Semantic Foundation of the Tagged Signal Model

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Outline

- Motivation
  - Heterogeneous embedded systems
  - The tagged signal model as a semantic metamodel

- Tagged process networks
- Timed process networks
- Discrete event process networks
- The metric structure of signals
- Conclusion
GM vehicles have long been equipped with microprocessors which today, depending on the vehicle, are used for:

- powertrain management
- anti-lock braking systems
- traction control
- stability enhancement
- supplemental inflatable restraint systems
- real-time damping
- navigation systems
- automatic climate control
- remote keyless entry
- head-up display
- entertainment systems
- entry control
- ...
Safety-Critical Applications of Heterogeneous Embedded Systems

Tools and Languages for Designing Heterogeneous Embedded Systems

- Simulink and Stateflow (The MathWorks)
- LabVIEW (National Instruments)
- Modelica (Modelica Association)
- CarSim (Mechanical Simulation Corp.)
- System Studio (Synopsys, Inc.)
- Advanced Design System (Agilent EEsof EDA)
- SystemC (Open SystemC Initiative)
- VHDL and Verilog with Analog and Mixed Signal Extensions
- GME (Vanderbilt University)
- Polis and Metropolis (UC Berkeley)
- Ptolemy and Ptolemy II (UC Berkeley)
The Metamodelling Approach


CDIF (CASE Data Interchange Format)
http://www.eigroup.org/cdif/intro.html

GME 5
The Generic Modeling Environment
http://www.isis.vanderbilt.edu/Projects/gme/

Metropolis Meta Model
http://embedded.eecs.berkeley.edu/metropolis/

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The Tagged Signal Model as a Semantic Metamodel

- Proposed as a Framework for Comparing Models of Computation (Lee & ASV, 1997)
- Opportunities Provided:
  - To compare certain properties of various models of computation, such as their notion of synchrony
  - To define formal relations among signals and process behaviors from different models of computation
  - To facilitate the cross-fertilization of results and proof techniques among models of computation
The Goal of This Presentation

Establish a Semantic Foundation of the Tagged Signal Model (TSM)

Explore the Opportunities Provided by the TSM Framework
Approach

• A Three-Step Development Process

1. Study the properties of a mathematical structure of (sets of) signals

2. Use this structure to characterize the processes that are functions on signal sets, such as continuity and causality

3. Determine conditions under which the characterizations are compositional
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**Signals**

- A signal has a **tag set** $T$, which is a partially ordered set.
- A **down-set** of $T$ is a downward-closed subset of $T$. $\mathcal{D}(T)$ is the set of down-sets of $T$.

![Diagram](image.png)

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

$$signal: D \rightarrow V$$
The Prefix Order on Signals

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

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

![Diagram of prefix order on signals](image)
The Order Structure of Signals

- For any poset $T$ of tags and set $V$ of values, let $S(T,V)$ be the set of all signals from down-sets of $T$ to $V$.
- With the prefix order $\preceq$, $S(T,V)$ is
  - a poset
  - a **complete partial order** (CPO)
  - a complete lower semilattice (i.e. any subset of signals have a “longest” common prefix)
Processes

- **Processes** are relations or functions among signal sets

\[
x \in S(T_1, V_1) \quad P \quad y \in S(T_2, V_2) \quad u \in S(T_3, V_3) \quad Q \quad v \in S(T_4, V_4)
\]

\[
P \subseteq S(T_1, V_1) \times S(T_2, V_2)
\]

- As functions among posets/CPOs, processes may be
  - Monotonic, \( r \preceq s \Rightarrow P(r) \preceq P(s) \)
  - (Scott) Continuous, \( P(\lor D) = \lor P(D) \)
  - Maximal
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 the processes $P$ and $Q$ are Scott-continuous, then $F$ is Scott-continuous.

Scott-continuity is compositional.
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Timed Process Networks

- Each signal in a tagged process network may have its own tag set.
- In a timed process network, all signals share the same totally ordered tag set.
Timed Signals

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

- $s(t) = 1$
- $s(1-1/k) = 1, \ k = 1, 2, \ldots$
- $s(k) = 1, \ k = 0, 1, 2, \ldots$
Timed Processes

\[ D = D_1 \cap D_2 \]
\[ s(t) = s_1(t) + \varepsilon s_2(t) \]

\[ D_2 = D_1 \oplus \{1\} \cup [0, 1) \]
\[ s_2(t) = s_1(t - 1), \text{ when } t \geq 1 \]
\[ s_2(t), \text{ otherwise} \]
A Timed Process Network Example

delay by 1

biased merge

X

Z

Y
A Non-Causal Process in the Network

Lookahead by 1

\[ z: \emptyset \rightarrow V_\varepsilon \]

