

Some Developments in the Tagged Signal Model

Xiaojun Liu
With J. Adam Cataldo, Edward A. Lee,
Eleftherios D. Matsikoudis, and Haiyang Zheng



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The Tagged Signal Model

- A set of *tags* T , e.g. $T = [0, \infty)$
- A set of *values* V , e.g. $V = \mathbf{N}$
- An *event* e is a pair of a tag and a value:
 $e = (t, v)$
- A *signal* s is a set of events, e.g.
 $clock_1 = \{(0.0, 1), (1.0, 1), (2.0, 1), \dots\}$
- A *process* P is a relation on signals

$$\begin{array}{c} s_1 \\ s_2 \end{array} \boxed{P} s_3 \quad P = \{ (s_1, s_2, s_3) \mid s_1 + s_2 - s_3 = 0 \}$$



Signals and Processes

	Signals	Processes
Physics	Velocities, Accelerations, and Forces	Newton's Laws
Electrical Engineering	Voltages and Currents	Resistors and Capacitors, Kirchhoff's Laws
Computer Science	Streams	Dataflow Processes

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Approach

- Study the mathematical structure of signal sets
 - Partial order/CPO, topological/metric space, algebra
- Study the properties of processes as relations/functions on signals
 - Continuity
 - Causality
 - Composition
- From the declarative to the imperative

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Signals

- Let T , a poset, be the set of all tags. Let $\mathcal{D}(T)$ be the set of down-sets of T .



- A signal is a function from a down-set $D \in \mathcal{D}(T)$ to some value set V ,

$$\text{signal}: D \rightarrow V$$

- Let $\mathcal{S}(T, V)$ be the set of all signals from down-sets of T to V .

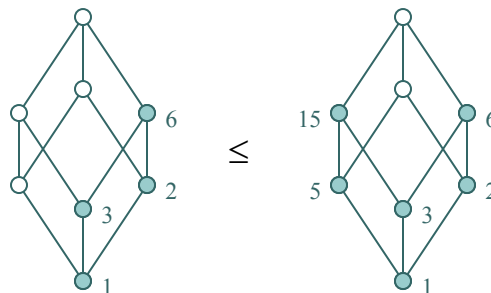
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Prefix Order on Signals

- A signal $s_1: D_1 \rightarrow V$ is a prefix of $s_2: D_2 \rightarrow V$, denoted $s_1 \leq s_2$, if and only if

$$D_1 \subseteq D_2, \text{ and } s_1(t) = s_2(t), \forall t \in D_1$$



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Prefix Order – Properties

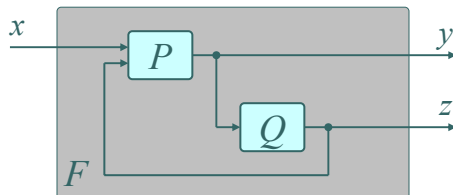
- For any poset T of tags and set V of values, $\mathcal{S}(T, V)$ with the prefix order is
 - a poset
 - a CPO
 - a complete lower semilattice (i.e. any subset of signals have a “longest” common prefix)

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Tagged Process Networks

- A direct generalization of Kahn process networks



$$(y, z) = F(x)$$

where (y, z) is the least solution of the equations

$$y = P(x, z)$$
$$z = Q(y)$$

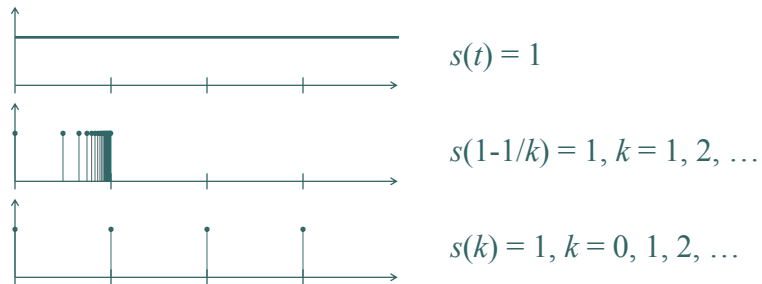
- If processes P and Q are Scott-continuous, then F is Scott-continuous.

