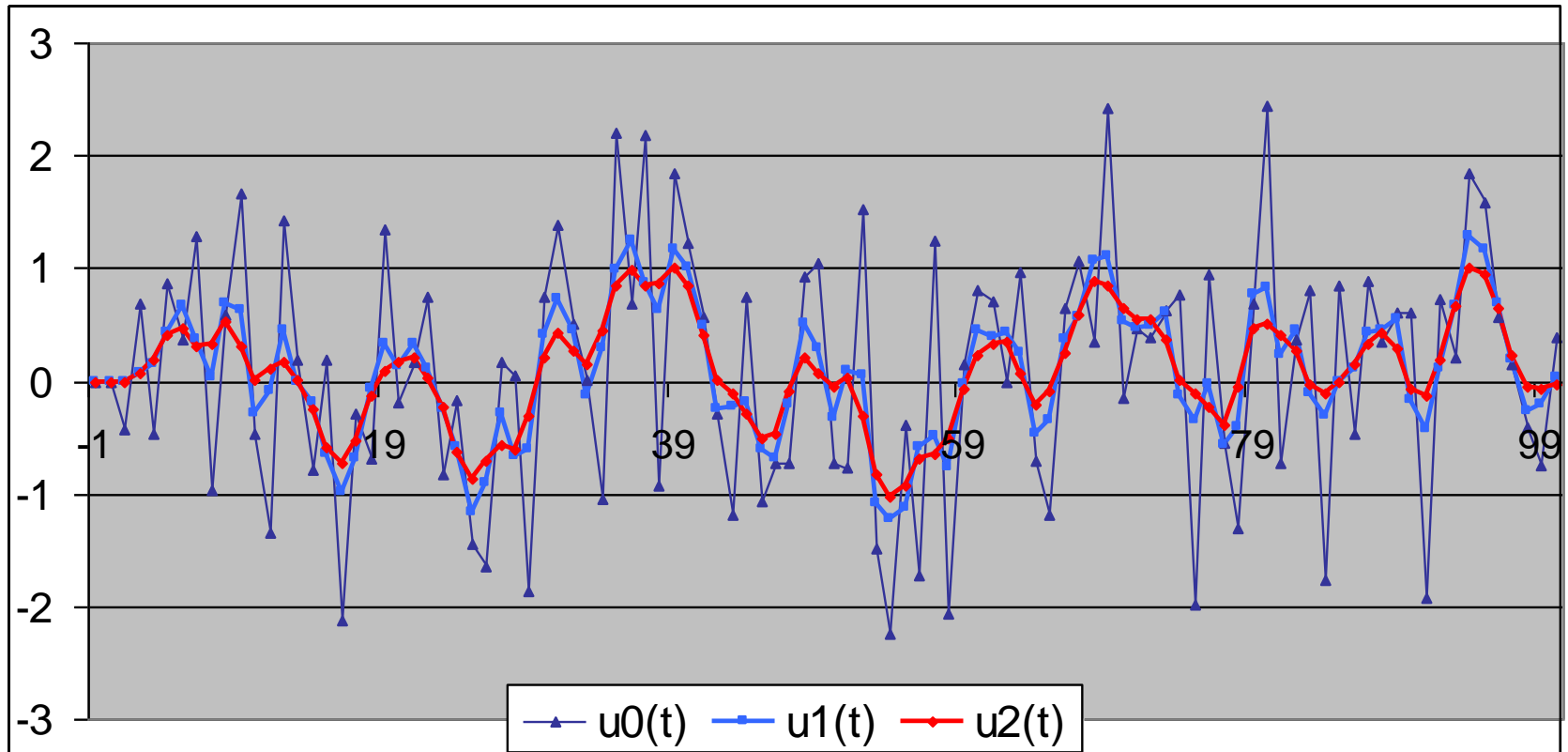
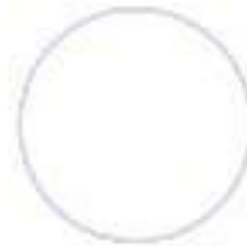
the stability condition in Corollary 2:

$$\begin{bmatrix} G & 0 \\ 0 & H \end{bmatrix} - \begin{bmatrix} R & S \\ E & 0 \end{bmatrix}^T \begin{bmatrix} G & 0 \\ 0 & H \end{bmatrix} \begin{bmatrix} R & S \\ E & 0 \end{bmatrix} \geq 0$$

is satisfied. There is no bullwhip effect.



Simulation



$\bar{u}_0(t) \sim \text{i.i.d. Gaussian}(0, 1)$, $\alpha = 0.4$, $\beta_0 = \beta_1 = 0.3$ ($K=1$)

Example (Stochastic Chain)

- Shipments lost with probability p
 - Transition probability matrix ($|\mathcal{M}| = 2$)

