T U M

I N S T I T U T F Ü R I N F O R M A T I K

Specification of Real-Time and Hybrid Systems in FOCUS

Olaf Müller
Peter Scholz

TUM-I9627
Juni 1996

T E C H N I S C H E U N I V E R S I T Ä T M Ü N C H E N

TUM-INFO-06-1996-I9627-350/1.-FI

Typescript: ---

Druck: Mathematisches Institut und
Institut für Informatik der
Technischen Universität München

Specification of Real-Time and Hybrid Systems in FOCUS *

Olaf Müller Peter Scholz
Institut für Informatik, Technische Universität München
D-80290 München, Germany
E-mail: {mueller,scholzp}@informatik.tu-muenchen.de

*This work is partially sponsored by the German Federal Ministry of Education and Research (BMBF) as part of the compound project "KorSys" and by BMW (Bayerische Motoren Werke AG).

Abstract

Functional specifications in FOCUS have been used to specify and verify designs of a number of reactive, discrete systems. In this paper we extend this specification style to deal with real-time and hybrid systems. As mathematical foundation we employ Banach's fixed point theory in metric spaces. The goal is to show that the theory used for discrete functional specifications smoothly carries over to real-time and hybrid systems. An example of a thermostat specification illustrates the method.

Contents

1 Introduction

Hybrid systems are dynamical systems consisting of both discrete and continuous components. They are used to model the behavior of embedded real-time systems in a physical environment. This topic is becoming very active in Computer Science, due to the increasing importance of embedded and real-time systems and the emergence of results showing that some techniques used for the specification and verification of reactive, digital systems can be adapted to deal with hybrid systems. By their nature hybrid systems form an interdisciplinary topic that lies at the junction of Computer Science and Control Theory.

From the control theory or dynamic systems viewpoint, it is interesting to investigate hybrid systems concerning the questions usually asked in these disciplines, such as the problems of analysis of dynamic behavior, realizability, and controller synthesis.

On the other hand, from the traditional computer scientist viewpoint, hybrid systems can be seen as a natural extension of reactive systems by the introduction of analog components into the model. Therefore, computer scientists rather investigate how to carry over their description and specification languages for reactive and/or real-time systems to hybrid systems, together with their proposed methodology for analysis, verification, and refinement.

In this paper we take the viewpoint of a computer scientist and extend the formalism of *functional specification* in FOCUS [BDD+93, Bro93, BD92] to deal with real-time and hybrid systems. Functional specifications describe the behavior of a system as a network of functions, where every function processes infinite streams of incoming messages and yields infinite streams of outgoing messages. In the discrete setting, several approaches have been taken to give functional specifications a semantics:

- In [Bro93, BD92] domain theory is used to develop a semantic model for discrete stream processing functions together with a tailored refinement methodology.
- In [GS96] metric spaces are employed to give a semantics for functionally specified, discrete mobile data-flow networks.

We follow the second approach and extend the static parts of [GS96] to a description and specification method for hybrid systems. Our goal is to show that only slight modifications must be carried through, so that the whole theory smoothly carries over to the hybrid case:

First, discrete streams of type $I\!N \to M$ have to be replaced by dense streams of type $I\!R_+ \to M$, where M denotes the set of all messages. Second, the property of a discrete stream processing function to be *strongly pulse driven* (delay between input and output of at least one time step) has to be changed to an adequate property for dense streams, called *delayed* (delay between input and output of at least $\delta > 0$). These modifications allow us to employ Banach's fixed point theorem as in the discrete case to prove the well-definedness of a functional specification.

1.1 Related Work

Recently, a number of description and specification languages for reactive and/or real-time systems together with their proposed methodology for analysis, verification, and refinement were extended to deal with hybrid systems. An overview of the growing field can be found in [GNRR93, AKNS95, AHS96].

For example, in [ACH+95] a theory of hybrid finite automata has been developed. For verification purposes, these automata are restricted to linear hybrid automata, where all variables follow piecewise-linear trajectories. For this subclass of systems the standard symbolic model checking techniques for reachability analysis can be carried over. Approximation techniques allow a treatment of systems whose verification problem is not decidable and for which the iterative verification procedures do not converge.

Besides model checking, also refinement techniques have been extended to deal with hybrid systems. As examples, that are by no means representative, we mention I/O Automata [LSVW95] and TLA, which has been extended to TLA+ [Lam93].

1.2 Overview

The rest of the paper is organized as follows: Section 2 introduces stream processing functions and relates them to the corresponding notions in the theory of metric spaces. In Section 3 composition operators are defined that are used to build networks out of single functions. In particular, the mathematical foundation of the feedback operator is presented. In Section 4 a short guideline for the specification of components with the concepts introduced so far is presented. Section 5 illustrates the specification method with the simple example of a thermostat. Finally, Section 6 gives a conclusion and highlights topics for future work.

