

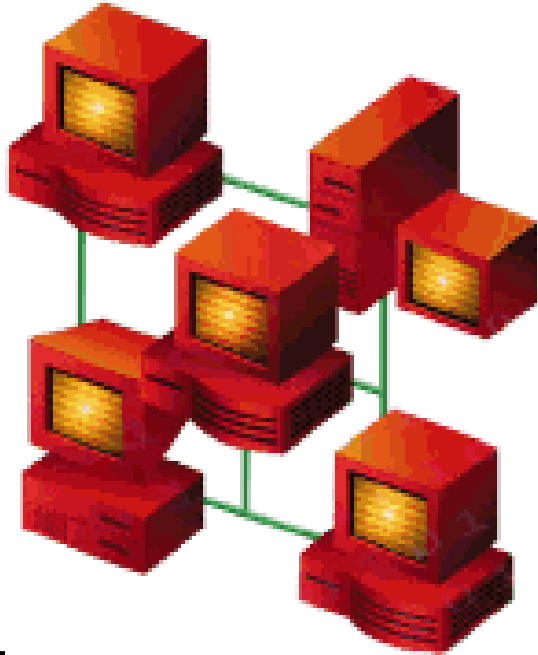
MEASUREMENT-BASED HYBRID FLUID-FLOW MODELS FOR FAST MULTI-SCALE SIMULATION

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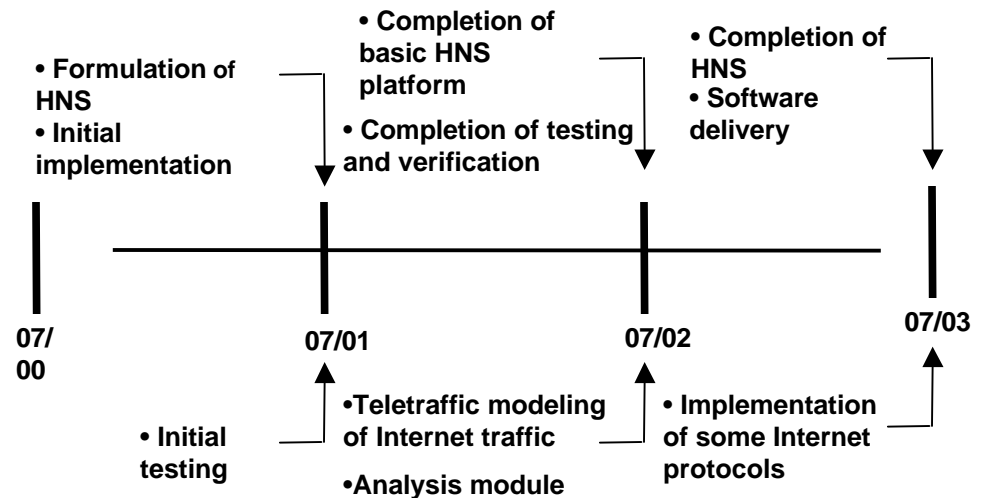
New Ideas

- *Novel measurement-based traffic modeling methodology based on general time-Series processes (e.g., Auto-regressive Modular)*
- *New hybrid discrete-continuous flow (HDFC) paradigm to combine discrete and continuous flows resulting in fast simulation and considerable modeling flexibility*

Impact

- **Accurate traffic modeling driven by measurement-based models**
- **Multi-scale simulation paradigm from packet transport to protocol-based messages**
- **Identification of generic scalable network topologies**

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MOTIVATION

- **Emerging high-speed packet-based telecommunications networks carry enormous traffic loads**
 - compressed video
 - file transfer
- **Network modeling and analysis technologies are urgently needed (*witness Internet congestion*)**
 - network control (admission and congestion)
 - network provisioning and planning