$$\mathcal{P} = \begin{bmatrix} 1-p & p \\ 1-p & p \end{bmatrix}$$

- The order-up-to policy
 - Safe mode

$$\bar{u}_1(t+1) = 2\bar{u}_0(t) - \bar{u}_0(t-2)$$

- Loss mode

$$\bar{u}_1(t+1) = \bar{u}_1(t-2) + 2\bar{u}_0(t) - \bar{u}_0(t-2) + u^\infty$$

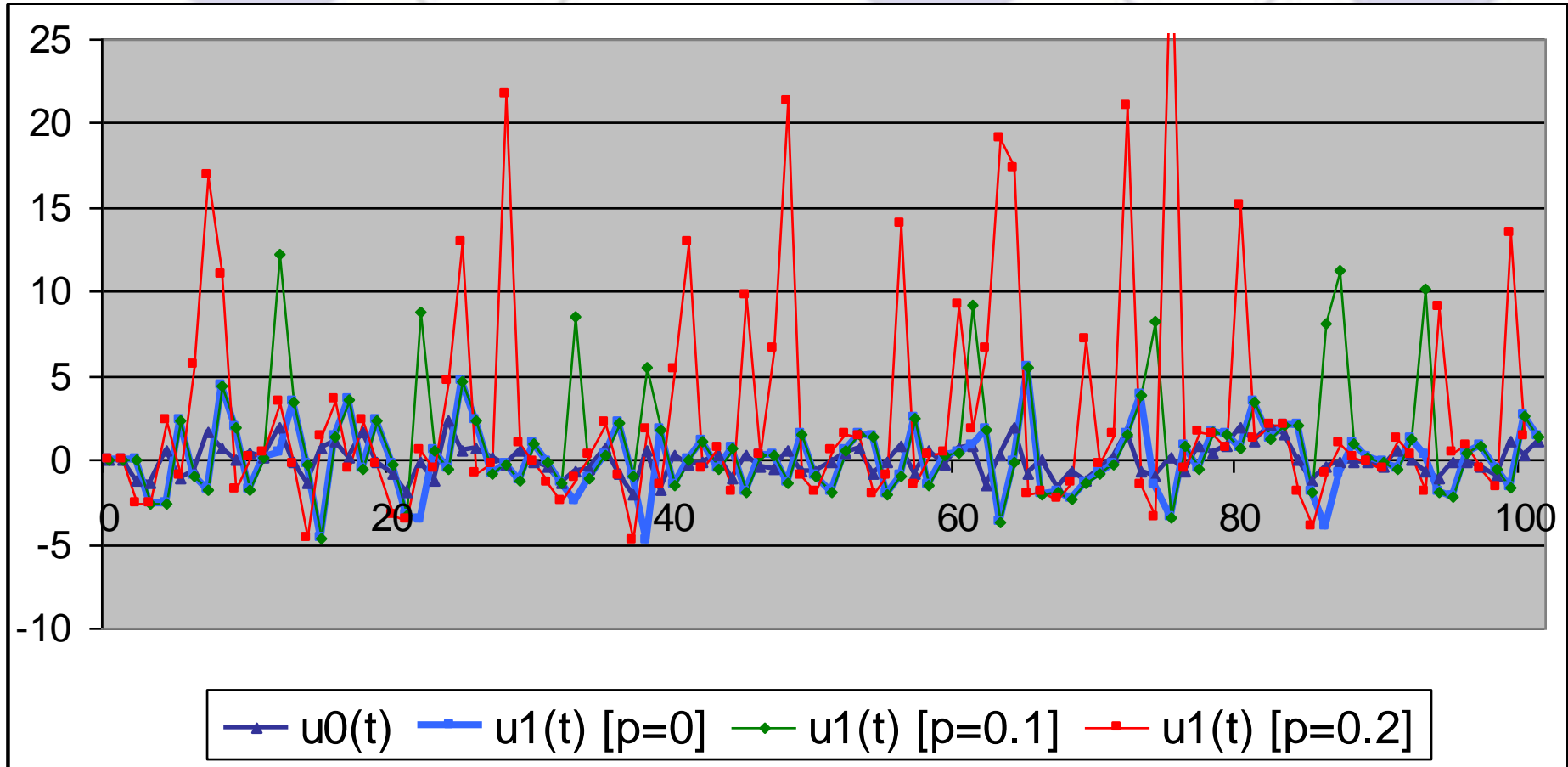
Example (Stochastic Chain)

- Let $\mathbf{u}_1(t) := [\bar{u}_1(t), \bar{u}_1(t-1), \bar{u}_1(t-2), u^\infty]^\top$, $\mathbf{u}_0(t) := [\bar{u}_0(t), \bar{u}_0(t-1), \bar{u}_0(t-2), u^\infty]^\top$.

- Safe mode ($m = 1$) $R_1 = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}, S_1 = \begin{bmatrix} 2 & 0 & -1 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix},$

- Loss mode ($m = 2$) $R_2 = \begin{bmatrix} 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}, S_2 = \begin{bmatrix} 2 & 0 & -1 & 1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}.$

- Result
 - A numerical search reveals that matrices G and H satisfying Theorem 2 do not exist, $\forall p \in [0, 1]$
 - The bullwhip effect exists

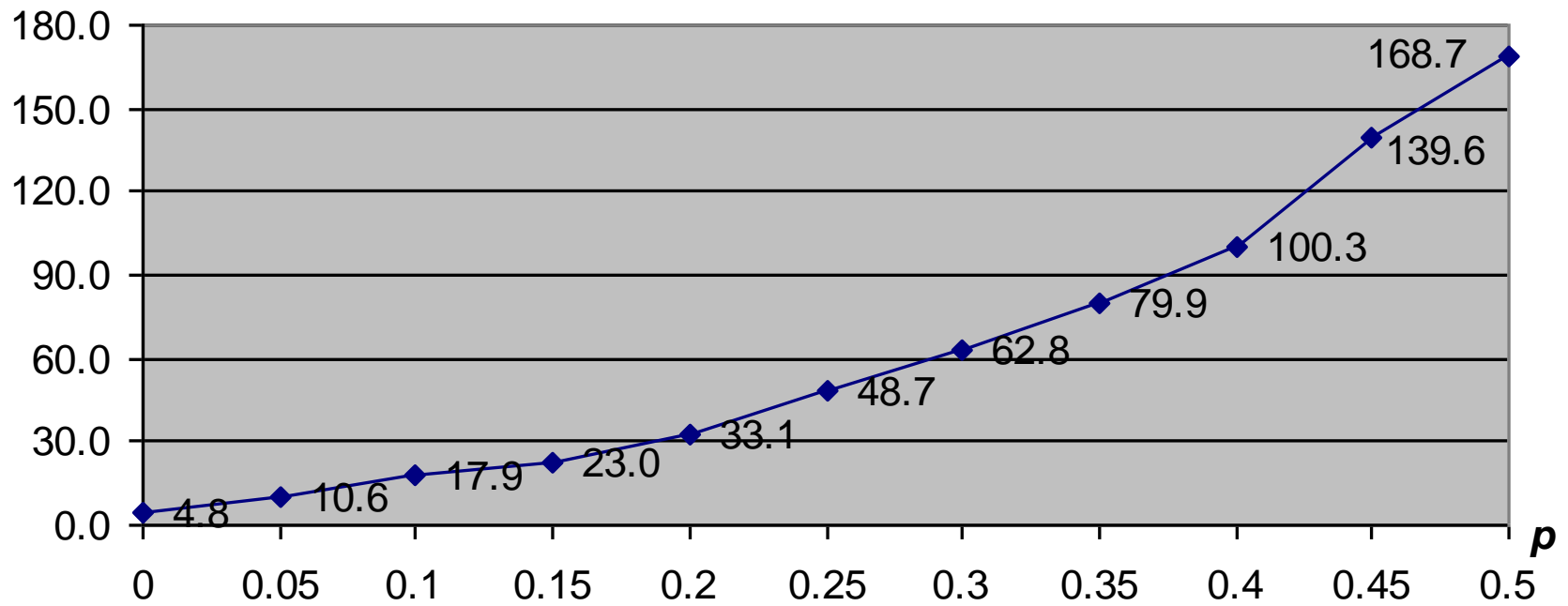


$\bar{u}_0(t) \sim \text{i.i.d. Gaussian}(0, 1)$

Plot $\bar{u}_0(t), \bar{u}_1(t)$ under both deterministic condition ($p = 0$) and stochastic conditions ($p = 0.1, p = 0.2$)



