
A traffic-like model for supply chains

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1. Introduction

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 - ▶ A supply chain is a string (or network) of processors (or stations) along which goods (or parts) are circulating
- ▶ At each processor, the goods undergo a transformation.
- ▶ Each processor requires a certain throughput time T to process a given good
- ▶ Each processor has a limited capacity q (maximum number of goods it is able to deliver per unit of time)

- In front of each processor, goods can be stored in buffer queues while waiting for being treated

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- ▶ Goods are picked up in the queues according to a given policy.
 - ▶ The simplest policy: FIFO (first in first out)
 - ▶ More complex policies (e.g. tagged 'hot lots' to be processed with higher priority)

- ▶ The simplest model to describe a supply chain

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- Discrete analog of a particle model in gas dynamics

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 - ▶ fluid models
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- ▶ The goal of this talk
 - ▶ 'Rigorously' derive a fluid model from a large particle limit of a DES model under FIFO policy
 - ▶ Refine this model into a kinetic model able to account for more complex policies (hot lots)

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- Review: [Daganzo]

2. A simple discrete event simulator

- Station S_m
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- $\tau(m, n) =$ time at which the part P_n enters the buffer queue of Station S_m
- Buffer queues are of infinite size (can be relaxed)

➤ First case: buffer of S_m non-empty

➤ S_m processes at full rate q_m

$$\implies \tau(m+1, n) = \tau(m+1, n-1) + \frac{1}{q_m}$$

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➤ Second case: buffer of S_m empty

➤ S_m processes part when it arrives

$$\implies \tau(m+1, n) = \tau(m, n) + T_m$$

▣▣▣▣▶ If buffer of S_m non-empty

$$\implies \tau(m + 1, n) \geq \tau(m, n) + T_m$$

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▣▶ If buffer of S_m empty

$$\implies \tau(m+1, n) \geq \tau(m+1, n-1) + \frac{1}{q_m}$$

▣▶ Collect the two cases into

$$\tau(m+1, n) = \max\left\{ \tau(m+1, n-1) + \frac{1}{q_m}, \tau(m, n) + T_m \right\}$$

- ▶ Investigate the 'thermodynamic limit'
 $M, N \rightarrow \infty$,
 - ▶ $M =$ Number of stations
 - ▶ $N =$ Number of parts
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 - ▶ $M =$ Number of stations
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- ▶ Idea: find that the DES is a discrete version of a conservation law in Lagrangian variable
 - ▶ $n =$ part index = 'mass' variable
 - ▶ $m =$ station index = 'space' variable

3. Continuity equation

- ➡ 'Position' (or Station number m) at which part P_n is at time t given by

$$\mu(t, n) = \frac{1}{M} \sum_{m=1}^M H(t - \tau(m, n))$$

where $H(x) = 1$ for $x > 0$ and 0 otherwise
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► 'Velocity' of part P_n

$$v(t, n) = \frac{d}{dt} \mu(t, n) = \frac{1}{M} \sum_{m=1}^M \delta(t - \tau(m, n))$$

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➡ Spacing between the parts (= 'specific volume')

$$\theta(t, n) = -\frac{\mu(t, n+1) - \mu(t, n)}{1/N}, \quad \theta(t, N) = 0$$

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►►► By construction

$$\frac{d}{dt} \theta(t, n) + \frac{v(t, n+1) - v(t, n)}{1/N} = 0$$

⇒ Define

⇒ $y = \text{mass variable} \in [0, 1]$ (part number)

⇒ $x = \text{space variable} \in [0, 1]$ (station number)

$$n = [Ny], \quad m = [Mx] \quad [\cdot] = \text{integer part}$$

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Assumption

$$\tau^{M,N}([Mx], [Ny]) \longrightarrow \tilde{\tau}(x, y)$$

as $M, N \rightarrow \infty$, as smoothly as we need

⇒ We have

$$\mu^{M,N}(t, [Ny]) \longrightarrow X(t, y)$$

where $t \rightarrow X(t, y)$ is the inverse function of $x \rightarrow \tilde{\tau}(x, y)$ (which is increasing)

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⇒ Proof: just a change of variable in the integral defining μ

⇒ $X(t, y)$ is the position in Lagrangian variables

⇒ We have

$$v^{M,N}(t, [Ny]) = \frac{d}{dt} \mu^{M,N}(t, [Ny]) \rightarrow \frac{dX}{dt}(t, y) := v(t, y)$$