2 Specification with Stream Processing Functions

We regard a distributed system as a network of components that exchange messages via directed channels. On every input or output channel messages are received from, or sent to, the environment. Therefore, every channel reflects an input or output communication history of the system.

The system itself is described by a set of functions, where each function processes input histories and produces output histories according to its specification. To describe underspecification or nondeterminism we use sets of functions instead of single functions.

2.1 Dense Communication Histories

Communication histories of discrete systems can be modeled by sequences of messages, i.e., functions of type $\mathbb{N} \to M$, where M denotes the set of all messages [Bro93, BDD+93]. For hybrid systems this model has to be extended to incorporate real time. One possibility is to add real time stamps. In the literature this is known as *sampling* semantics [MP93]. Here, instead, we develop a *super dense* semantics and therefore introduce real time or *dense* streams.

Let M be the (potentially infinite) set of all messages. A *dense stream* x over a set M is represented by a total function $x : \mathbb{R}_+ \to M$, where $\mathbb{R}_+$ denotes the set of all non-negative real numbers. Since we describe reactive systems, which continuously respond to stimuli from the environment, time never halts, and we use $\mathbb{R}_+$ as the time scale instead of time intervals. The set of all dense streams is denoted by $M^{\mathbb{R}_+}$. For every dense stream x we abbreviate the restriction $x|_{[0,t]}$ by $x \downarrow t$.

In order to motivate the usefulness of this definition we have adapted the example of a thermostat from [ACH+95], where it is presented by means of hybrid automata.

Example 1 (Dense Stream) *The temperature of a room in a cool environment can be modeled by a dense stream x. We assume that without the presence of any heater, the temperature decreases according to the exponential function $x(t) = \Theta e^{-Kt}$, where t denotes the time, Θ the initial temperature, and K is a positive constant determined by the room.*

A mathematical treatment of functional specifications requires dealing with feedback loops. In the discrete case, dealing with streams of type $\mathbb{N} \to M$, the semantics of such loops has been successfully described as least fixed points of functions over domains [Bro93, BDD+93]. The underlying mathematical model is Scott's domain theory [SG90, Win93]. Fixed points of stream processing functions over dense streams, however, are more naturally and elegantly described by the fixed point theory of Banach. It is based upon the mathematical background of metric spaces. In order to specify loops of stream processing functions in Section 3, we therefore introduce the main concepts of metric space theory.

Definition 1 (Metric Space) *A* metric space *is a pair (D, d) consisting of a nonempty set D and a mapping $d : D \times D \to \mathbb{R}$, called a* metric *or a* distance, *which has the following properties:*

(**1**) $\forall x, y \in D: \quad d(x,y) = 0 \quad \Leftrightarrow \quad x = y$
(**2**) $\forall x, y \in D: \quad d(x,y) = d(y,x)$
(**3**) $\forall x, y, z \in D: \quad d(x,y) \leq d(x,z) + d(z,y).$

We need a metric for dense streams, which is defined in the sequel.

Definition 2 (The Metric of Streams) *The metric space of dense streams $(M^{\mathbb{R}_+}, d)$ is for all $x, y \in M^{\mathbb{R}_+}$ defined as follows:* $d(x,y) = inf\{2^{-t} \mid t \in \mathbb{R}_+ \wedge x \downarrow t = y \downarrow t\}$.

From this definition a metric $d^{(n)}$ for n-tuples of streams $(M^{\mathbb{R}_+})^n$ can be easily derived.

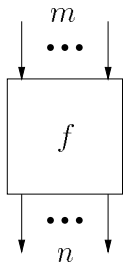

Figure 1: Stream Processing Function

Let $n \in I\!N$ and $x, y \in (M^{I\!R_+})^n$ then $d^{(n)}(x, y)$ is defined as

$$d^{(n)}(x, y) = max\{d(x_i, y_i) \mid 1 \leq i \leq n\}.$$

A metric space (D, d) is called *complete* whenever each Cauchy sequence converges to an element of D [Eng77]. The metric space on stream tuples $((M^{I\!R_+})^n, d^{(n)})$ is complete [Eng77]. Complete metric spaces are a presupposition for Banach's fixed point theorem. This theorem, which will be explained later on, guarantees — under certain assumptions — the existence of a unique fixed point of loops in functional specifications.

2.2 Stream Processing Functions

Components of real time or hybrid systems can be functionally specified by stream processing functions over dense streams. Components are connected by directed channels to form a network. Each channel links an *input port* to an *output port*. A (m, n)-ary stream processing function with m input and n output ports is a function f with

$$f : (M_1^{I\!R_+})^m \rightarrow (M_2^{I\!R_+})^n$$

where M_1 and M_2 represent two (not necessarily different) sets of messages. The graphic notation of f is pictured in Fig. 1. If we want to express some kind of nondeterminism we describe components by a set of stream processing functions rather than by a single function.