Average variance amplification (MJLS)



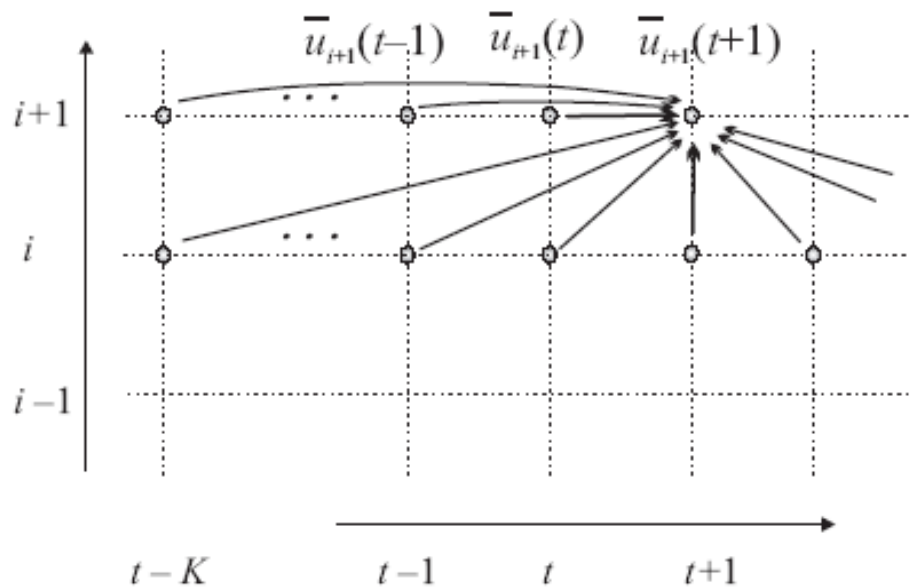
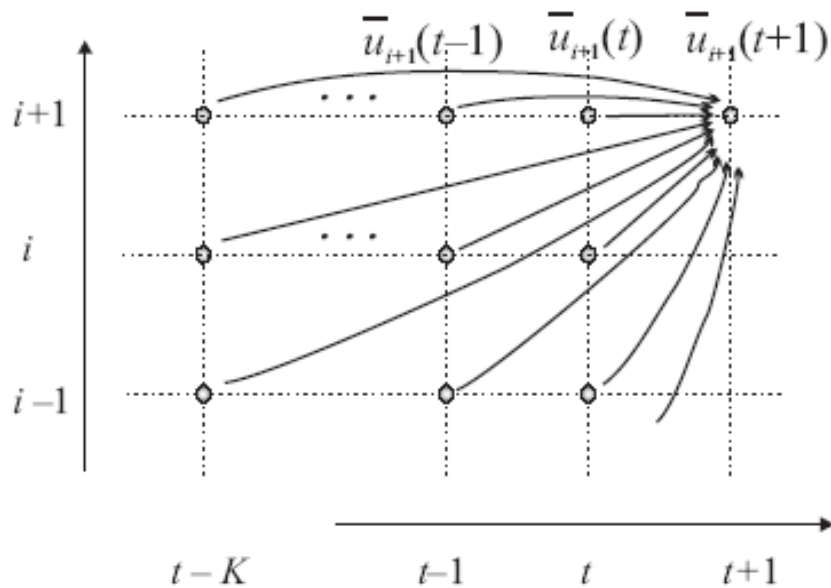


Outline

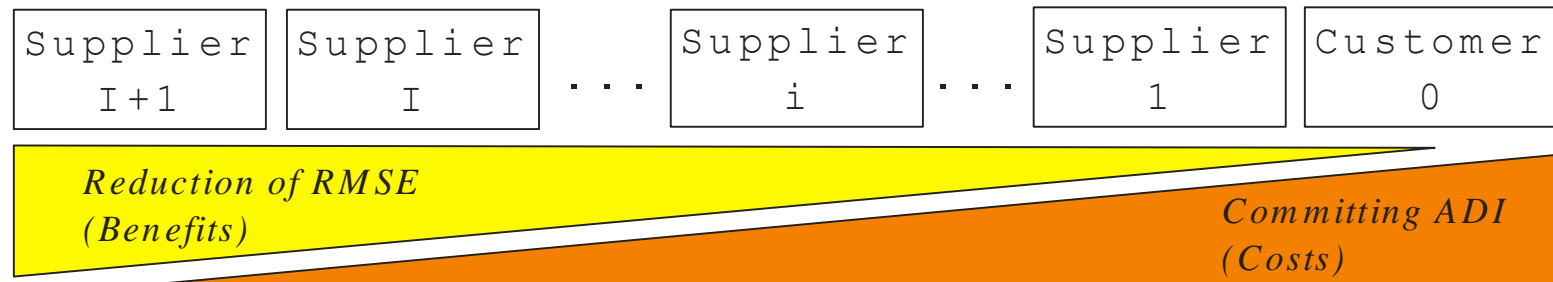
- Basics
 - Background
 - The bullwhip effect
- Deterministic chain stability
 - System formulation
 - Analytical results
- Stochastic chain stability
 - System formulation
 - Analytical results
- Toward optimality
 - Decentralized negotiations
 - Advance order commitment

Achieving Stability / Optimality

- Sharing information among suppliers (Lee *et al.*, 2000; Simchi-Levi and Zhao, 2003; etc.)
- Advance demand information (ADI) (Hariharan and Zipkin, 1995; Ouyang and Daganzo, 2005; etc.)