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Timed Signals

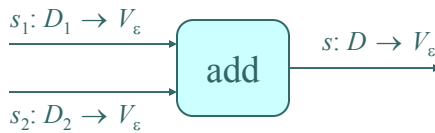
- Let $T = [0, \infty)$, and $V_\varepsilon = V \cup \{\varepsilon\}$, where ε represents the absence of value, $\mathcal{S}(T, V_\varepsilon)$ is the set of timed signals.



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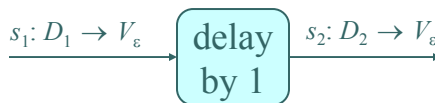


Timed Processes



$$D = D_1 \cap D_2$$

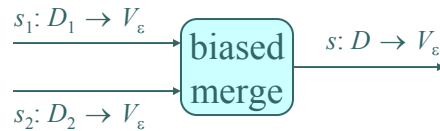
$$s(t) = s_1(t) +_\varepsilon s_2(t)$$



$$D_2 = D_1 \oplus \{1\} \cup [0, 1)$$

$$s_2(t) = s_1(t-1), \text{ when } t \geq 1$$

$$\varepsilon, \text{ when } t \in [0, 1)$$



$$D = D_1 \cap D_2$$

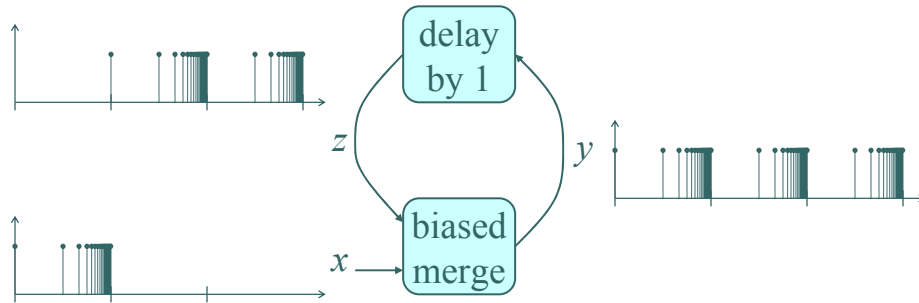
$$s(t) = s_1(t), \text{ when } s_1(t) \in V$$

$$s_2(t), \text{ otherwise}$$

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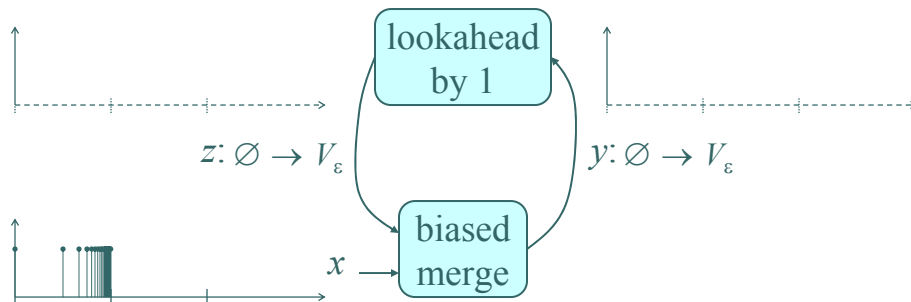
A Timed Process Network



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A Non-Causal Process in the Network



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Causality

- A timed process P is causal if
 - It is monotonic, i.e. for all s_1, s_2
$$s_1 \leq s_2 \Rightarrow P(s_1) \leq P(s_2)$$
 - For all $s: D_1 \rightarrow V_1, P(s): D_2 \rightarrow V_2$
$$D_1 \subseteq D_2$$
- A timed process P is strictly causal if it is monotonic, and
 - For all $s: D_1 \rightarrow V_1, P(s): D_2 \rightarrow V_2$
$$D_1 \subset D_2 \text{ or } D_2 = [0, \infty)$$