Our operational understanding that stream processing functions model interacting components leads to a basic requirement for them. An interactive component is not capable to take back an output message that it has already emitted. This requirement can be fulfilled by a certain kind of stream processing functions, namely behaviors.

A stream processing function is said to be a *behavior* if its input until time t completely determines its output until time t. It is said to be a *delayed behavior* if its input until time t completely determines its output until time $t + \delta$ for $\delta > 0$. In other words, a delayed behavior imposes a delay of at least an arbitrarily small real value between input and output. Here, δ denotes the delay of f. It is quite realistic to assume components to

be delayed because reactive systems always need a certain time to react. Instantaneous reactions, however, can be expressed by (non-delayed) behaviors.

Definition 3 ((Delayed) Behavior) *A (m,n)-ary stream processing function f is called a* behavior *if*

$$\forall x,y \in (M^{\mathbb{R}_+})^m,\ t \in \mathbb{R}_+ : x \downarrow t = y \downarrow t \Rightarrow f(x) \downarrow t = f(y) \downarrow t$$

and a delayed behavior (with delay $\delta > 0$) *if*

$$\forall x,y \in (M^{\mathbb{R}_+})^m,\ t \in \mathbb{R}_+ : x \downarrow t = y \downarrow t \Rightarrow f(x) \downarrow (t+\delta) = f(y) \downarrow (t+\delta).$$

Note that the operator $\downarrow$ is overloaded to stream tuples in a point-wise style, i.e., $x \downarrow t$ for a stream tuple $x \in (M^{\mathbb{R}_+})^m$ denotes the tuple we get by applying $\downarrow t$ to each component of x.

The equivalent property in Scott's theory is monotonicity. From a theorem by Knaster and Tarski it is well-known that monotonic functions over complete partial orders have a least fixed point [Win93].

We model specifications by sets of (delayed) behaviors. They can be composed into networks of functions, which themselves behave as (delayed) behaviors. For this purpose, we will introduce three composition operators in the next section. For one of them, the feedback operator, the existence of a unique fixed point of the feedback loop is guaranteed only for *delayed* behaviors. To prove this formally we introduce a notion corresponding to delayed behaviors in metric space theory.

Definition 4 (Lipschitz Functions) *Let (D_1,d_1) and (D_2,d_2) be metric spaces and let $f : D_1 \to D_2$ be a function. We call f a* Lipschitz function *if there is a constant $c \geq 0$ such that the following condition is satisfied for all $x,y \in D_1$:*

$$d_2(f(x),f(y)) \leq c \cdot d_1(x,y).$$

The Lipschitz constant $Lip(f)$ of a Lipschitz function f is denoted by the infimum of all c that fulfill the above mentioned inequation. If $Lip(f) \leq 1$ we call f non-expansive. *If $Lip(f) < 1$ we call f* contractive.

The following theorem relates the notions of behaviors and delayed behaviors to non-expansiveness and contractivity. Whereas the first ones have a operational justification, the latter ones represent their transfer to metric space theory and will be used as a requirement for Banach's fixed point theorem.

Theorem 1 *A stream processing function is a delayed behavior iff it is contractive with respect to the metric of stream tuples. A stream processing function is a behavior iff it is non-expansive with respect to the metric of stream tuples.*

Proof 1 *We prove the first statement of the theorem. First, we prove the only-if-direction. Suppose that $d^{(m)}(x,y) = 2^{-t_0}$ and that f is a delayed behavior with delay δ. $d^{(m)}(x,y) =$*

2^{-t_0} *implies that* $x \downarrow t_0 = y \downarrow t_0$. *Therefore,* $f(x) \downarrow (t_0 + \delta) = f(y) \downarrow (t_0 + \delta)$. *Finally, we get* $inf\{2^{-t} \mid t \in \mathbb{R}_+ \wedge f(x) \downarrow t = f(y) \downarrow t\} \leq 2^{-(t_0+\delta)} = 2^{-\delta} \cdot d^{(m)}(x, y)$. *Since* $2^{-\delta} < 1$ *for all* $\delta > 0$, f *is contractive.*

Now, we prove the if-direction. Suppose that $d^{(m)}(x, y) = 2^{-t_1}$, $d^{(n)}(f(x), f(y)) = 2^{-t_2}$, *and that* f *is contractive, i.e.,* $\exists c < 1 : \forall x, y : \; d^{(n)}(f(x), f(y)) \leq c \cdot d^{(m)}(x, y)$. *Then* $2^{t_1 - t_2} \leq c < 1 = 2^0$. *This implies because of the monotonicity of the logarithmic function that* $t_1 < t_2$. *We can find some* $\delta > 0$ *with* $t_1 + \delta = t_2$. *As a consequence we get* $x \downarrow t_1 = y \downarrow t_1 \Rightarrow f(x) \downarrow (t_1 + \delta) = f(y) \downarrow (t_1 + \delta)$. *In other words,* f *is a delayed behavior. The second equivalence can be proven accordingly.*

3 Composition Operators

The definition of networks is the main structuring principle on the functional specification level. There is no (semantical) difference in principle between a single component and a network of components. A network can be defined either by recursive equations or by special composition operators. We choose the second alternative and consider three basic composition operators, namely *sequential/parallel composition* and *feedback.*

In our functional specification technique, networks of components can be represented by directed graphs, where the nodes represent components and the edges represent point-to-point, directed communication channels (see, for instance, Fig. 2).