Advance Order Commitments



- Downstream suppliers committing to advance orders ...
 - Reduces upstream order variations (introduces benefits)
 - Increases own costs
- Able to quantify benefits / costs for every supplier
- Idealized optimum
 - Coordination among suppliers
 - Total benefits exceeds total costs for sufficiently long chains



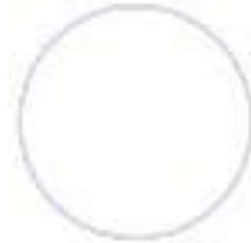
Decentralized Negotiations

- Negotiations
 - Neighboring suppliers negotiate discounts for advance order commitments and RMSE reductions
- If suppliers are not greedy
 - system reaches the same optimum as if there was a coordinating agent
- If suppliers are greedy and impatient
 - system may reach sub-optimum



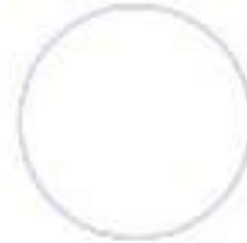
Summary

- System-control framework
 - Supply chains
 - The bullwhip effect
- System-level stability
 - Deterministic inhomogeneous chain
 - Stochastic homogeneous chain
- System-level optimality
 - Advance order commitment
 - Decentralized negotiations



Questions?

Thank you!



Back-up Slides

Bullwhip Effect Metrics

Time domain

- ✓ Worst-case variance amplification

(L₂ norm)

$$\max_{\forall \{\bar{u}_0\} \neq 0} \frac{\left(\sum_{t=0}^{\infty} \bar{u}_I^2(t) \right)^{\frac{1}{2}}}{\left(\sum_{t=0}^{\infty} \bar{u}_0^2(t) \right)^{\frac{1}{2}}}$$

- × Average-case variance amplification for white-noise input sequence
- × Variance amplification with certain demand process

Frequency domain

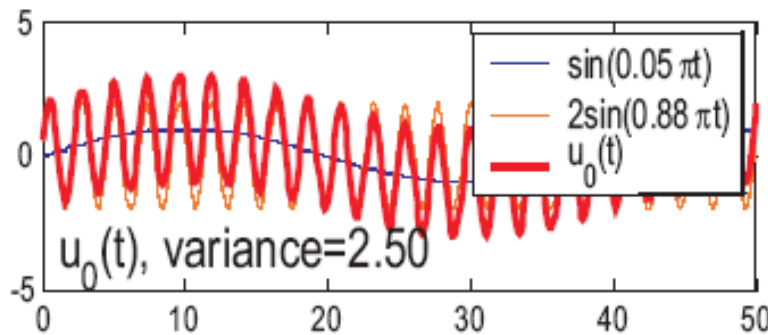
- ✓ The peak value on the Bode plot
- (H_∞ norm)

$$\max_{\forall w \in [0, 2\pi)} |T_I(e^{jw})|$$

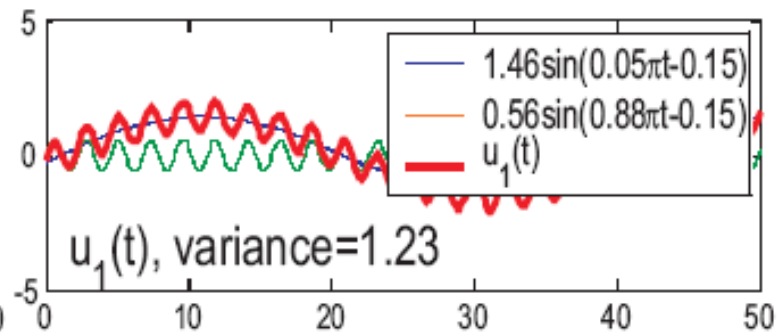
- × Noise-bandwidth
- (H₂ norm)

Bullwhip Effect Metrics

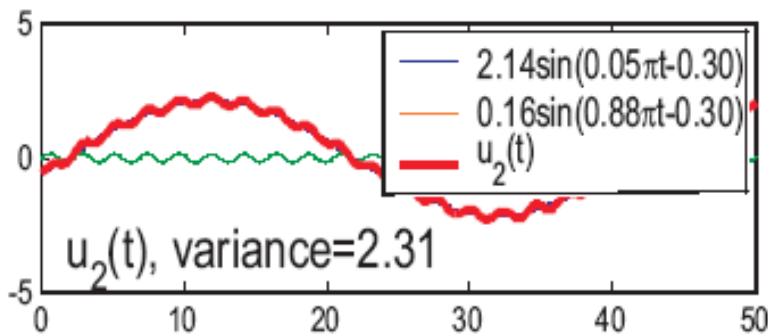
- The “general replenishment rule” proposed by Dejonckheere (2003a)
- Two sinusoidal input signals: $\sin(0.05\pi t) + 2 \sin(0.88\pi t)$
- Amplification ratio: 1.464 and 0.282 through each stage; phase change: -0.1507 and -0.1515



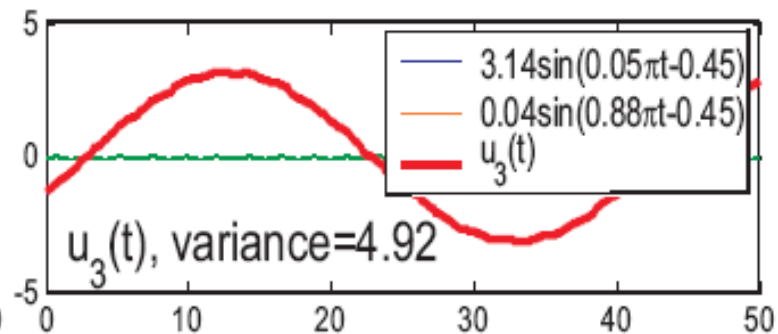
(a)



(b)



(c)



(d)

Theorem 1 Proof (Deterministic Chains)

Boyd and Desoer (1985) showed that $\log \|T_I(e^\sigma)\|$ is subharmonic with regard to σ and satisfies the Poisson Inequality:

$$\log |T_I(e^y)| \leq \frac{1}{\pi} \int_{-\infty}^{+\infty} \log |T_I(e^{jw})| \frac{y dw}{y^2 + w^2}, \forall y \in (0, \infty)$$