PROJECT GOALS

- **PROBLEM: Emerging high-speed packet-based telecommunications networks are hard to analyze**
 - current analytical models cannot capture teletraffic burstiness and are overly optimistic
 - simulation of complex networks is either infeasible, or takes forever to complete
- **SOLUTION GOALS: Develop a new modeling and simulation paradigm**
 - hybrid simulation paradigm that combines traditional discrete flows with continuous ones
 - multi-scale simulation paradigm from packet transport to protocol-based messages
 - accurate teletraffic modeling driven by measurement-based teletraffic models

TECHNICAL CHALLENGES

- **How to achieve a high expressive power of simulation models by capturing multiple scales?**
 - transaction level (discrete and continuous flows)
 - message level (subject to prescribed protocols)
- **How to speed up simulation runs?**
 - large and complex network models give rise to enormous numbers of packet-based events
 - traditional simulation would require prohibitive computational resources to process
- **How to achieve a high accuracy of predicted performance measures?**
 - burstiness modeling
 - measurement-based teletraffic modeling

NEW IDEAS

- **New Hybrid Discrete-Continuous Flow (HDCF) paradigm combines discrete and continuous flows**
 - fast simulation takes advantage of fluid transport
 - flexible modeling allows modeler to assign type of flows (traditional discrete jobs or fluid-flow streams)
- **Traffic model**
 - new accurate measurement-based teletraffic modeling methods via ARM (AutoRegressive Modular processes), e.g., TES, QTES

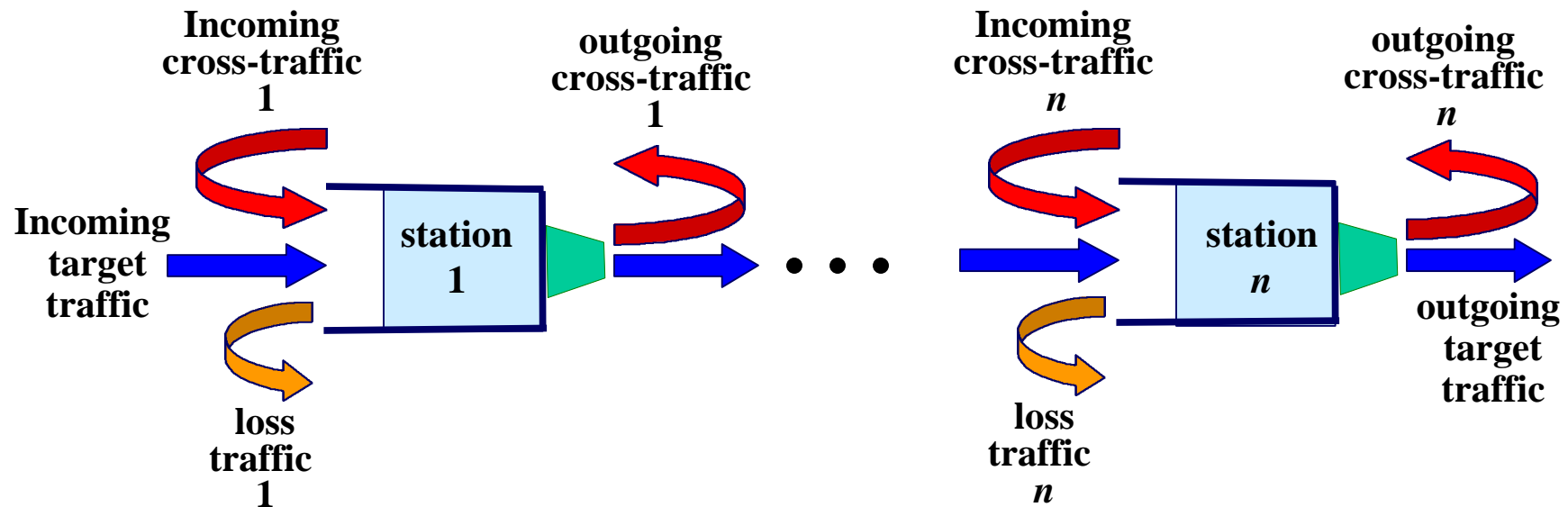
BENEFITS TO DOD

- **Ability to model and simulate complex networks**
 - accurate simulation models whose computational complexity using traditional models is currently infeasible or prohibitive
 - flexible paradigm allows users to control trade-off between computational complexity and model fidelity
- **Integration with network applications**
 - modeling of Next Generation Internet at multiple levels of detail
 - network design and capacity planning
 - network control (buffer management, service allocation)