3.1 Sequential Composition

Sequential composition is simply defined by functional composition of two stream processing functions. The graphic representation of this composition is pictured in Fig. 2.

Definition 5 (Sequential Composition) *Let* f *and* g *be* (m, n)*-ary and* (n, k)*-ary stream processing functions, respectively. Then* $f \circ g$ *is the* (m, k)*-ary stream processing function defined by*
$(f \circ g)(x) = g(f(x))$.

The following theorem and corollary depict important properties of the sequential composition:

Theorem 2 *The sequential composition of two Lipschitz functions* $f : D_1 \to D_2$ *and* $g : D_2 \to D_3$ *is a Lipschitz function with constant* $Lip(f) \cdot Lip(g)$.

Proof 2 $d_3(g(f(x_1)), g(f(x_2))) \leq Lip(g) \cdot d_2(f(x_1), f(x_2)) \leq Lip(g) \cdot Lip(f) \cdot d_1(x_1, x_2)$.

Corollary 1 *The sequential composition of two behaviors is a behavior. The sequential composition of two delayed behaviors with delays* δ_1 *and* δ_2, *respectively, is a delayed behavior with delay* $\delta_1 + \delta_2$. *The sequential composition of a behavior and a delayed behavior is a delayed behavior.*

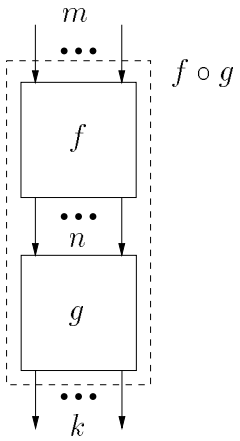

Figure 2: Sequential Composition

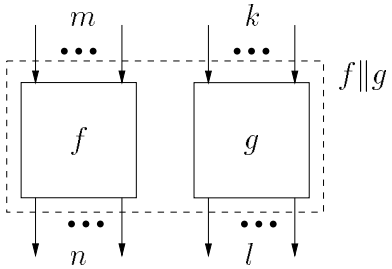

Figure 3: Parallel Composition

Due to the above theorem, the proof of this corollary is obvious.

3.2 Parallel Composition

The parallel composition is defined intuitively. Sticking two components orthogonally together yields a component which input/output ports consists of all input/output ports of the composed components (see Fig. 3). Formally:

Definition 6 (Parallel Composition) *Let f and g be (m,n)-ary and (k,l)-ary stream processing functions. Then $f\|g$ is the $(m+k,n+l)$-ary stream processing function defined by*

$$(f\|g)(x_1,\ldots,x_{m+k}) = (f(x_1,\ldots,x_m), g(x_{m+1},\ldots,x_{m+k})).$$

As for the sequential composition, an equivalent property can also be formulated for the parallel composition:

Theorem 3 *The parallel composition of two behaviors is a behavior. The parallel composition of two delayed behaviors with delays δ_1 and δ_2, respectively, is a delayed behavior*

with delay $min(\delta_1, \delta_2)$. The parallel composition of a behavior and a delayed behavior is a behavior.

Proof 3 *We prove the second statement of the theorem. Let f be a (m,n)-ary delayed behavior with delay δ_1 and g be a (k,l)-ary delayed behavior with delay δ_2. Without loss of generality we assume that $\delta_1 < \delta_2$. Let $x, y \in (M^{\mathbb{R}_+})^k$, then $g(x)\downarrow(t+\delta_2) = g(y)\downarrow(t+\delta_2)$ implies that $g(x)\downarrow(t+\delta_1) = g(y)\downarrow(t+\delta_1)$. The other statements can be proven accordingly.*

Note that the sequential composition of a behavior and a delayed behavior is a delayed behavior, whereas the parallel composition of a behavior and a delayed behavior is "only" a behavior.

3.3 Feedback Operator

Systems described by functional specifications may contain loops. In the graphic notation, this is denoted by circular graphs (Fig. 4). The feedback operator feeds k output channels back to k input channels of a $(m+k, n+k)$-ary delayed behavior.