Divide both sides by y , and let $y \rightarrow 0^+$,

$$\begin{aligned} \lim_{y \rightarrow 0^+} \frac{1}{y} \log |T_I(e^y)| &\leq \lim_{y \rightarrow 0^+} \frac{1}{\pi} \int_{-\infty}^{+\infty} \log |T_I(e^{jw})| \frac{dw}{y^2 + w^2} \\ &= \frac{1}{\pi} \int_{-\infty}^{+\infty} \log |T_I(e^{jw})| \frac{dw}{w^2} \end{aligned}$$

Note that $T_I(z) = \prod_{i=1}^I T_{i-1,i}(z)$, therefore

$$\lim_{y \rightarrow 0^+} \frac{1}{y} \log |T_I(e^y)| = \sum_{i=1}^I \lim_{y \rightarrow 0^+} \frac{1}{y} \log |T_{i-1,i}(e^y)|.$$

By Taylor expansion at $y = 0$,

$$\begin{aligned} T_{i-1,i}(e^y) &= T_{i-1,i}(e^y)|_{y=0} + [T_{i-1,i}(e^y)]'_y|_{y=0} \cdot y + o(y) \\ &= 1 + \frac{1 + B_i(1)l_i - C_i(1)}{A_i(1) + B_i(1)} \cdot y + o(y) \end{aligned}$$

Theorem 1 Proof (Deterministic Chains)

At the neighborhood of 0^+ , $\frac{1+B_i(1)l_i-C_i(1)}{A_i(1)+B_i(1)} \cdot y + o(y) \ll 1$, therefore

$$\begin{aligned} |T_{i-1,i}(e^y)| &= \left| 1 + \frac{1+B_i(1)l_i-C_i(1)}{A_i(1)+B_i(1)} \cdot y + o(y) \right| \\ &= 1 + \frac{1+B_i(1)l_i-C_i(1)}{A_i(1)+B_i(1)} \cdot y + o(y), \end{aligned}$$

By l'Hôpital's Rule,

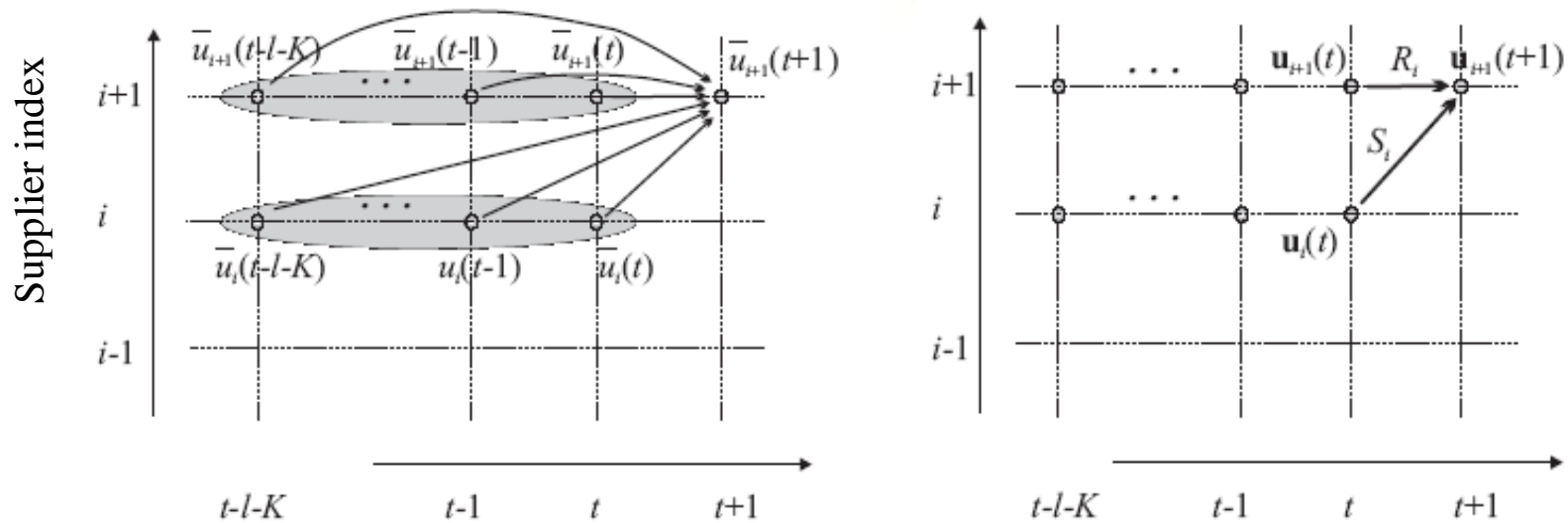
$$\begin{aligned} \lim_{y \rightarrow 0^+} \frac{1}{y} \log |T_{i-1,i}(e^y)| &= \lim_{y \rightarrow 0^+} \frac{|T_{i-1,i}(e^y)|'_y}{|T_{i-1,i}(e^y)|} \\ &= \lim_{y \rightarrow 0^+} \frac{\frac{1+B_i(1)l_i-C_i(1)}{A_i(1)+B_i(1)} + O(y)}{1 + \frac{1+B_i(1)l_i-C_i(1)}{A_i(1)+B_i(1)} \cdot y + o(y)} \\ &= \frac{1+B_i(1)l_i-C_i(1)}{A_i(1)+B_i(1)}. \end{aligned}$$

we have

$$\sum_{i=1}^I \frac{1+B_i(1)l_i-C_i(1)}{A_i(1)+B_i(1)} \leq \frac{1}{\pi} \int_{-\infty}^{+\infty} \log |T_I(e^{jw})| \frac{dw}{w^2}.$$

[Back](#)