Biased merge

\[ y: \emptyset \rightarrow V_\varepsilon \]
Causality

• A timed process $P$ is **causal** if
  – it is monotonic, and
  – for all $s: D_1 \to V_1$, $P(s): D_2 \to V_2$, $D_1 \subseteq D_2$

• A timed process $P$ is **strictly causal** if
  – it is monotonic, and
  – for all $s: D_1 \to V_1$, $P(s): D_2 \to V_2$, $D_1 \subset D_2$, or $D_2 = T$
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.
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.

Causality + continuity is compositional.
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Discrete Event Signals

- A timed signal $s : D \rightarrow V_\varepsilon$ is a discrete event signal if for all $t \in D$, $s$ is present at a finite number of times before $t$.

\[
\begin{align*}
\text{dom}(s) &= [0, \infty) \\
rep(s) &= 1, k = 0, 1, 2, \ldots \\
\text{DE, Non-Zeno} & & \\
\text{Not DE} & & \\
\text{DE, Zeno} & & \\
\end{align*}
\]
The Order Structure of DE Signals

• For any totally ordered set $T$ of tags and set $V$ of values, the set of all DE signals $S_d(T,V)$, with the prefix order $\preceq$, is
  - a poset
  - a complete partial order
  - a complete lower semilattice (i.e. any subset of DE signals have a “longest” common prefix)
A DE Process Network Example

delay by 1

biased merge

x

y

z
If DE processes $P$ and $Q$ are Scott-continuous, then $F$ is a DE process and is Scott-continuous.

\[(y, z) = F(x)\]

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

\[y = P(x, z)\]
\[z = Q(y)\]

Discreteness + Scott-continuity is compositional.
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Non-Zeno DE Process Networks

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

$F(x) = (y, z)$ where $(y, z)$ is the least solution of the equations

$y = P(x, z)$

$z = Q(y)$

$F$ is non-Zeno in the sense that if $x$ is non-Zeno, $F(x)$ is non-Zeno.

Discreteness + Causality + Scott-continuity is compositional.
A Brief Recap

- Tagged Signal Model
- Kahn Process Networks
  - Tagged Process Networks
  - Causality
  - Non-Zenoness
- Timed Process Networks
  - Discreteness
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Generalized Ultrametric

Let $X$ be a set and $\Gamma$ be a poset with a minimum element $0$. A function

$$d: X \times X \to \Gamma$$

is a **generalized ultrametric** if for all $x, y, z \in X$ and $\gamma \in \Gamma$,

$$d(x, y) = 0 \iff x = y$$
$$d(x, y) = d(y, x)$$
$$d(x, y) \leq \gamma \text{ and } d(y, z) \leq \gamma \Rightarrow d(x, z) \leq \gamma$$
A Generalized Ultrametric on Tagged Signals

- For any poset $T$ of tags, the set of generalized ultrametric distances, $\Gamma_T$, is
  \[ \Gamma_T = (\mathcal{D}(T), \supseteq) \]
- For any tag set $T$ and value set $V,$
  \[
d_{ds}: S(T, V) \times S(T, V) \rightarrow \Gamma_T,
\]
  \[
d_{ds}(s_1, s_2) = \begin{cases} 
  \text{dom}(s_1 \land s_2) & \text{if } s_1 \neq s_2, \\
  T & \text{if } s_1 = s_2.
\end{cases}
\]
  is a generalized ultrametric.
A Metric “Generator”

- If the tag set $T$ is totally ordered, $\Gamma_T$ is also totally ordered.
- For any totally ordered set $\Sigma$ with a minimum element $\sigma_0$, and a function $g: \Gamma_T \rightarrow \Sigma$

such that
- $g$ is monotonic
- $g(\gamma) = \sigma_0 \Leftrightarrow \gamma = 0$

$g \circ d_{ds}$ is a (generalized) ultrametric on $S(T, V)$. 
Summary

- The Order Structure of Tagged Signals and Tagged Process Networks
- Causality, Discreteness, and Non-Zenoness in Timed Process Networks
- A Generalized Ultrametric on Tagged Signals
- A Formulation of Processes as Labeled Transition Systems for Comparing Implementations or Simulations of Tagged Processes
The End

Questions?