Definition 7 (Feedback Operator) *Let $f : (M_1^{\mathbb{R}_+})^m \times (M^{\mathbb{R}_+})^k \to (M_2^{\mathbb{R}_+})^n \times (M^{\mathbb{R}_+})^k$ be a $(m+k, n+k)$-ary delayed behavior. Then $\mu^k f$ is a (m,n)-ary delayed behavior such that the value $(z_1, \ldots, z_n)$ of $(\mu^k f)(x_1, \ldots, x_m)$ is calculated as follows:*

$$(z_1, \ldots, z_n, y_1, \ldots, y_k) = f(x_1, \ldots, x_m, y_1, \ldots, y_k)$$

where $(y_1, \ldots, y_k)$ is the solution of the equation

$$(y_1, \ldots, y_k) = g_{(x_1,\ldots,x_m)}(y_1, \ldots, y_k).$$

Here $g_{(x_1,\ldots,x_m)}$ is defined as a (k,k)-ary delayed behavior:

$$g_{(x_1,\ldots,x_m)}(y_1, \ldots, y_k) = \pi_{n+1,n+k}(f(x_1, \ldots, x_m, y_1, \ldots, y_k))$$

where $\pi_{n+1,n+k}$ denotes the projection on the last k ports.

The central issue of our contribution is that the fixed point operator is well-defined, i.e., that the unique solution of

$$(y_1, \ldots, y_k) = g_{(x_1,\ldots,x_m)}(y_1, \ldots, y_k)$$

exists. The existence of this fixed point is guaranteed by Banach's fixed point theorem:

Theorem 4 (Banach's Fixed Point Theorem) *Let (D,d) be a complete metric space and $f : D \to D$ a contractive function. Then there exists an $x \in D$, such that the following holds:*

(1) $x = f(x)$ *(x is a fixed point of f)*

(2) $\forall y \in D : \; y = f(y) \Rightarrow y = x$ *(x is unique)*

(3) $\forall z \in D : \; x = lim_{n\to\infty} f^n(z)$ *where*

$$f^0(z) = z$$
$$f^{n+1}(z) = f(f^n(z))$$

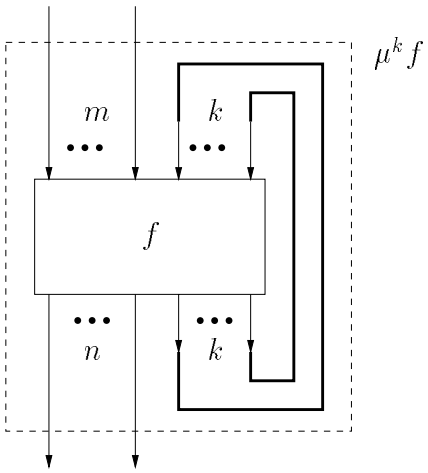

Figure 4: Feedback Operator

Proof 4 *For instance, see [Sut75].*

In the context of this paper, we can apply Banach's theorem in the following way. First of all, the metric space $((M^{\mathbb{R}_+})^k, d^{(k)})$ is complete. Secondly, f is a $(m+k, n+k)$-ary delayed behavior and therefore contractive. Remember that f need not to be a basic stream processing function, but can also be a composed, delayed behavior. Moreover, also $g_{(x_1,\ldots,x_m)} : (M^{\mathbb{R}_+})^k \to (M^{\mathbb{R}_+})^k$ is by definition a contractive function. Altogether, all assumptions of Banach's fixed point theorem are fulfilled and the existence of a unique fixed point $(y_1, \ldots, y_k)$ of $g_{(x_1,\ldots,x_m)}$ is ensured. Hence, every delayed behavior has a unique fixed point.

Banach's fixed point theorem is the counterpart of Knaster/Tarski's fixed point theorem in the theory of metric spaces. However, note that Knaster/Tarski's theorem only guarantees the existence of a *least* fixed point, i.e., that potentially more than one fixed point can exist. In contrast, Banach's fixed point theorem guarantees the existence of a unique fixed point.

Again it is a straightforward proof to show that the feedback $\mu^k f$ is a delayed behavior, provided that f is a delayed behavior.

4 Specification of Components

The here presented notion of specification of components is defined according to [BDD+93]. A component is modeled by a non-empty set of behaviors, which is represented by a predicate on functions. Each function from this set corresponds to one particular, deterministic behavior.

Hence, a functional specification of a component C is given by the predicate

$$P : ((M_1^{\mathbb{R}_+})^m \to (M_2^{\mathbb{R}_+})^n) \to \mathbb{B}$$

which describes the following set S of (m,n)-ary behaviors

$$\{f : (M_1^{\mathbb{R}_+})^m \to (M_2^{\mathbb{R}_+})^n \mid P(f) \wedge f \text{ is a behavior}\}.$$

This is denoted by $[\![C]\!] = S$. Every (m,n)-ary behavior describes a potential input/output behavior of the component. The composition operators defined in Section 3 can easily be lifted to sets:

$$\begin{aligned} [\![C_1 \circ C_2]\!] &= \{f \circ g \mid f \in [\![C_1]\!] \wedge g \in [\![C_2]\!]\} \\ [\![C_1 \parallel C_2]\!] &= \{f \parallel g \mid f \in [\![C_1]\!] \wedge g \in [\![C_2]\!]\} \\ [\![\mu^k C]\!] &= \{\mu^k(f) \mid f \in [\![C]\!]\}. \end{aligned}$$

If the above set only contained one single element, it would represent a deterministic component.