System Dynamics



- Let $\mathbf{u}_i(t) := [\bar{u}_i(t), \bar{u}_i(t-1), \dots, \bar{u}_i(t-l_i-K)]'$, then

$$\mathbf{u}_i(t+1) = R_i \cdot \mathbf{u}_i(t) + S_i \cdot \mathbf{u}_{i-1}(t), \forall i, t > 0$$

where

$$R_i = \begin{bmatrix} \alpha_0^i & \alpha_1^i & \alpha_2^i & \dots & \alpha_{K+l_i-1}^i & \alpha_{K+l_i}^i \\ 1 & 0 & 0 & \dots & 0 & 0 \\ 0 & 1 & 0 & \dots & 0 & 0 \\ 0 & 0 & 1 & \dots & 0 & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & 0 & \dots & 1 & 0 \end{bmatrix} \quad S_i = \begin{bmatrix} \beta_0^i & \beta_1^i & \dots & \beta_{K+1}^i & 0 & \dots & 0 \\ 0 & 0 & \dots & 0 & 0 & \dots & 0 \\ \vdots & \vdots & & \vdots & \vdots & \dots & \vdots \\ 0 & 0 & \dots & 0 & 0 & \dots & 0 \end{bmatrix}$$

Almost-surely stability (Mariton 1995)

Theorem 2 (Time stability). The system described above is almost surely stable if for $w(t) \equiv 0$ and every initial state $(u(0), \theta(0))$,

$$Pr\{\lim_{t \rightarrow \infty} \|u(t)\| = 0\} = 1.$$

With an ergodic Markov chain $\theta(t)$, almost sure stability is achieved when

$$\sum_{m=1}^M \pi_m |\sigma(R_m)| < 1$$

where π_m is the long-run probability of mode m , and $|\sigma(R_m)|$ is the spectral radius of matrix R_m .

Variance Amplification Bounds

Theorem 4 (Bounds for bullwhip effect metric)

The bullwhip effect metric (variance amplification) is bounded by μ ($\mu > 0$), if there exists non-zero matrices $G \geq 0$ and $H := \text{diag}(h_0, h_1, \dots, h_K) \geq 0$ such that

$$\begin{bmatrix} G & 0 \\ 0 & \mu \cdot H \end{bmatrix} - \sum_{m=1}^M p_{nm} \begin{bmatrix} R_m & S_m \\ E & 0 \end{bmatrix}^T \begin{bmatrix} G & 0 \\ 0 & H \end{bmatrix} \begin{bmatrix} R_m & S_m \\ E & 0 \end{bmatrix} \geq 0, \forall n \in \mathcal{M},$$

where E is the identity matrix.

Theorem 2 Proof (Stochastic Chains)

Define a scalar function $V(u) := gu^2$. $g \geq 0$, we have

$$\mathbb{E}_{\theta(0), \dots, \theta(T)} \sum_{t=0}^T [V(u(t+1)) - V(u(t))] = \mathbb{E}_{\theta(0), \dots, \theta(T)} V(u(T+1)) \geq 0$$

we need to prove that

$$\mathbb{E} \left[\sum_{t=0}^{\infty} u^2(t) \right] \leq \sum_{t=0}^{\infty} w^2(t).$$

Define the expected L_2 norm of a truncated stochastic sequence $\{u(0), u(1), \dots, u(T)\}$ as

$$\begin{aligned} & \mathbb{E}_{\theta(0), \dots, \theta(T-1)} \left[\sum_{t=0}^T u^2(t) \right] \\ & \leq \sum_{t=0}^T w^2(t) + \mathbb{E}_{\theta(0), \dots, \theta(T)} \left[\sum_{t=0}^T (u^2(t) - w^2(t) + V(u(t+1)) - V(u(t))) \right] \\ & = \sum_{t=0}^T w^2(t) + \sum_{t=0}^T \mathbb{E}_{\theta(0), \dots, \theta(T)} [u^2(t) - w^2(t) + gu^2(t+1) - gu^2(t)] \\ & = \sum_{t=0}^T w^2(t) - \sum_{t=0}^T \mathbb{E}_{\theta(0), \dots, \theta(t)} \left\{ [\mathbf{u}(t)' \mathbf{w}(t)'] F_{\theta(t)} \begin{bmatrix} \mathbf{u}(t) \\ \mathbf{w}(t) \end{bmatrix} \right\} \end{aligned}$$

Theorem 2 Proof (Stochastic Chains)

where

$$F_{\theta(t)} := \begin{bmatrix} G & 0 \\ 0 & H \end{bmatrix} - \begin{bmatrix} R_{\theta(t)} & S_{\theta(t)} \\ I & 0 \end{bmatrix}' \begin{bmatrix} G & 0 \\ 0 & H \end{bmatrix} \begin{bmatrix} R_{\theta(t)} & S_{\theta(t)} \\ I & 0 \end{bmatrix},$$

and

$$\mathbf{u}(t) = [u(t), u(t-1), \dots, u(t-l-K)]', \quad \mathbf{w}(t) = [w(t), w(t-1), \dots, w(t-l-K)]'.$$

For $\forall t, 0 \leq t \leq T$,

$$\begin{aligned} & \mathbb{E}_{\theta(0), \dots, \theta(t)} \left\{ [\mathbf{u}(t)' \mathbf{w}(t)'] F_{\theta(t)} \begin{bmatrix} \mathbf{u}(t) \\ \mathbf{w}(t) \end{bmatrix} \right\} \\ (a) = & \mathbb{E}_{\theta(0), \dots, \theta(t-1)} \left\{ \mathbb{E}_{\theta(t)} \left[[\mathbf{u}(t)' \mathbf{w}(t)'] F_{\theta(t)} \begin{bmatrix} \mathbf{u}(t) \\ \mathbf{w}(t) \end{bmatrix} \middle| \theta(0), \dots, \theta(t-1) \right] \right\} \\ (b) = & \mathbb{E}_{\theta(0), \dots, \theta(t-1)} \left\{ [\mathbf{u}(t)' \mathbf{w}(t)'] \mathbb{E}_{\theta(t)} [F_{\theta(t)} | \theta(0), \dots, \theta(t-1)] \begin{bmatrix} \mathbf{u}(t) \\ \mathbf{w}(t) \end{bmatrix} \right\} \\ (c) = & \mathbb{E}_{\theta(0), \dots, \theta(t-1)} \left\{ [\mathbf{u}(t)' \mathbf{w}(t)'] \left(\sum_{m=1}^M p_{\theta(t-1)m} F_m \right) \begin{bmatrix} \mathbf{u}(t) \\ \mathbf{w}(t) \end{bmatrix} \right\} \\ (d) \geq & 0. \end{aligned}$$