FLUID-FLOW MODEL SOLUTION

- **Problem:** traditional discrete-event simulation at packet level is often *infeasible*
 - time complexity (too many packets to process)
 - space complexity (too many packets to store)
- **Proposed Solution:** *fluid-flow models*
 - transactions (customers, packets) become fluid
 - random discrete arrivals become random arrival rates
 - random discrete services become random service rates
 - random routing becomes rate thinning and merging
 - sample paths governed by differential equations
 - complex networks are modeled as TT/CT (Target-Traffic /Cross-Traffic) networks

THE GENERIC TT/CT NETWORK MODEL

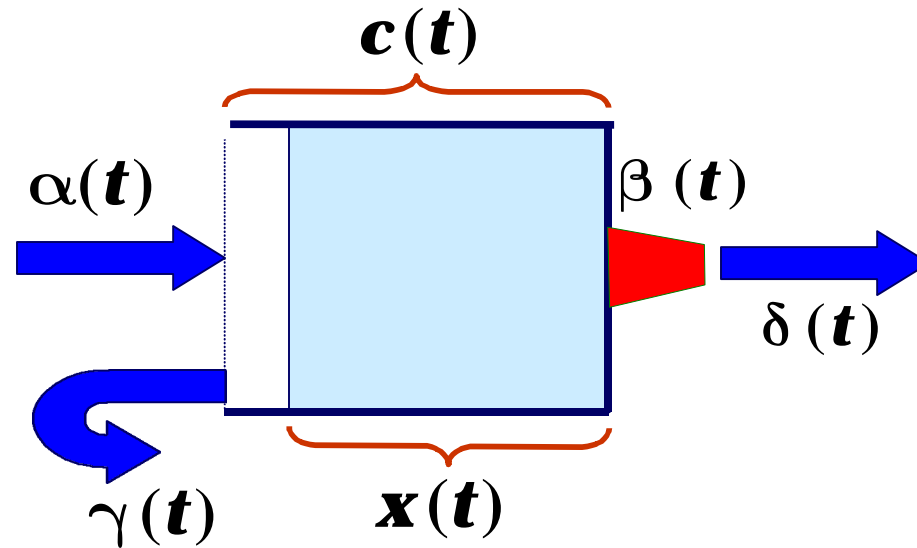


- **The generic TT/CT (Target-Traffic / Cross-Traffic) network model is a useful class of networks**
 - simple HDCF network with tandem topology
 - reduced complexity renders simulation scalable in path size n
 - accurate measurement- based teletraffic modeling and generation methods (e.g., QTES) already developed under a previous DARPA/ITO project

FLUID-FLOW MODELING IMPLICATIONS

- **Fluid-flow simulation implications**
 - events correspond to rate changes, which occur far less frequently than packet arrivals, service and routing
 - rate changes affect all downstream flows in a fluid-flow network, so fluid-flow events are more expensive than packet-flow events
 - overall, a fluid-flow simulation **usually** runs much faster than its packet-flow counterpart

BASIC CONTINUOUS-FLOW MODEL (CFM)



Defining Processes

$\alpha(t)$ = inflow rate at time t

$\beta(t)$ = service rate at time t

$c(t)$ = capacity rate at time t

Derived Processes

$x(t)$ = workload at time t

$\gamma(t)$ = loss rate at time t

$\delta(t)$ = outflow rate at time t

DEFINING PROCESSES ASSUMPTIONS

- **The time horizon is an interval $[0, T)$**
- **The inflow rate process $\{\alpha(t)\}_{t=0}^T$ satisfies**
 - with probability 1, the sample paths $\alpha(x)$ are piecewise continuous and continuously differentiable in their continuity intervals
- **The service rate process $\{\beta(t)\}_{t=0}^T$ satisfies**
 - with probability 1, the sample paths $\beta(x)$ are piecewise continuous and continuously differentiable in their continuity intervals
- **The buffer capacity rate process $\{c(t)\}_{t=0}^T$ satisfies**
 - with probability 1, the sample paths $c(x)$ are piecewise continuous and continuously differentiable in their continuity intervals

