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Structure, behavior, and function of complex systems: The structure, behavior, and function modeling language

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Abstract

Teleological modeling is fundamental to understanding and explaining many complex systems, especially engineered systems. Research on engineering design and problem solving has developed several ontologies for expressing teleology, for example, functional representation, function–behavior–structure, and structure–behavior–function (SBF). In this paper, we view SBF as a programming language. SBF models of engineering systems have been used in several computer programs for automated design and problem solving. The SBF language captures the expressive power of the earlier programs and provides a basis for interactive construction of SBF models. We provide a precise specification of the SBF language. We also describe an interactive model construction tool called SBFAuthor.

Keywords: Behavior; Function; Ontology; Structure; Teleology

1. INTRODUCTION

Teleological modeling is fundamental to understanding and explaining many complex systems, especially engineered systems. By complex system, we mean a system that not only has many interconnected components but also has causal processes at multiple levels of abstraction such that a causal process at one level emerges out of component interactions at a lower level. By teleological model of a system, we mean a representation that specifies both the functions of the system and the causal processes that result in the system functions. It follows that a teleological model of a complex system would specify the system functions and the causal processes that result in them, at multiple abstraction levels.

Of course, teleological models are not necessarily useful for understanding or explaining all artifacts. We may view artifacts as lying on a spectrum of teleological explanation. At one end of this spectrum are artifacts for which teleological explanations are very useful. For example, the door system in my office has a useful teleological explanation: its functions (closing and opening) are accomplished by causal processes (pushing or pulling on the handle of the door, which creates a torque about the hinges that connect the door to the wall, which generates an angular motion about the hinges, and so on). Although the physical structure of the door (e.g., its connection to the wall) both affords and constrains its motion, its causal processes mediate between its structure and its functions. At the other extreme of the teleological spectrum are artifacts whose function directly emerges from the shape of its structural components. For example, the function of the desk on which my computer sits (providing a flat space for placing objects) directly emerges from the shape of its structure (a large flat top) without any mediation by a causal process. Many artifacts lie in the middle of this teleological spectrum, containing some subsystems that are teleological and others that are not. For example, the function of the chair that I am sitting on while I type these words directly emerges from the shape of its structure. In contrast, the chair also contains a lever so that when I push the lever down, it lowers the height of the seat; this function is accomplished by a causal process. In this paper, we are concerned solely with the ontology of complex systems in which causal processes mediate between their structure and functions.

Computational research on engineering design and problem solving has led to several ontologies for teleological

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modeling such as functional representation (Sembugamoorthy & Chandrasekaran, 1986; Chandrasekaran, 1994; Chandrasekaran & Josephson, 2000), function–behavior–structure (FBS; Gero, 1990; Gero et al., 1992), and function–behavior–state (Umeda et al., 1990, 1996; Umeda & Tomiyama, 1997). By *ontology*, we mean a vocabulary of knowledge representation (cf. Bunge, 1977), who views ontology as pertaining to the nature of the world). The structure–behavior–function (SBF) knowledge representation language (Bhatta & Goel, 1994, 1997; Goel et al., 1996; Prabhakar & Goel, 1998; Goel & Bhatta, 2004) provides one such ontology for teleological modeling.

Although SBF modeling is fairly mature by now (as is indicated by the years of publication in the list of references), recently our work has explored computational tools for interactive construction of SBF models, which raises several new research issues. The design literature typically described ontologies for teleological modeling only at a high level, often using teleological models of specific systems as illustrations. This is because a high-level specification of the ontology often is sufficient to explain a computational theory or model. However, a high-level specification of our SBF ontology is inadequate for developing an interactive tool for constructing an SBF model. This is because model building is an open-ended task, and the higher the level of specification of the ontology the less constrained the task becomes and the less guided is the human user. Users may potentially use the interactive tool to construct a variety of SBF models in a variety of ways, many of which might be incorrect or incomplete. Furthermore, the tool may need to incorporate model checking functionality to provide feedback to the users about the correctness of their models, creating a need for more precise specification of the ontology. Thus, for the purpose of developing an interactive tool, we need to view SBF as a programming language with a well-defined syntax and semantics. An SBF language would both constrain the actions of a user as well as afford the construction of useful SBF models. The language would also enable automated model checking. In addition, it would enable the construction of agents that both guide a user in constructing an SBF model as well as critique a constructed model. The primary goal of this paper is to specify the abstract syntax of the SBF language and to describe how the static semantics of the syntax enables interactive construction of SBF models.