(velocity in Lagrangian variables)

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(velocity in Lagrangian variables)

⇒ and

$$\theta^{M,N}(t, [Ny]) = \frac{\mu^{M,N}(t, [Ny] + 1) - \mu^{M,N}(t, [Ny])}{1/N}$$
$$\rightarrow -\frac{\partial X}{\partial y}(t, y) := \theta(t, y)$$

(specific volume)

⇒ Since

$$\frac{\partial}{\partial y} \left(\frac{dX}{dt} \right) = \frac{d}{dt} \left(\frac{\partial X}{\partial y} \right)$$

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⇒ The continuity equation is satisfied

$$\frac{\partial \theta}{\partial t} + \frac{\partial v}{\partial y} = 0$$

4. Constitutive relation

▣ was written

$$\tau([Mx] + 1, [Ny]) = \max \left\{ \tau([Mx] + 1, [Ny] - 1) + \frac{1}{q([Mx])}, \tau([Mx], [Ny]) + T([Mx]) \right\}$$

➡ After dividing by $1/N$ and some rearrangement

$$\max\left\{\frac{\tau([Mx] + 1, [Ny] - 1) - \tau([Mx] + 1, [Ny])}{1/N} + \frac{N}{q([Mx])}, \frac{N}{M} \left(\frac{\tau([Mx], [Ny]) - \tau([Mx] + 1, [Ny])}{1/M} + MT([Mx]) \right)\right\} = 0$$

⇒ $N \rightarrow \infty, q^{M,N} \rightarrow \infty$ and

$$\frac{N}{q^{M,N}([Mx])} \rightarrow \frac{1}{\bar{q}(x)}$$

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$$MT^{M,N}([Mx]) \rightarrow \bar{T}q(x)$$

⇒ $\frac{N}{M} \rightarrow 1$ Number of parts and stations are of the same order of magnitude

► Under scaling assumptions, as $N, M \rightarrow \infty$:

$$\max\left\{-\frac{\partial \tilde{\tau}}{\partial y} + \frac{1}{\tilde{q}}, -\frac{\partial \tilde{\tau}}{\partial x} + \tilde{T}\right\} = 0$$

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➡ Using that $t \rightarrow X(t, y)$ is the inverse fct of $x \rightarrow \tilde{\tau}(x, y)$

$$\frac{\partial \tilde{\tau}}{\partial x}(X(t, y), y) = \frac{1}{v(t, y)}, \quad \frac{\partial \tilde{\tau}}{\partial y}(X(t, y), y) = \frac{\theta(t, y)}{v(t, y)}$$

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Gives

$$\max\left\{-\frac{\theta(t, y)}{v(t, y)} + \frac{1}{\tilde{q}(X(t, y))}, -\frac{1}{v(t, y)} + \tilde{T}(X(t, y))\right\} = 0$$

➡ Either max is attained for 1st argument and

$$v(t, y) = q(X(t, y))\theta(t, y) \leq \frac{1}{T(X(t, y))}$$

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➡ or max is attained for 2nd argument and

$$v(t, y) = \frac{1}{T(X(t, y))} \leq q(X(t, y))\theta(t, y)$$

➡ Thus

$$v(t, y) = \min\left\{\frac{1}{T(X(t, y))}, q(X(t, y))\theta(t, y)\right\}$$

Supply chain model (in Lagrang. coord.) 27

$$\frac{\partial \theta}{\partial t} + \frac{\partial v}{\partial y} = 0$$

$$v(t, y) = \min\left\{\frac{1}{T(X(t, y))}, q(X(t, y))\theta(t, y)\right\}$$

$$-\frac{\partial X}{\partial y}(t, y) = \theta(t, y)$$

$$\frac{\partial \theta}{\partial t} + \frac{\partial v}{\partial y} = 0$$

$$v(t, y) = \min\left\{\frac{1}{T(X(t, y))}, q(X(t, y))\theta(t, y)\right\}$$

$$-\frac{\partial X}{\partial y}(t, y) = \theta(t, y)$$

▣► Last eq. equivalent to:

$$X(t, y) = \int_y^1 \theta(t, z) dz$$

5. Passage to Eulerian variables

➡ Obtain a model in (t, x) rather than (t, y)