CFM AND DISCRETE-EVENT SIMULATION

- **Suppose that with probability 1, all defining processes of the basic CFM satisfy**
 - all defining sample paths are piecewise constant
 - the number of jumps in finite time intervals is finite
- **Then**
 - the CFM is a DEDS (Discrete-Event Dynamic System)
 - the CFM can be simulated by a discrete-event simulation
 - the superposition of all jump time points of all defining processes over any finite interval can be written (with probability 1) as a strictly increasing finite sequence

$$\{[t_i, t_{i+1})\}_{i=0}^N$$

- in particular, for any initial interval,

$$t_0 = 0, \quad t_{N+1} = T$$

WORKFLOW PROCESS

- The workload process $\{\mathbf{x}(t)\}_{t=0}^T$ is governed by

$$\frac{d}{dt}\mathbf{x}(t) = \begin{cases} \mathbf{0}, & \text{if } \mathbf{x}(t) = \mathbf{0} \text{ and } \alpha(t) \leq \beta(t) \\ \mathbf{c}(t), & \text{if } \mathbf{x}(t) = \mathbf{c}(t) \text{ and } \alpha(t) - \beta(t) \geq \mathbf{c}(t) \\ \alpha(t) - \beta(t), & \text{otherwise} \end{cases}$$

- Let all defining sample paths be piecewise constant, with finite number of jumps in finite time intervals
 - then the workload process is piecewise linear, and its values at **event times** can be computed recursively by

$$\mathbf{x}_{i+1} = \min\{\max\{\mathbf{x}(t_i) + [\alpha(t_i) - \beta(t_i)][t_{i+1} - t_i], \mathbf{0}\}, \mathbf{c}(t_{i+1})\}$$

for a given initial value $\mathbf{x}(0) = \mathbf{x}_0$

OUTFLOW RATE PROCESS

- The outflow rate process $\{\delta(\mathbf{t})\}_{t=0}^T$ is defined by

$$\delta(\mathbf{t}) = \begin{cases} \alpha(\mathbf{t}), & \text{if } \mathbf{x}(\mathbf{t}) = \mathbf{0} \\ \beta(\mathbf{t}), & \text{if } \mathbf{x}(\mathbf{t}) > \mathbf{0} \end{cases}$$

- if the defining sample paths are piecewise constant, then the loss rate process is piecewise constant, and can be computed from the workload process
- The average outflow (throughput) over $[0, T)$ is

$$\bar{\delta}(T) = \frac{1}{T} \int_0^T \delta(\mathbf{t}) dt$$

LOSS RATE PROCESSES

- The loss rate process $\{\gamma(\mathbf{t})\}_{\mathbf{t}=0}^T$ is defined by

$$\gamma(\mathbf{t}) = \begin{cases} \alpha(\mathbf{t}) - \beta(\mathbf{t}) - \mathbf{c}(\mathbf{t}), & \text{if } \mathbf{x}(\mathbf{t}) = \mathbf{c}(\mathbf{t}) \text{ and} \\ \alpha(\mathbf{t}) \text{ } \beta(\mathbf{t}) + \mathbf{c}(\mathbf{t}) & \\ \mathbf{0}, & \text{otherwise} \end{cases}$$

- if the defining sample paths are piecewise constant, then the loss rate process is piecewise constant, and can be computed from the workload process