2. AN ILLUSTRATIVE EXAMPLE OF SBF MODELS

To motivate the SBF programming language, let us consider a simple example: teleological modeling of a gyroscope follow-up, a device used in gyrocompasses on ships. (Briefly, the gyrocompass, with its ability to track true north compared to the unreliable magnetic north, is an instrument for navigation and piloting aboard many ships.) A gyroscope is an assembly with a very rapidly spinning top. A gyroscope follow-up automatically tracks and amplifies the movement of a spinning gyro. The follow-up servo can drive any number of gyrocompasses located anywhere on a ship, each of which replicates the reading of the central gyro. Figure 1 illustrates the structure of a simple gyroscope follow-up with no feedback control.

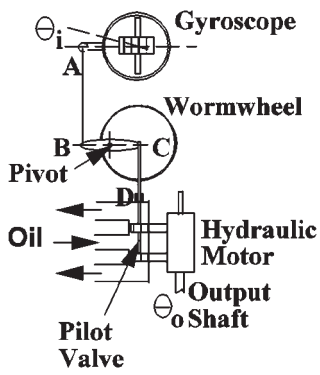


Fig. 1. A schematic diagram of the gyroscope follow-up.

2.1. Structure

In SBF models, structure is represented in terms of components, the substances contained in the components, and connections among the components. The specification of a component includes its functional abstractions, where a component can have multiple such functions. The specification of a substance includes its properties. Substances can be abstract, as, for example, angular momentum.

2.2. Function

A function is represented as a schema that specifies its preconditions and its postconditions. The function schema contains a reference to the behavior that accomplishes the function. This schema also may specify conditions under which the specified behavior achieves the given function

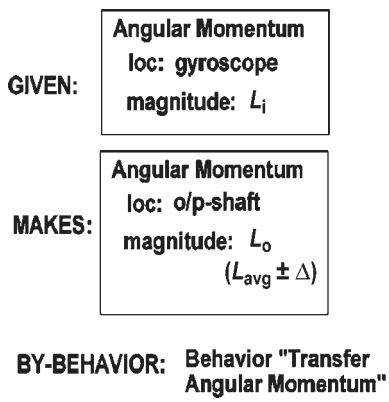


Fig. 2. A representation of function.

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(e.g., an external stimulus). Figure 2 illustrates the schema for the function of the gyroscope follow-up of Figure 1. Informally, the function specifies that the device takes as input angular momentum of magnitude L_i and of clockwise direction at the input (gyroscope) location and produces a proportional angular momentum of magnitude L_o and of clockwise direction at the output shaft location; L_o fluctuates over a large range, that is, $L_o = L_{avg} \pm \Delta$, where Δ can be large.

2.3. Behavior

A behavior is represented as a sequence of states and transitions between them. The states and the transitions are represented as state and transition schemas, respectively. The states in a behavior specify the evolution in the values of the parameters of substances and/or components. Continuous state variables are discretized, and temporal ordering is subsumed by causal ordering. Each state transition in a behavior is annotated by the causes for the transition. Causal explanations for state transitions may include physical laws, mathematical equations, functions of its subsystems, structural constraints, other behaviors, or a state or transition in another behavior. Figure 3 shows the behavior that explains how angular momentum from the input gyroscope location is transferred to the output shaft location. The functional context specified by the annotation `USING-FUNCTION` in Transition 3–4 indicates that the transition occurs because of the primitive function "`CREATE` Angular Momentum" of Hydraulic-Motor.