□

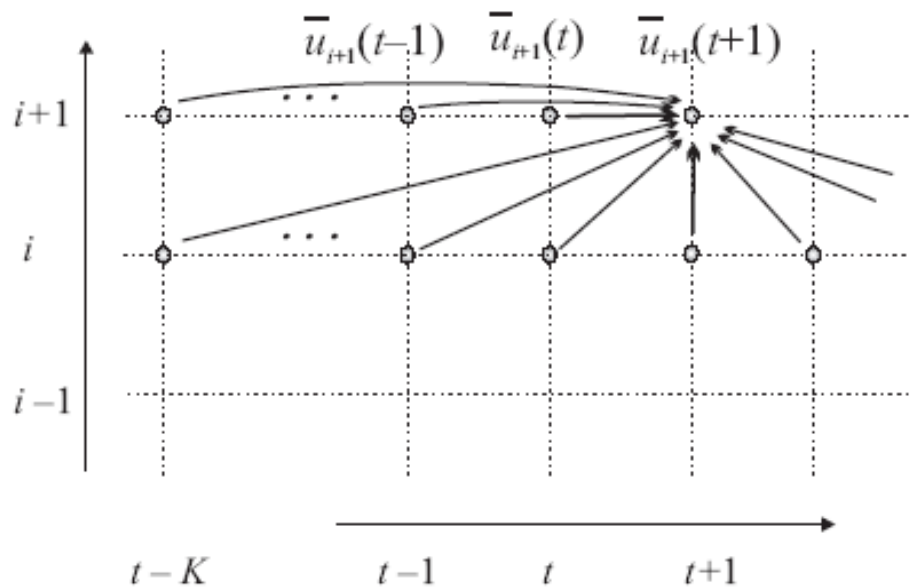
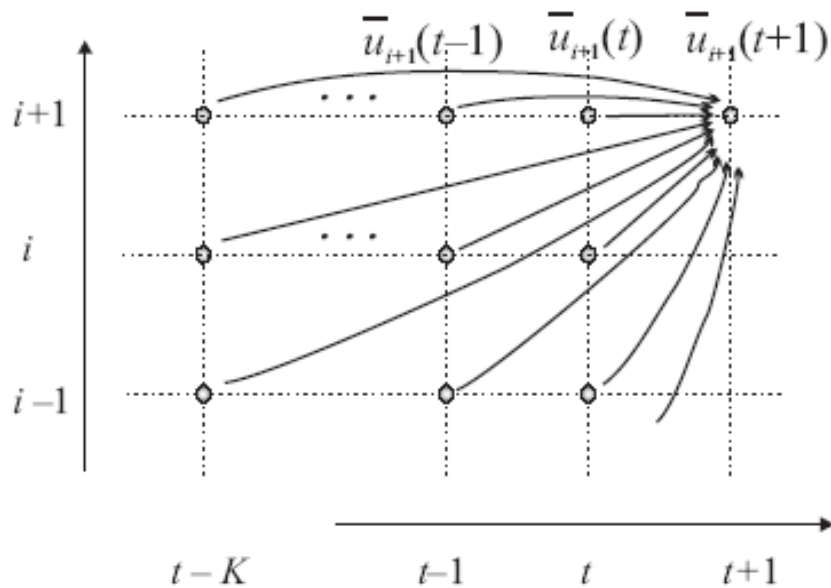


Outline

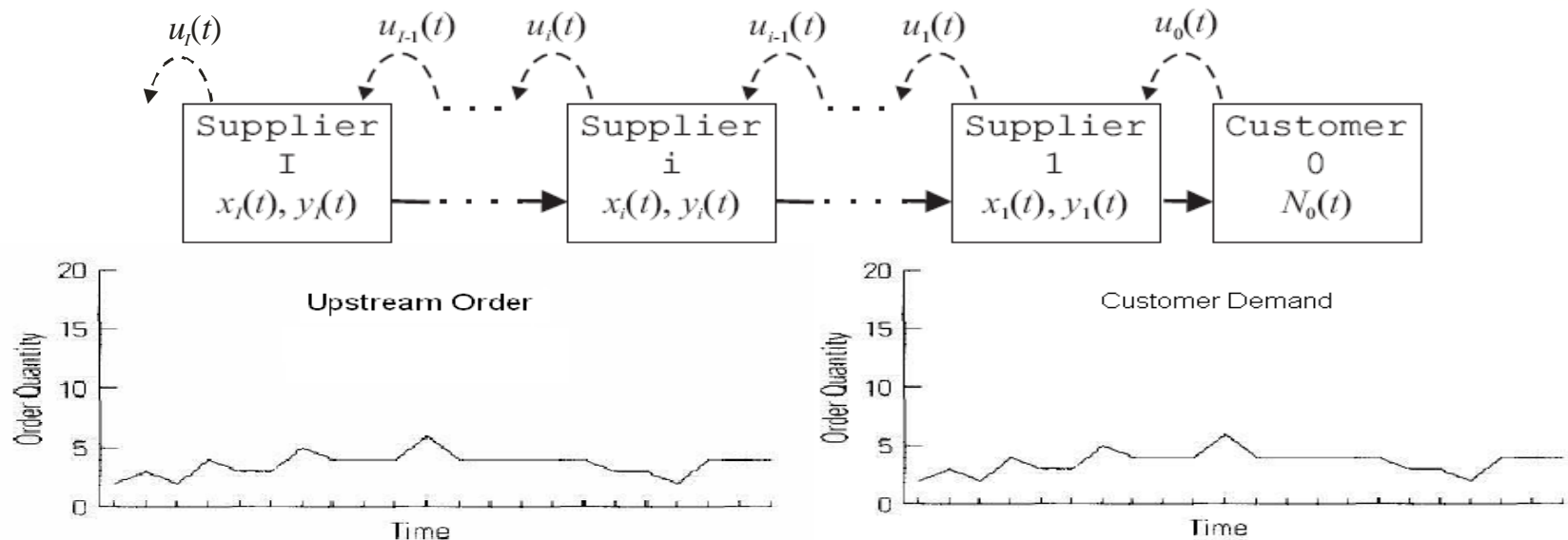
- Basics
 - Background
 - The bullwhip effect
- Deterministic inhomogeneous chain
 - System formulation
 - Analytical results
- Stochastic homogeneous chain
 - System formulation
 - Analytical results
- A decentralized contracting scheme
 - Advance order commitment

Achieving System-level Stability

- Sharing information among suppliers (Lee *et al.* 2000; Simchi-Levi and Zhao, 2003; etc.)
- Advance demand information (Hariharan and Zipkin 1995; Ouyang and Daganzo, 2005; etc.)