- ▶▶▶ Obtain a model in (t, x) rather than (t, y)
- ▶▶▶ Classical procedure in gas dynamics
 - ▶▶ Coordinate change:

$$x = X(t, y) = \int_y^1 \theta(t, z) dz \quad \text{strictly } \searrow \text{ of } y$$

$$y = Y(t, x) \quad \text{inverse fct}$$

Using that $y \rightarrow Y(t, x)$ is the inverse fct of $x \rightarrow X(t, y)$

$$\frac{\partial Y}{\partial x}(t, x) = -\frac{1}{\theta(t, Y(t, x))} := \rho(t, x)$$

Number density of parts at x at time t

- ⇒ Using that $y \rightarrow Y(t, x)$ is the inverse fct of $x \rightarrow X(t, y)$

$$\frac{\partial Y}{\partial x}(t, x) = -\frac{1}{\theta(t, Y(t, x))} := \rho(t, x)$$

Number density of parts at x at time t

- ⇒ and

$$\frac{\partial Y}{\partial t}(t, x) = -\frac{v(t, Y(t, x))}{\theta(t, Y(t, x))} := \rho(t, x)u(t, x)$$

$u(t, x) = v(t, Y(t, x))$ velocity in Eulerian coord.

⇒ Since

$$\frac{\partial}{\partial t} \left(\frac{\partial Y}{\partial x} \right) = \frac{\partial}{\partial x} \left(\frac{\partial Y}{\partial t} \right)$$

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Since

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The continuity equation is satisfied

$$\frac{\partial \rho}{\partial t} + \frac{\partial \rho u}{\partial x} = 0$$

From the constitutive relation in Lagrangian variable, we get

$$\rho u(t, x) = \min \left\{ \frac{1}{T(x)} \rho(t, x), q(x) \right\}$$

$$\frac{\partial \rho}{\partial t} + \frac{\partial \rho u}{\partial x} = 0$$

$$\rho u(t, x) = \min\left\{\frac{1}{T(x)}\rho(t, x), q(x)\right\}$$

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$$\rho u(t, x) = \min\left\{\frac{1}{T(x)}\rho(t, x), q(x)\right\}$$

➡ Hyperbolic model with flux constraint

6. Kinetic model

Fluid model

$$\frac{\partial \rho}{\partial t} + \frac{\partial \rho u}{\partial x} = 0$$

$$\rho u = \min\{\rho V_0, q\}, \quad V_0 = \frac{1}{T}$$

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Particle interpretation

$$\dot{X} = \begin{cases} V_0 \\ 0 \end{cases} \quad \dot{\rho} = \begin{cases} -\rho \partial_x V_0 & \text{if } \rho V_0 \leq q \\ -\partial_x q & \text{if } \rho V_0 > q \end{cases}$$

- Either particle moves or is blocked according to whether the 'free' flux ρV_0 is below or exceeds the threshold q .

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- ▶▶▶▶ Either particle moves or is blocked according to whether the 'free' flux ρV_0 is below or exceeds the threshold q .
- ▶▶▶▶ Kinetic model: 'regularization' of this singular dynamics
- ▶▶▶▶ Introduce an attribute variable ξ to each particle
 - ▶▶▶▶▶ Particles move with actual velocity

$$V(t, x, \xi) \leq V_0$$

- ▶▶▶▶▶ s.t. total flux $\leq q$

⇒ $f(x, \xi, t)$ density of parts at time t , position x ,
with attribute ξ

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⇒ Density ρ and flux ρu

$$\rho = \int f(x, \xi, t) d\xi, \quad \rho u = \int f(x, \xi, t) V(x, \xi, t) d\xi$$

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⇒ Density ρ and flux ρu

$$\rho = \int f(x, \xi, t) d\xi, \quad \rho u = \int f(x, \xi, t) V(x, \xi, t) d\xi$$

⇒ Maximal possible flux Q :

$$Q = \int f(x, \xi, t) V_0(x) d\xi = \rho V_0$$

Flux if there would be no capacity limitation

- Higher priority to parts with lower attribute values
 - Note: same methodology would apply for other policies

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- Procedure: move parts by increasing attribute number with maximum allowed speed V_0 until processor capacity is reached