LOSS VOLUME PROCESSES

- The loss volume $L(t_1, t_2)$ over $[t_1, t_2)$ is defined by

$$L(t_1, t_2) = \int_{t_1}^{t_2} \gamma(t) dt$$

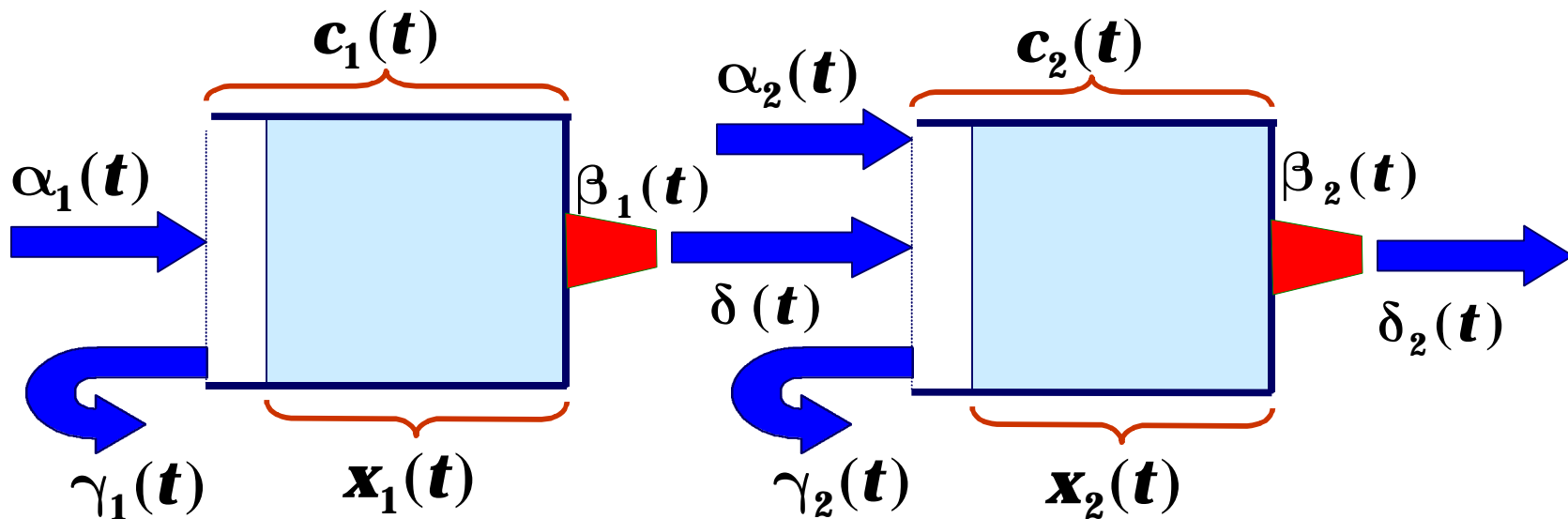
- Let all defining sample paths be piecewise constant, with finite number of jumps in finite time intervals
 - then the partial loss volumes over $[t_1, t_2)$ are given by

$$L(t_i, t_{i+1}) = \begin{cases} [\alpha(t_i) - \beta(t_i)][t_{i+1} - t_i] + \mathbf{x}(t_i) - \mathbf{c}(t_{i+1}), & \text{if } \mathbf{x}(t_i) = \mathbf{c}(t_{i+1}) \\ \mathbf{0}, & \text{otherwise} \end{cases}$$

- The loss fraction over $[t_1, t_2)$ is defined by

$$L_f(t_1, t_2) = \int_{t_1}^{t_2} \gamma(t) dt / \int_{t_1}^{t_2} \alpha(t) dt$$