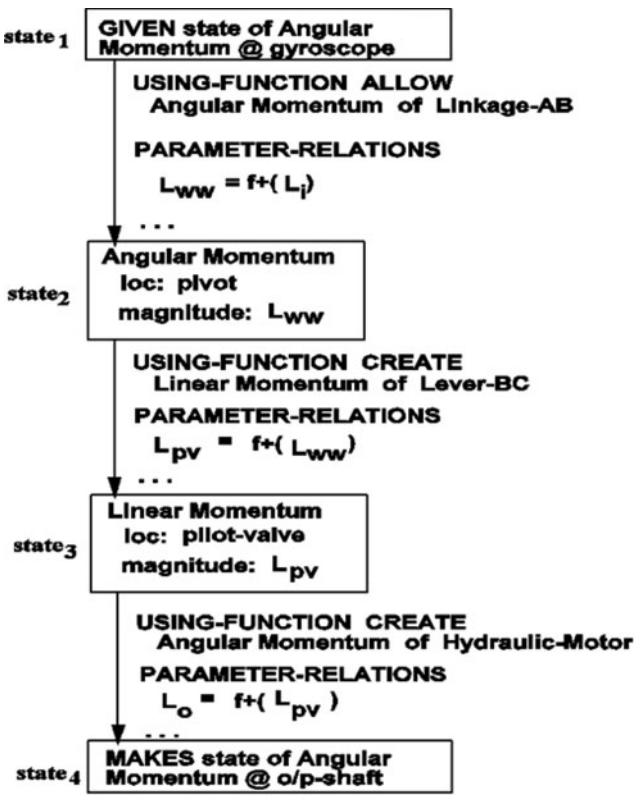


Fig. 3. A representation of behavior.

3. BACKGROUND OF SBF MODELS

Before we go further, we need to situate SBF modeling in the context of teleological modeling of complex systems in general. In *Sciences of the Artificial*, Simon (1996) viewed a system as having an inner and an outer environment. Further, he viewed a function of the system as an abstraction of its inner environment, lying at the interface of its inner and outer environments. Simon suggested that, from the perspective of understanding design of complex systems, the right kind of explanations specified how the function of a system arises out of the components and processes in the inner environment of the system.

In the early 1980s, Chandrasekaran and associates developed a functional representation (FR) scheme for representing teleology (Sembugamoorthy & Chandrasekaran, 1986; Chandrasekaran, 1994). They used the FR representation of a system for explanation and diagnosis. The FR representation of a system explains how its causal processes accomplish its functions and how the causal processes hierarchically compose the functions of the system's components into the functions of the systems as a whole. Thus, FR not only represented the functions of a system explicitly, but also used the functions as indices into the causal processes (called behaviors) that accomplished the functions; the behaviors, in turn, indexed the functions of the subsystems, and so on. At about the same time, Rasmussen developed an SBF scheme for representing teleology (Rasmussen, 1985; Rasmussen et al., 1994). Rasmussen and colleagues used the SBF representations of very large complex systems, such as an electrical power plant, to interactively aid human operators in troubleshooting the system. Their SBF representations emphasized representation of the functional roles of the structural components of the system.

In the late 1980s, Gero et al. (Gero, 1990; Gero et al., 1992) and Tomiyama et al. (Umeda et al., 1990, 1996; Umeda & Tomiyama, 1997) independently developed FBS schemes for representing teleology. (Tomiyama sometimes calls his representation scheme the FBS model.) Gero and Tomiyama used their respective FBS schemes for understanding the process of designing engineered systems and for aiding human designers in the design process. In the 1990s, Chakrabarti and Bligh (1996), Stone and Wood (2000), and Mizoguchi and his colleagues (Sasajima et al., 1995; Kitamura et al., 2004), among others, developed similar functional representation schemes and for similar uses. It is important to note that, as a consequence, there is not one but several functional representation schemes currently in the field. Although the various functional representation schemes have many features in common, their ontologies often are quite different. For example, behavior in Chandrasekaran's FR scheme refers

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to a causal process, whereas behavior in Gero's FBS representation pertains to the properties of a structural component. Erden et. al (2008) provide a recent useful review of several major functional representation schemes.