In most cases components are modeled not only by behaviors but by *delayed* behaviors. Delayed behaviors with delay δ usually have an undefined output stream during the interval $[0,\delta)$. Thus, during this interval, nothing can be said about the input/output behavior of the component. The component remains underspecified in this time and therefore behaves non-deterministically.

To abolish this underspecification, we can assume that the component generates a predefined value during the interval $[0,\delta)$, as we shall see in the example.

5 Example

In this section we give a functional specification of a thermostat, a simple hybrid system used as an introductory example in [ACH+95]. The temperature of a room is controlled by a thermostat, which continuously senses the temperature and turns a heater on and off. The temperature is governed by differential equations.

When the heater is off, the temperature $Temp$ of the environment, denoted by the dense stream x, decreases according to the function $x(t) = \Theta e^{-Kt}$ (see Example 1). When the heater is on, the temperature of the environment follows the function $x(t) = \Theta e^{-Kt} + h(1-e^{-Kt})$, where h is a constant that depends on the power of the heater, Θ is the initial temperature of the room, and K is a constant determined by the environment. K can be considered to be direct proportional to the geometric size of the room. We wish to keep the temperature between min and max degrees and turn the heater on and off accordingly.

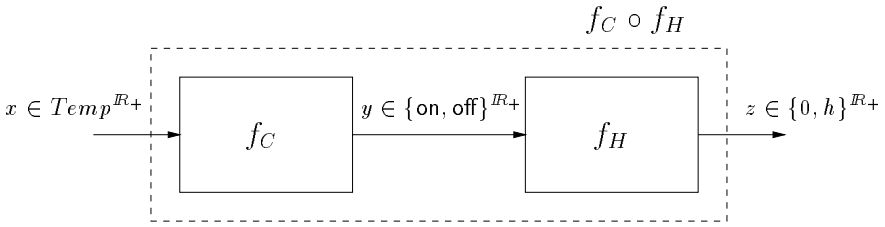

Figure 5: Thermostat Modeled as Open System

5.1 Thermostat as Open System

The controlling part of the resulting system for this informal description is shown in Fig. 5. The system consists of the two components *Control* and *Heater*. The first one is described by a set *Control* of functions f_C, described by the predicate

$$P_C : (Temp^{\mathbb{R}+} \to \{\mathsf{on}, \mathsf{off}\}^{\mathbb{R}+}) \to \mathbb{B}.$$

Each function f_C with $P_C(f_C) = true$ produces signals off or on, if the incoming stream of temperature signals overshoots max or undershoots min, respectively. These signals serve as an input stream for the *Heater*:

$$P_H : (\{\mathsf{on}, \mathsf{off}\}^{\mathbb{R}+} \to \{0, h\}^{\mathbb{R}+}) \to \mathbb{B}$$

that produces the corresponding heating power, which can be 0 or h. Note that we model only the heating power of the heater, but not the resulting absolute temperature. The temperature of the room is regarded as part of the system's environment. This is different from [ACH+95], where the temperature is an inherent part of the system description. Therefore, the environment is there modeled as part of the system.

In fact, the model of hybrid automata does not emphasize on an interface concept to the environment, so that [ACH+95] describes merely closed systems without dividing the overall specification into system and environment. The advantage of our approach is its modularity, which allows us to separate the environment from the system specification. This is one of the essential issues of our approach. The application of our functional specification method to the thermostat example shows that indeed only the environment behaves continuously. The system itself, i.e., *Controller* and *Heater* behave as value-discrete components. They produce signals on, off, 0, and h. The environment, however, is characterized by the temperature, which is denoted by a real-valued ($Temp$) stream.

In the sequel, we give the precise specifications of the components *Control* and *Heater*. First of all, we define $Control = \{f_C \mid P_C(f_C) \wedge f_C$ is a delayed behavior$\}$:

$$f_C(x) = y$$

where the output stream y is defined by the predicate P_C:

$$\forall y \in \{\mathsf{on}, \mathsf{off}\}^{I\!R_+}\ \forall t \in I\!R_+ : \begin{array}{lcl} x(t) \leq \mathsf{min} & \Rightarrow & y(t+\delta_C) = \mathsf{on} \quad \wedge \\ x(t) \geq \mathsf{max} & \Rightarrow & y(t+\delta_C) = \mathsf{off} \quad \wedge \\ \mathsf{min} < x(t) < \mathsf{max} & \Rightarrow & y(t+\delta_C) = y(t). \end{array}$$