- Higher priority to parts with lower attribute values
 - Note: same methodology would apply for other policies
- Procedure: move parts by increasing attribute number with maximum allowed speed V_0 until processor capacity is reached
- Number of parts with attribute $\leq \alpha$ is

$$\int_{-\infty}^{\alpha} f(x, \xi, t) d\xi$$

- ➡ If parts with attribute $\leq \alpha$ all move with maximal speed V_0 , the associated flux is

$$\beta(x, \alpha, t) = V_0(x) \int_{-\infty}^{\alpha} f(x, \xi, t) d\xi$$

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- ➡ The fonction

$$\alpha \in \mathbb{R} \rightarrow \beta(x, \alpha, t) \in [0, Q]$$

is increasing

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- ▶▶▶ Denote $\beta^{-1}(x, \cdot, t)$ its inverse

- Processors process all parts with attribute $\leq \alpha$
- where α s.t. associated flux = processor capacity

$$\beta(x, \alpha, t) = q \quad (\Leftrightarrow \alpha = \beta^{-1}(x, q, t))$$

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 - in which case $\alpha = \infty$

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- Then

$$\beta(x, \alpha, t) = \min\{q, Q\}$$

- ▶ Note: processor velocity V_0 can be attribute dependent

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$$V_0 = V_0(x, \xi)$$

- Actual velocity

$$V(x, \xi, t) = \begin{cases} V_0(x, \xi) & \text{if } \xi \leq \alpha(x, t) \\ 0 & \text{if } \xi > \alpha(x, t) \end{cases}$$

- ➡ Note: processor velocity V_0 can be attribute dependent

$$V_0 = V_0(x, \xi)$$

- ➡ Actual velocity

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- ➡ or

$$V(x, \xi, t) = V_0(x, \xi)H(\alpha(x, t) - \xi) \quad H = \text{Heaviside}$$

▣ Since β ↗ fonction of α , we have

$$\begin{aligned} H(\alpha(x, t) - \xi) &= H(\beta(x, \alpha(x, t), t) - \beta(x, \xi, t)) \\ &= H(\min\{q, Q\} - \beta(x, \xi, t)) \\ &= H(q - \beta(x, \xi, t)) \quad (\text{since } \beta \leq Q) \end{aligned}$$

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➡ and

$$\begin{aligned} \beta(x, \xi, t) &= \int_{-\infty}^{\xi} V_0(x, \xi') f(x, \xi', t) d\xi' \\ &= \int_{\mathbb{R}} V_0(x, \xi') f(x, \xi', t) H(\xi - \xi') d\xi' \end{aligned}$$

➡ Finally

$$V(x, \xi, t) = V_0(x, \xi) H(q - \beta(x, \xi, t))$$

$$\beta(x, \xi, t) = \int_{\mathbb{R}} V_0(x, \xi') f(x, \xi', t) H(\xi - \xi') d\xi'$$

► By analogy with fluid model

$$\dot{X} = V(X, \Xi, t) = V_0(X, \Xi)H(q - \beta(X, \Xi, t))$$

$$\dot{f} = -f(\partial_x V)|_{(X, \Xi, t)}$$

$$\dot{\Xi} = 0$$

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$$\dot{X} = V(X, \Xi, t) = V_0(X, \Xi)H(q - \beta(X, \Xi, t))$$

