Functional Hybrid Modeling from an Object-Oriented Perspective

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Background (1)

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- Yampa is an instance of FRP embedded in Haskell.
- One central idea: *first-class* reactive components (or models):
 - enables highly structurally dynamic systems to be described declaratively;
 - opens up for meta-modelling without additional language layers.

Background (2)

- Additional interesting aspects:
 - full power of a modern functional language available;
 - polymorphic type system;
 - well-understood underlying semantics.

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 - a powerful, fully-declarative, non-causal modelling language supporting highly structurally dynamic systems;
 - a semantic framework for studying modelling and simulation languages supporting structural dynamism.

 The idea of FHM goes back a few years (PADL 2003). UK research funding (EPSRC) secured very recently. Thus still work in very early stages.

The Rest of the Talk

- A brief introduction to FRP/Yampa as a background.
- Sketch the key ideas of how this may be generalized to a non-causal setting.

Signal functions

Key concept: *functions on signals* (first class).

$$x \qquad y \qquad f$$

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Intuition:

Signal $\alpha \approx \text{Time} \rightarrow \alpha$ x :: Signal T1 y :: Signal T2 SF $\alpha \ \beta \approx \text{Signal} \ \alpha \rightarrow \text{Signal} \ \beta$ f :: SF T1 T2

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Additionally, *causality* required: output at time t must be determined by input on interval [0, t].

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$$\begin{array}{c|c} x(t) & f & y(t) \\ \hline state(t) & \end{array}$$

state(t) summarizes input history x(t'), $t' \in [0, t]$.

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state(t) summarizes input history x(t'), $t' \in [0, t]$. From this perspective, signal functions are: • stateful if y(t) depends on x(t) and state(t)• stateless if y(t) depends only on x(t)Integral is an example of a stateful signal function.

Programming with signal functions

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For example, serial composition:

$$f \rightarrow g \rightarrow$$

A *combinator* can be defined that captures this:

 $(\gg) :: SF \ a \ b \to SF \ b \ c \to SF \ a \ c$

Note: plain function operating on first-class signal function.

The Arrow framework (1)

These diagrams convey the general idea:



The Arrow framework (2)

Some derived combinators:



Example: Constructing a network



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 $\begin{aligned} loop (arr (\lambda(x, y) \to ((x, y), x)) \\ \gg (fst f \\ \gg (arr (\lambda(x, y) \to (x, (x, y))) \gg (g \nleftrightarrow h))) \end{aligned}$

The Arrow notation



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The Arrow notation



proc $x \to do$

rec

$$u \leftarrow f \prec (x, v)$$
$$y \leftarrow g \prec u$$
$$v \leftarrow h \prec (u, x)$$
$$returnA \prec y$$



Some switching combinators:

• switch :: SF $a (b, Event c) \rightarrow (c \rightarrow SF a b)$ $\rightarrow SF a b$

• $pSwitchB :: Functor \ col \Rightarrow$ $col \ (SF \ a \ b)$ $\rightarrow SF \ (a, \ col \ b) \ (Event \ c)$ $\rightarrow (col \ (SF \ a \ b) \rightarrow c \rightarrow SF \ a \ (col \ b))$ $\rightarrow SF \ a \ (col \ b)$

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- Supports hybrid (mixed continuous and discrete time) systems: option type *Event* represents discrete-time signals.
- Supports dynamic system structure through switching combinators:



Example: Space Invaders



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- First-class *relations* on signals instead of functions on signals to enable non-causal modeling.
- Employ state-of-the-art symbolic and numerical methods for sound and efficient simulation.
- Adapted switch constructs.

First class signal relations

The type for a relation on a signal of type Signal α :

 $\mathrm{SR}\;\alpha$

Specific relations use a more refined type; e.g. the derivative relation:

der :: SR (Real, Real)

Since a signal carrying pairs is isomorphic to a pair of signals, *der* can be understood as a binary relation on two signals.

Defining relations

The following tentative construct denotes a signal relation:

sigrel pattern where equations

The pattern introduces *signal variables* which at each point in time are going to be bound to to a "sample" of the corresponding signal.

Given p :: t, we have: sigrel p where ... :: SR t

Equations

Let $e_i :: t_i$ be non-relational expressions possibly introducing new signal variables.

Point-wise equality; the equality must hold for all points in time:

 $e_1 = e_2$

Relation "application"; the relation must hold for all points in time:

 $sr \diamond e_3$

Here, *sr* is an *expression* having type $SR t_3$.

Equations: examples

Consider a differential equation like x' = f(x, y). This equation could be written: $der \diamond (x, f(x, y))$

For convenience, *syntactic sugar* closer to standard mathematical notation could be considered:

 $\operatorname{der}(x) = f(x, y)$

Here, **der** is **not** a pure function operating only on instantaneous signal values since it depends on the history of the signal.

Modeling electrical components (1)

The type Pin is assumed to be a record type describing an electrical connection. It has fields v for voltage and i for current.

twoPin :: SR (Pin, Pin, Voltage) *twoPin* = **sigrel** (p, n, v) where v = p.v - n.v

$$p.i + n.i = 0$$

Modeling electrical components (2)

 $\begin{aligned} \textit{resistor} :: \texttt{Resistance} &\to \texttt{SR} (\texttt{Pin},\texttt{Pin}) \\ \textit{resistor}(r) = \textbf{sigrel} (p, n) \textbf{ where} \\ & twoPin \diamond (p, n, v) \\ r \cdot p.i = v \\ \textit{inductor} :: \texttt{Inductance} &\to \texttt{SR} (\texttt{Pin},\texttt{Pin}) \\ \textit{inductor}(l) = \textbf{sigrel} (p, n) \textbf{ where} \\ & twoPin \diamond (p, n, v) \\ & l \cdot \textbf{der}(p.i) = v \end{aligned}$

Modeling electrical components (3)

 $\begin{array}{l} capacitor :: \texttt{Capacitance} \to \texttt{SR} \ (\texttt{Pin}, \texttt{Pin}) \\ capacitor(c) = \textbf{sigrel} \ (p, n) \ \textbf{where} \\ twoPin \diamond (p, n, v) \\ c \cdot \textbf{der}(v) = p.i \end{array}$

Modeling an electrical circuit (1)

simpleCircuit :: SR Current
simpleCircuit = sigrel i where

 $\begin{aligned} resistor(1000) & (r1p, r1n) \\ resistor(2200) & (r2p, r2n) \\ capacitor(0.00047) & (cp, cn) \\ inductor(0.01) & (lp, ln) \\ vSourceAC(12) & (acp, acn) \\ ground & gp \end{aligned}$

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Modeling an electrical circuit (2)



connect acp, r1p, r2pconnect r1n, cpconnect r2n, lpconnect acn, cn, ln, gpi = r1p.i + r2p.i

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- Adapting non-causal modelling and simulation methods to a setting with first class signal relations: causality analysis, symbolic processing code generation after each switch.
- Guaranteeing compositional correctness statically through the type system to the extent possible; e.g. employing dependent types to keep track of variable/equation balance.