Here $\delta_C > 0$ denotes the delay of the component *Control*. However, this specification leaves the value $y(t)$ in the interval $[0, \delta_C)$ unspecified. We can abolish this under-specification by simply extending f_C. We define $y(t) = \mathsf{off}$ in this interval and get a deterministic component, i.e., a one-element set. Now, we specify the $Heater = \{f_H \mid P_H(f_H) \wedge f_H \text{ is a delayed behavior}\}$:

$$f_H(y) = z$$

where the output stream z is defined by the Boolean predicate P_H:

$$\forall z \in \{0, h\}^{I\!R_+}\ \forall t \in I\!R_+ : \begin{array}{lcl} y(t) = \mathsf{off} & \Rightarrow & z(t+\delta_H) = 0 \quad \wedge \\ y(t) = \mathsf{on} & \Rightarrow & z(t+\delta_H) = h. \end{array}$$

Again, to avoid under-specification, we define $z(t) = 0$ for $t \in [0, \delta_H)$ and get a deterministic component, represented by a one-element set. Being a deterministic component, the whole thermostat can then be described using the sequential composition

$$Control \circ Heater.$$

This component has delay $\delta_C + \delta_H$ according to Corollary 1.

5.2 Thermostat as Closed System

To model the continuous part of the specification, we add the environment

$$Env = \{f_E \mid P_E(f_E) \wedge f_E \text{ is a behavior}\}$$

to it, where the predicate P_E has type

$$P_E : (\{0, h\}^{I\!R_+} \rightarrow Temp^{I\!R}) \rightarrow I\!B$$

and we get a closed system (Fig. 6).

Env is specified as a component that cools the temperature down according to the exponential function Θe^{-Kt} (see also Example 1), if the *Heater* is off. When it is on, the temperature follows the function $\Theta e^{-Kt} + h(1 - e^{-Kt})$. We combine these two functions to one function $x(t) = \Theta e^{-Kt} + z(t) \cdot (1 - e^{-Kt})$ and get:

$$f_E(z) = x$$

where the output stream $x \in Temp^{I\!R_+}$ is defined by the differential equation:

$$x'(t) = z(t) - K\Theta x(t)$$

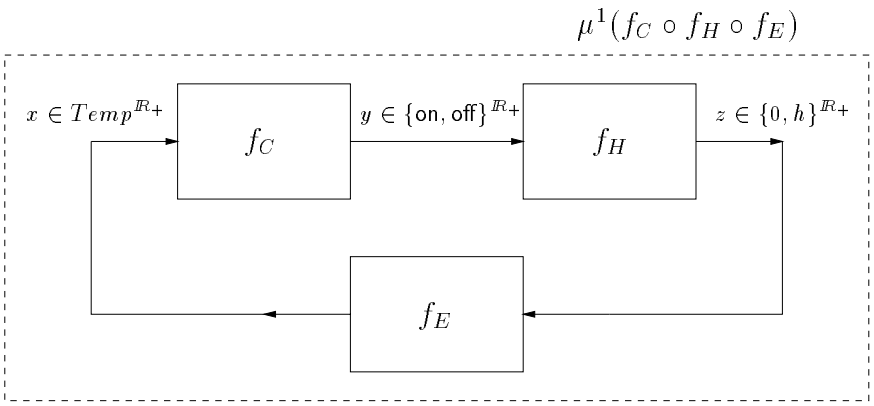

Figure 6: Thermostat Modeled as Closed System

where $x'(t)$ denotes the first differentiation of $x(t)$. Using the product rule for differentiations, it can be calculated as follows:

$$\begin{aligned} x'(t) &= -\Theta K e^{-Kt} + z'(t) - (z'(t) \cdot e^{-Kt} + z(t) \cdot (-K) e^{-Kt}) \\ &= K(z(t) - \Theta) e^{-Kt} \\ &= z(t) - K\Theta x(t) \end{aligned}$$

and get $P_E(f_E) :\Leftrightarrow$

$$\forall z \in \{0, h\}^{\mathbb{R}+} \; \forall x \in Temp^{\mathbb{R}+} \; \forall t \in \mathbb{R}_+ : f_E(z) = x \wedge x'(t) = z(t) - K\Theta x(t)$$

as overall result. Env and $Control \circ Heater$ form a closed system in the shape of a feedback:

$$\mu^1(Control \circ Heater \circ Env).$$

This definition is well-defined, as the occurring fixed point is uniquely determined according to our theory in Section 3: as $Control \circ Heater$ contains one single contractive function with delay $\delta_C + \delta_H$, $Control \circ Heater \circ Env$ is contractive according to Corollary 1, even if all functions in Env have no delay at all. Therefore, Banach's fixed point theorem can be applied.

6 Conclusion and Further Work

We have shown that the specification formalism of discrete timed stream processing functions can easily be extended to deal with real-time and hybrid systems. We could give functional specifications with feedback a semantical foundation by introducing the concept of delayed behaviors that allows us to employ Banach's fixed point theorem.

Characteristic of our approach is that our functional model naturally reflects the physical and conceptual structure of the system and its environment. In particular, it is possible

to distinguish clearly between system and environment. In the thermostat example this structural clarity has been documented. Furthermore, we have the impression that the concept of well-known mathematical functions leads to a simple and clear specification style.