$$\dot{f} = -f(\partial_x V)|_{(X, \Xi, t)}$$

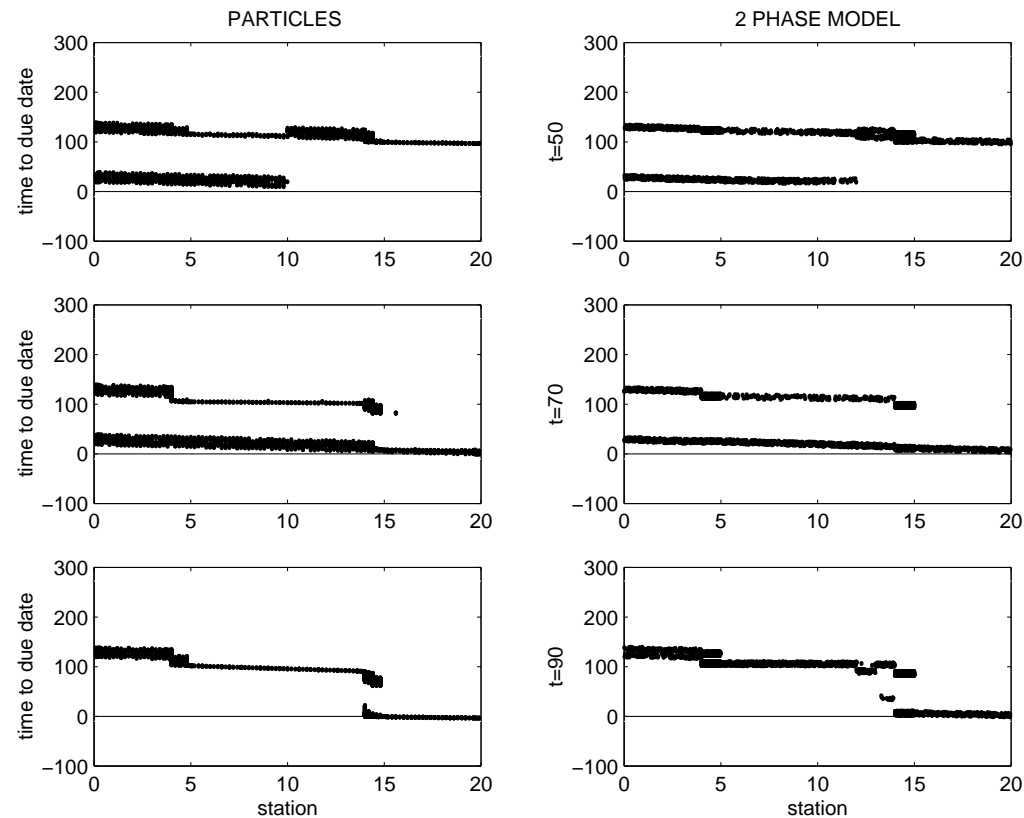
$$\dot{\Xi} = 0$$

► Characteristics of the first order kinetic eq.

$$\partial_t f + \partial_x (V f) = 0$$



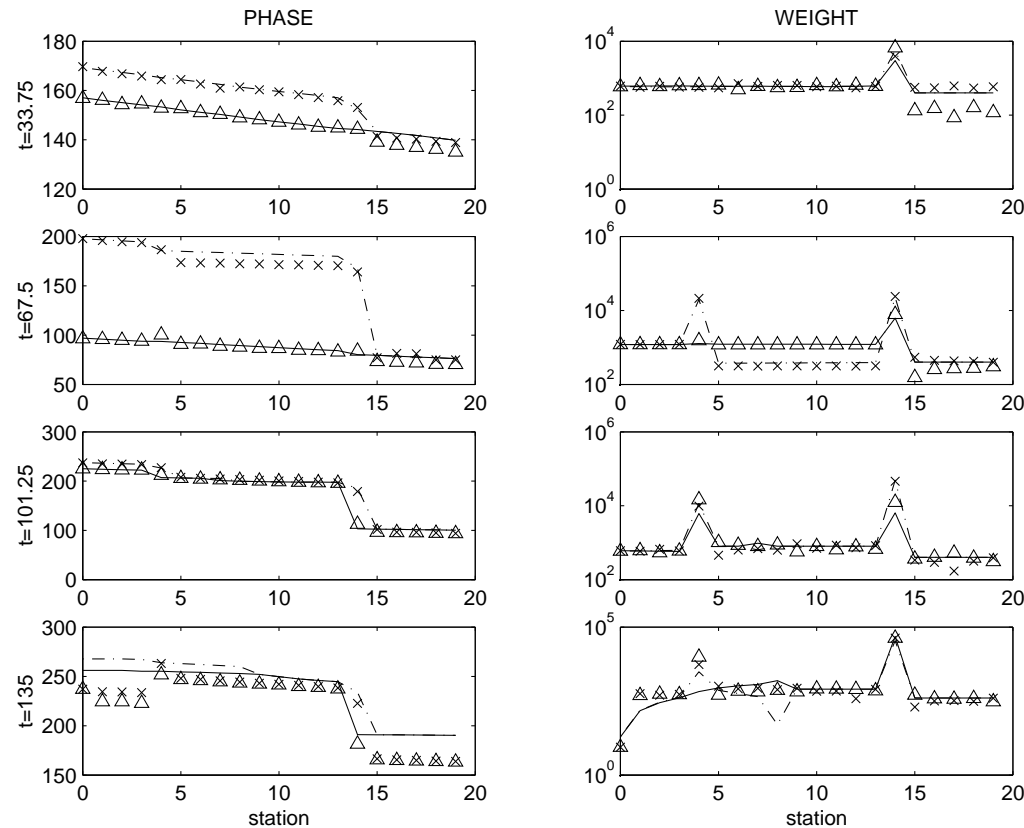
Case 2: Multiphase and kinetic models 61