CFM NETWORKS



Defining Processes

$\alpha_i(t)$ = inflow rate at node i

$\beta_i(t)$ = service rate at node i

$c_i(t)$ = capacity rate at node i

Derived Processes

$x_i(t)$ = workload at node i

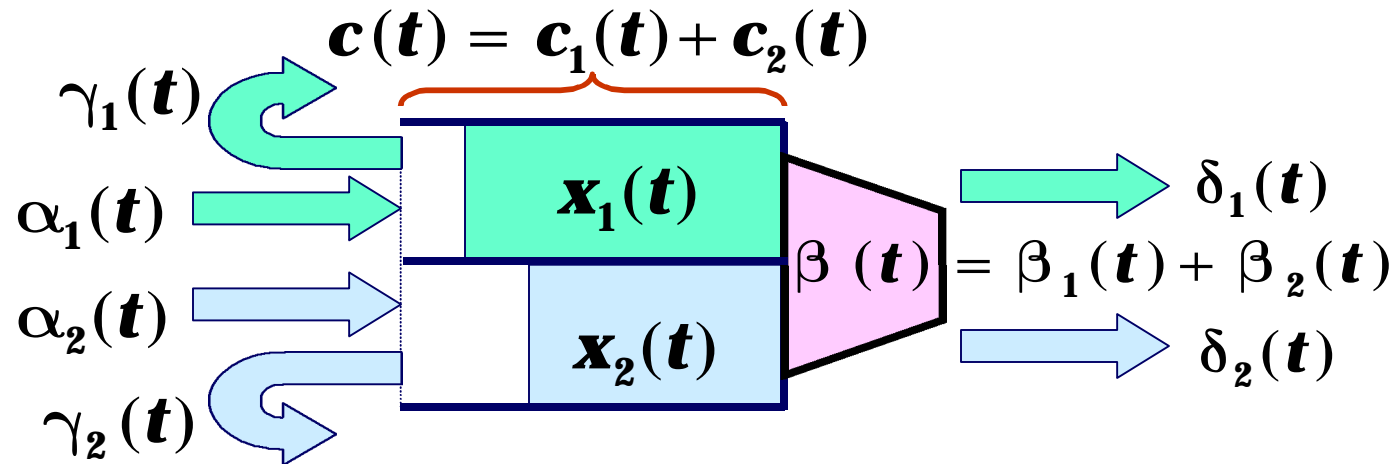
$\gamma_i(t)$ = loss rate at node i

$\delta_i(t)$ = outflow rate at node i

CFM NETWORKS (Cont.)

- A **CFM network** is a set of *interacting* basic CFM nodes
 - shared buffer
 - shared server
- **Basic CFM nodes may be interconnected**
 - flows have an itinerary of multiple nodes
 - flows may split and merge
- **For piecewise constant defining processes**
 - CFM network is a DEDS, with all derived processes being piecewise constant
 - CFM network is amenable to discrete-event simulation

MULTIPLE-FLOW CFM's



Defining Processes

$\alpha_i(t) = i$ - **th** inflow rate

$\beta_i(t) = i$ - **th** service rate

$\beta(t) =$ total service rate

$c_i(t) = i$ - **th** capacity rate

$c(t) =$ total buffer capacity

Derived Processes

$x_i(t) = i$ - **th** workload

$\gamma_i(t) = i$ - **th** loss rate

$\delta_i(t) = i$ - **th** outflow rate

INTERACTING EQUAL-PRIORITY FLOWS

- **Basic equal-priority CFM₁ and CFM₂**
- **Defining processes**
 - inflow rates are $\alpha_i(\mathbf{t}), i=1,2$
 - service rates are $\beta_i(\mathbf{t}), i = 1,2$, subject to a shared total service rate $\beta(\mathbf{t}) = \beta_1(\mathbf{t}) + \beta_2(\mathbf{t})$
 - buffer capacities are $\mathbf{c}_i(\mathbf{t}), i=1,2$, subject to a shared total buffer capacity $\mathbf{c}(\mathbf{t}) = \mathbf{c}_1(\mathbf{t}) + \mathbf{c}_2(\mathbf{t})$
- **Derived processes**
 - workloads $\mathbf{x}_i(\mathbf{t}), i=1,2$, computed separately
 - loss rates $\gamma_i(\mathbf{t}), i=1,2$, computed separately
 - outflow rates $\delta_i(\mathbf{t}), i=1,2$, computed separately