In the discrete case a verification methodology by (structural, behavioral, and interface) refinements is well studied and understood. Further work should explore how to carry over these results to the hybrid setting. Generally, there are several possibilities to extend our specification method with a verification methodology:

- As our specification style provides a clear distinction between environment and system, it seems to be natural to refine the system to a discrete description. Therefore a transformation from analog (but inherently discrete) to discrete specifications has to be investigated. In the case of the thermostat, e.g., the system itself (*Controller* and *Heater*) is translated to a discrete system by simply abstracting dense streams to discrete streams ($\mathbb{N} \to M$), as it is already value-discrete. Using such a transformation the *system* can be refined by the well known discrete verification methodology, whereas the *environment* is still described with continuous mathematics.

- Alternatively, one could stay in the hybrid specification style and investigate how the discrete refinement concepts carry over to dense streams and continuous values.

- Apart from these verification approaches through refinements it would be interesting to investigate property checking methods. As hybrid model checkers such as HyTech [ACH+95] are inherently connected to state based descriptions, an appropriate formalism has to be developed for the functional description style. An advantage of property checking is that it verifies properties of both system and environment, whereas refinements cover the systems behavior only.

Finally, it would be interesting to analyze another type of streams as functions of type $\mathbb{N} \to M \times \mathbb{R}$, yielding a sampling semantics.

Acknowledgment

Thanks are owed to Manfred Broy who provided first ideas concerning both dense streams and behaviors. The authors have benefited from many discussion with Ketil Stølen on this and on related topics. We also thank Radu Grosu and Ketil Stølen for the stimulating technical report on discrete timed streams.

References

[ACH+95] R. Alur, C. Courcoubetis, N. Halbwachs, T.A. Henzinger, P.-H. Ho, X. Nicollin, A. Olivero, J. Sifakis, and S. Yovine. The algorithmic analysis of hybrid systems. *Theoretical Computer Science*, 138(1):3–34, 1995.

[AHS96] R. Alur, T.A. Henzinger, and E.D. Sontag. *Hybrid Systems III*, volume 1066. Springer Verlag, 1996. Lecture Notes in Computer Science.

[AKNS95] P. Antsaklis, W. Kohn, A. Nerode, and S. Sastry. *Hybrid Systems II*, volume 999. Springer Verlag, 1995. Lecture Notes in Computer Science.

[BD92] M. Broy and C. Dendorfer. Modelling of operating system structures by timed stream processing functions. *Journal of Functional Programming*, 2(1):1–21, 1992.

[BDD+93] M. Broy, F. Dederichs, C. Dendorfer, M. Fuchs, T. F. Gritzner, and R. Weber. The Design of Distributed Systems: An Introduction to Focus — Revised Version. Technical Report TUM-I9202-2, Technische Universität München, Fakultät für Informatik, 80290 München, Germany, 1993.

[Bro93] M. Broy. Interaction Refinement – The Easy Way. In M. Broy, editor, *Program Design Calculi*, volume 118 of *NATO ASI Series F: Computer and System Sciences*. Springer, 1993.

[Eng77] R. Engelking. *General Topology*. PWN - Polish Scientific Publishers, 1977.

[GNRR93] R.L. Grossman, A. Nerode, A.P. Ravn, and H. Rischel. *Hybrid Systems*, volume 736. Springer Verlag, 1993. Lecture Notes in Computer Science.

[GS96] R. Grosu and K. Stølen. A Model for Mobile Point-to-Point Dataflow Networks without Channel Sharing. In *Proc. of the 5th International Conference on Algebraic Methodology and Software Technology AMAST'96, Munich*, 1996. Also available as Technical Report TUM-I9527, Technische Universität München, `http://www4.informatik.tu-muenchen.de/~grosu/amast96.html`.

[Lam93] L. Lamport. Hybrid Systems in TLA+. In R.L. Grossman et al., editor, *[GNRR93]*, 1993.

[LSVW95] N. Lynch, R. Segala, F. Vaandrager, and H.B. Weinberg. Hybrid I/O automata. Technical Report CS-R9578, CWI, Computer Science Department, Amsterdam, 1995. To appear in: Hybrid Systems III. Available under `http://www.cs.kun.nl/~fvaan/`.

[MP93] Z. Manna and A. Pnueli. Verifying Hybrid Systems. In Grossman et al., editor, *[GNRR93]*, 1993.

[SG90] D. Scott and C. Gunter. Semantic Domains and Denotational Semantics. In *Handbook of Theoretical Computer Science*, chapter 12, pages 633 – 674. Elsevier Science Publisher, 1990.

[Sut75] W. A. Sutherland. *Introduction to metric and topological spaces.* Claredon Press - Oxford, 1975.

[Win93] G. Winskel. *The Formal Semantics of Programming Languages.* The MIT Press, 1993.