Phases as a function of 'space' (or DOC)
at various times (top to down)

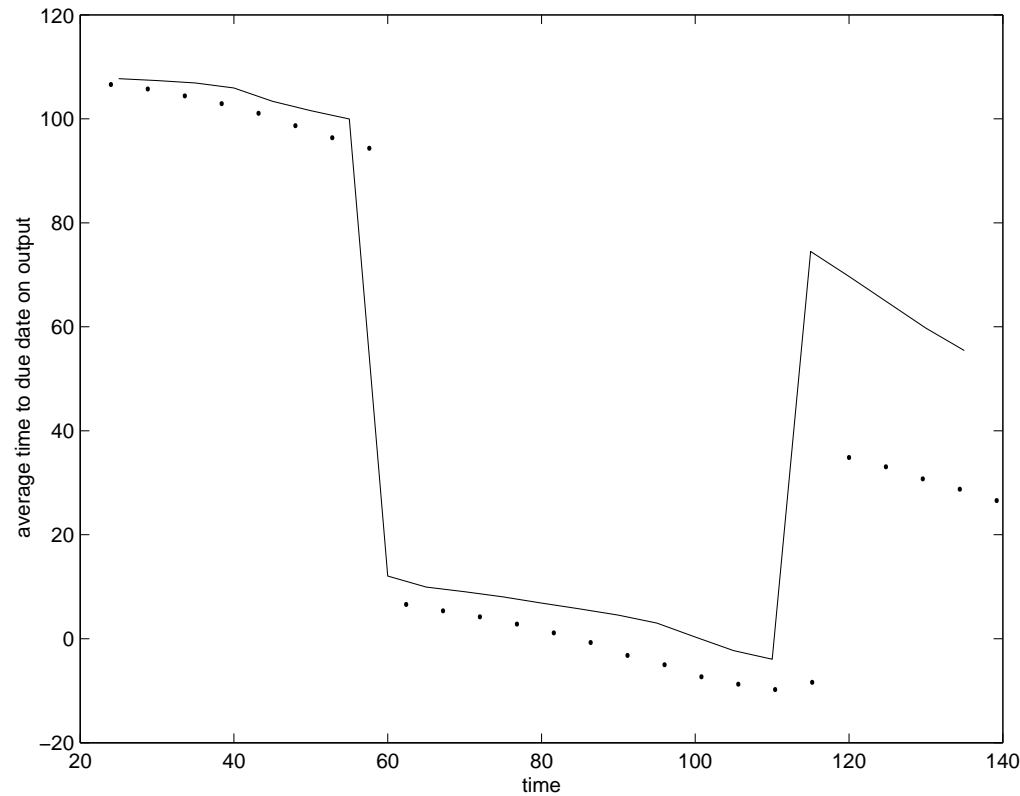
Left: kinetic model Right: two-phase model

Case 2: Multiphase and kinetic models 62



Phases (left) and densities (right) as a function of 'space'
at various times (top to down)

Kinetic (\times , Δ) and two-phase (—, - · -) models



Expected time to due-date m_1/m_0 in the last cell
as a function of time
Kinetic (· · ·) and two-phase (—) models

9. Conclusion

- (Rigorous) derivation of continuum model from Discrete Event Simulator
 - Nonlinear hyperbolic model with saturated flux
 - Reproduces DES model satisfactorily even in situations where capacity has large variations

- (Rigorous) derivation of continuum model from Discrete Event Simulator
 - Nonlinear hyperbolic model with saturated flux
 - Reproduces DES model satisfactorily even in situations where capacity has large variations
- Kinetic model with an internal variable (policy attribute)
 - Simple closure recovers the fluid model
 - Multiphase closure allows to implement policies
 - Correct agreement between two-phase and kinetic models

⇒ Fluid model

⇒ Networks

- ▣ Fluid model
 - ▣ Networks

- ▣ Kinetic model
 - ▣ Randomness (random failures)

- ▶ Fluid model
 - ▶ Networks

- ▶ Kinetic model
 - ▶ Randomness (random failures)

- ▶ More complex models
 - ▶ Orders, payments, etc.