INTERACTING PREEMPTIVE-PRIORITY FLOWS

- Basic CFM₁ of higher priority than basic CFM₂
 - Defining processes
 - inflow rates are $\alpha_i(\mathbf{t})$, $i=1,2$
 - service rates are $\beta_i(\mathbf{t})$, $i=1,2$, subject to a shared total service rate $\beta(\mathbf{t}) = \beta_1(\mathbf{t}) + \beta_2(\mathbf{t})$ such that

$$\beta_1(\mathbf{t}) = \begin{cases} \beta(\mathbf{t}), & \text{if } \mathbf{x}_1(\mathbf{t}) > \mathbf{0} \\ \alpha_1(\mathbf{t}), & \text{if } \mathbf{x}_1(\mathbf{t}) = \mathbf{0} \end{cases} \quad \beta_2(\mathbf{t}) = \beta(\mathbf{t}) - \beta_1(\mathbf{t})$$
 - buffer capacities are $\mathbf{c}_i(\mathbf{t})$, $i=1,2$, subject to a shared total buffer capacity $\mathbf{c}(\mathbf{t}) = \mathbf{c}_1(\mathbf{t}) + \mathbf{c}_2(\mathbf{t})$ such that

$$\mathbf{c}_1(\mathbf{t}) = \begin{cases} \mathbf{c}(\mathbf{t}), & \text{if } \mathbf{x}_1(\mathbf{t}) > \mathbf{0} \\ \mathbf{0}, & \text{if } \mathbf{x}_1(\mathbf{t}) = \mathbf{0} \end{cases} \quad \mathbf{c}_2(\mathbf{t}) = \mathbf{c}(\mathbf{t}) - \mathbf{c}_1(\mathbf{t})$$
 - Derived processes are computed separately
-

NETWORK SERVICE RATE ALLOCATION

- CFM simulation requires a service rate allocation algorithm, invoked on state changes
- **Input**
 - network nodes $\{1, \dots, n\}$
 - current inflow rate at each node $\{\alpha_j : j=1, \dots, n\}$
 - current workload at each node $\{x_j : j=1, \dots, n\}$
 - network total service rate β to be allocated to nodes
- **Output**
 - service rates at each node $\beta_j, j=1, \dots, n$, such that

$$\sum_{j=1}^n \beta_j = \beta$$

SERVICE RATE ALLOCATION ALGORITHM

- **Initialize**

set $\mathbf{N} \leftarrow \{1, \dots, n\}$

set $\mathbf{Z} \leftarrow \{j \in \mathbf{N} : \mathbf{x}_j = \mathbf{0} \text{ and } \alpha_j < \beta/n\}$

set $\mathbf{b} \leftarrow \beta$

- **Main loop**

while ($\mathbf{Z} \neq \emptyset$)

set $\beta_j \leftarrow \alpha_j$ for all $j \in \mathbf{Z}$

set $\mathbf{b} \leftarrow \mathbf{b} - \sum_{j \in \mathbf{Z}} \alpha_j$

set $\mathbf{N} \leftarrow \mathbf{N} - \mathbf{Z}$

set $\mathbf{Z} \leftarrow \{j \in \mathbf{N} : \mathbf{x}_j = \mathbf{0} \text{ and } \alpha_j < \beta/|\mathbf{N}|\}$

- **Finalize**

if ($\mathbf{N} \neq \emptyset$)

set $\beta_j \leftarrow \mathbf{b}/|\mathbf{N}|$ for all $j \in \mathbf{N}$

FLUID VS. PACKET TRANSPORT

- **Main simulation events in packet-based transport**
 - arrivals, service completions, packet loss
- **Main simulation events in CFM**
 - changes in arrival rate, service rate and capacity rate
- **Comparison of Computational Complexity**
 - packet-based transport has enormous number of events, each being local
 - CFM transport has far fewer events in single-node models
 - events in CFM transport are global (rate re-computation)
 - in feed-forward CFM networks, rate re-computation is fast, but events grow quadratically via the *ripple effect*
 - in general CFM networks, rate re-computation is hard, and events can grow explosively via the *echo effect*

PROPOSED CFM RESEARCH

- **CFM telecommunications applications**
 - network design, planning and provisioning
 - network resource allocation

- **CFM software tools**
 - object-oriented CFM simulator architecture and software