Advance Order Commitments



- Downstream suppliers placing advance orders ...
 - Reduces upstream order variations (introduces benefits)
 - Increases own costs
- Able to quantify benefits / costs for every supplier
- Feasibility
 - Total benefits exceeds total costs (an “imaginary broker” can profit)



A Decentralized Option

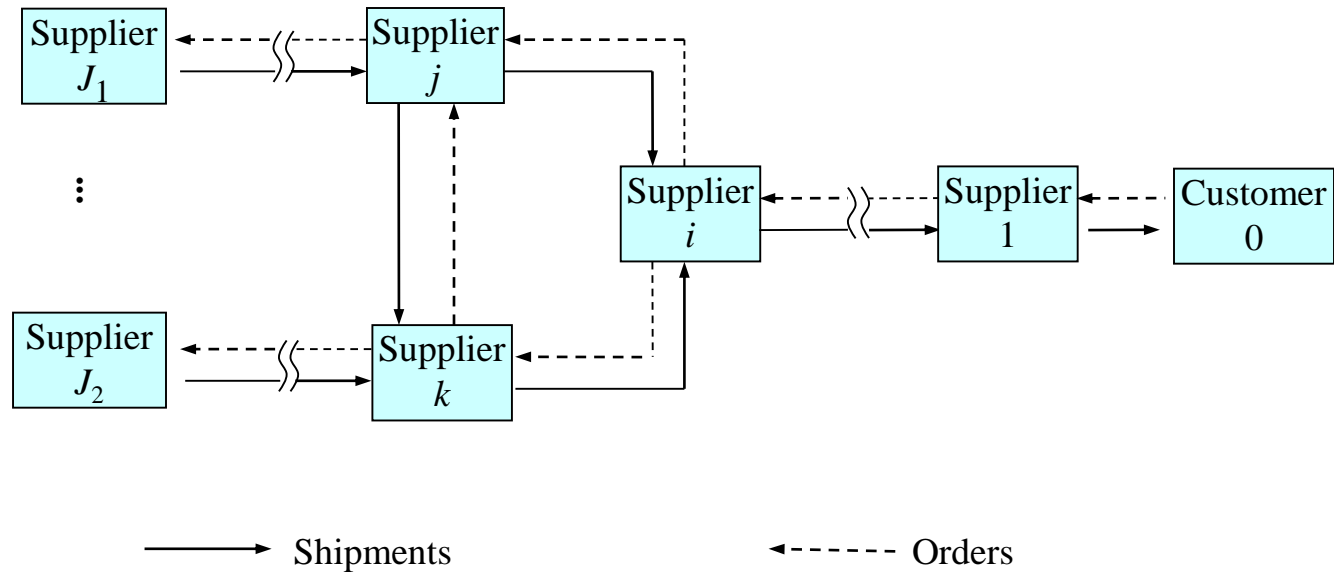
- Contracting
 - Neighboring suppliers negotiate discounts for advance order commitments and variance reductions
 - System reaches the same equilibrium as if there was a “broker”
- No coordinating agent is necessary



Extensions

- Supply network
- Nonlinear system
 - ordering policy
 - operation (e.g., load-dependent lead time)
- Endogenous MJLS

Supply Networks



- $G = (V \cup W, E)$, $V = \{\text{supplier node}\}$, $W = \{\text{customer node}\}$, $E = \{\text{ordering arc}\}$;
- Inventory $\mathbf{x}(t) = \{x_1(t), x_2(t), \dots, x_n(t)\}$, orders $\mathbf{u}(t) = \{u_{ij}(t): (i, j) \in E\}$;
- lead time $\{l_{ij}\}$, shipment loss $\{\rho_{ij}\}$, $\forall (i, j) \in E$.



Supply Networks

System dynamics:

Inventory

dynamics: $x_i(t+1) = x_i(t) + \sum_{j:(i,j) \in A} (\rho_{ij} \cdot u_{ij}(t - l_{ij})) - \sum_{k:(k,i) \in E} u_{ki}(t)$, $i = 1, \dots, n$, $\mathbf{x}(0) = \mathbf{0}$;

Ordering
policy:

$$u_{ij}(t) = \sum_{k:(i,k) \in E} A_{ik}(P) u_{ik}(t) + \frac{1}{\rho_{ij}} \left[\alpha_{ij}^0 x_i(t) + B_{ij}(P) \cdot \sum_{k:(k,i) \in E} u_{ki}(t) \right], \forall (i,j) \in E, i > 0$$

Supply Networks

Motion Equations:

$$X_i(z) = \frac{1}{z-1} \left[\sum_{j:(i,j) \in E} (\rho_{ij} \cdot U_{ij}(z) \cdot z^{-l_{ij}}) - \sum_{k:(k,i) \in E} U_{ki}(z) \right], i = 1, \dots, n$$

$$U_{ij}(z) = \frac{(B_{ij}(z^{-1}) - \alpha_{ij}^0 (z-1)^{-1})}{\rho_{ij} (1 - A_{ij}(z^{-1}) - \alpha_{ij}^0 z^{-l_{ij}} (z-1)^{-1})} \cdot \sum_{k:(k,i) \in E} U_{ki}(z) + \frac{\sum_{k:(i,k) \in E, k \neq j} [U_{ik}(z) \cdot (\alpha_{ij}^0 \rho_{ik} z^{-l_{ik}} (z-1)^{-1} + A_{ik}(z^{-1}))]}{\rho_{ij} (1 - A_{ij}(z^{-1}) - \alpha_{ij}^0 z^{-l_{ij}} (z-1)^{-1})}, \forall (i,j) \in E, i > 0$$

Transfer Function:

$$T_i(z) = \frac{\sum_{j:(i,j) \in E} U_{ij}(z)}{U_{01}(z)}$$