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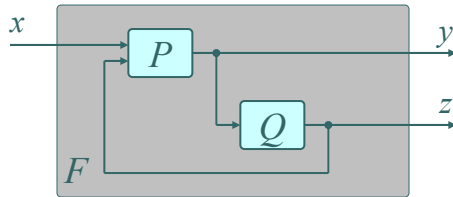
Causality and Continuity

- Neither implies the other.
- A process may be continuous but not causal, e.g. “lookahead by 1”.
- A process may be causal but not continuous, e.g. one that produces an output event after counting an infinite number of input events.

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Causal Timed Process Networks



$$(y, z) = F(x)$$

where (y, z) is the least solution of the equations

$$y = P(x, z)$$

$$z = Q(y)$$

- If processes P and Q are causal and continuous, and at least one of them is strictly causal, then F is causal and continuous.

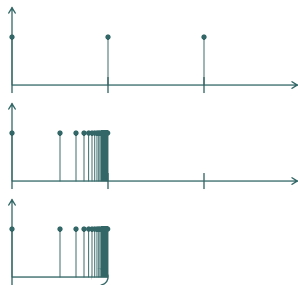
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Discrete Event Signals

- A timed signal $s: D \rightarrow V_\varepsilon$ is a discrete event signal if for all $t \in D$

$$s^{-1}(V) \cap [0, t] \text{ is a finite set}$$



$$\text{dom}(s) = [0, \infty) \quad \text{DE, Non-Zeno}$$

$$s(k) = 1, k = 0, 1, 2, \dots$$

$$\text{dom}(s) = [0, \infty) \quad \text{Not DE}$$

$$s(1-1/k) = 1, k = 1, 2, \dots$$

$$\text{dom}(s) = [0, 1) \quad \text{DE, Zeno}$$

$$s(1-1/k) = 1, k = 1, 2, \dots$$

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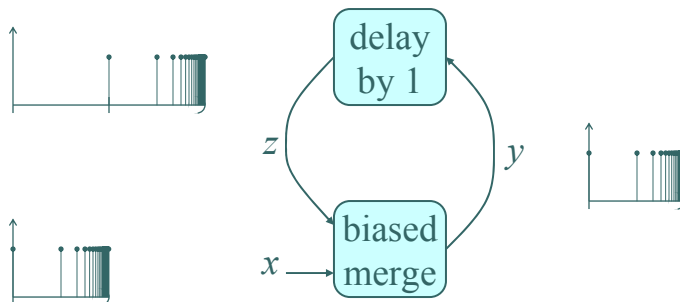
Discrete Event Signals - Properties

- For $T = [0, \infty)$ and any set V of values, the set of all discrete event signals with the prefix order is
 - a poset
 - a CPO
 - a complete lower semilattice (i.e. any subset of signals have a “longest” common prefix)

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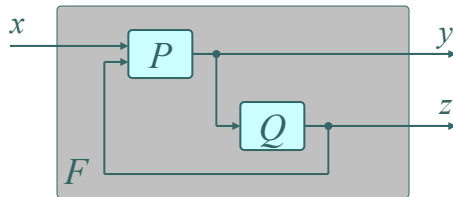
A Discrete Event Process Network



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A Sufficient Condition for Non-Zeno Composition



$$(y, z) = F(x)$$

where (y, z) is the least solution of the equations

$$y = P(x, z)$$

$$z = Q(y)$$

- If processes P and Q are discrete, causal and continuous, and at least one of them is strictly causal, then F is discrete, causal and continuous.
- F is non-Zeno in the sense that if x is non-Zeno, $F(x)$ is non-Zeno.

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Conclusions

- Progress in developing the foundation of the tagged signal model
- Extend Kahn process networks to tagged process networks
- Develop discrete event semantics as a special case of tagged process networks
- Develop a sufficient condition for the non-Zeno composition of discrete event processes

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