The origin of our SBF representation of teleology (Bhatta & Goel, 1994, 1997; Goel et al., 1996; Prabhakar & Goel, 1998; Goel & Bhatta, 2004) lies in Chandrasekaran's FR scheme (e.g., Goel & Chandrasekaran, 1989; Chandrasekaran et al., 1993). In particular, our SBF models both combine FR with Bylander's component–substance ontology and primitive functions (Bylander, 1991) and extend FR to support the inferences needed for automated design (Goel & Chandrasekaran 1989, 1992; Goel, 1992; Goel et al., 1997). SBF models share the main features of the FR scheme:

1. functions of devices are represented explicitly;
2. functions act as indices into internal causal behaviors responsible for them;
3. behaviors are represented as an ordered sequence of states;
4. state transitions in a behavior are annotated by the causal explanations for them;
5. the causal explanations can be of several types, for example, component function, structural relation, domain principle, and another behavior; and
6. the component function explanations for transitions act as indices into functions at the next (lower) level of aggregation.

SBF models also extend the FR scheme:

1. SBF models use a component–substance ontology of devices, which enables a more precise specification of states in a behavior or in a function;
2. SBF models use an ontology of primitive functions based on the component–substance ontology, which enables a more precise specification of state transitions in a behavior;
3. in SBF models, separate behaviors are constructed for substances and components, which makes for a more precise specification of behaviors as a whole;
4. the functions of systems are viewed as the intended subset of its output behaviors, and SBF models allow specification of all output behaviors of a device;
5. the internal causal behaviors in an SBF model may branch and merge; and
6. the internal causal behaviors admit inverse causality and bidirectional causality.

Although (4), (5), and (6) above enhance the expressive power of FR, (1), (2), and (4) afford more precise and accurate inferences needed for automated case-based design.

We have used SBF models in several design systems as briefly described below in Section 5. In addition, the functional representations in the NIST design repository (Szykman, Racz, et al., 2000; Szykman, Sriram, et al., 2000) have much in common with SBF models. Anthony et al. (2001) have developed a mark-up tool for interactive labeling of three-dimensional computer-aided design (CAD) models by structures, behaviors, and functions.

4. ABSTRACT SYNTAX OF THE SBF LANGUAGE

This section presents a top-down, syntax-oriented grammatical description of SBF. The notation used is a variant of Backus–Naur Form (BNF; International Organization for Standardization, 1996) in which syntactic definitions are described using production rules in which the term being defined appears on the left of a separator (`:=`), and its definition appears on the right as a sequence of terminal and nonterminal symbols. *Terminal symbols* denote categories of atomic words in an SBF model. Important categories in SBF are `STRINGS` and `INTEGERS`. A specific literal string may also appear in the grammar between straight apostrophes as in "`pump`." Other textual names denote *nonterminal* units that are defined in other rules in the grammar.

Several operators are used in the syntactic definitions. Juxtaposition denotes concatenation; "`|`" denotes alternative; "`[T]`" denotes optionality, where `T` is any string of symbols; "`{T}*`" and "`{T}+`" denote, respectively, any number of occurrences of `T` and any nonzero number of occurrences. Finally, "`// . . .`" denotes a comment that proceeds from the slashes to the end of the line.

Together, the following set of rules comprises an abstract syntax for SBF. The syntactic description is abstract because it avoids concrete details such as punctuation and keywords.

4.1. SBF model

At the highest level, an SBF specification looks like the following:

```
SBFModel := STRING                  // Model name
            [STRING]                // Description
            StructureModel
            FunctionModel
            BehaviorModel
            {Stimulus}+
```

That is, an SBF specification comprises six parts, appearing consecutively: a name, an optional description, specific submodels for structure, function, and behavior, and one or more external stimuli to which the system being modeled might react. These latter four constituents are now described in corresponding subsections.

4.2. Structure model

```
StructureModel := {Element}+
                  {Connection}*
```

A StructureModel is merely one or more `Elements` and the `Connections` among them.

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```
Element := INTEGER                    // Element Id
           STRING                     // Element name
           [STRING]                   // Description
           {Property}*
           (Component | Substance)
           {INTEGER}*                 // Subelement Ids
```

An `Element` consists of a unique identifier followed by a name, an optional description, and a list of `Properties`. An `Element` is either a physical `Component` or a `Substance`. Moreover, an `Element` can itself have subcomponents, leading to a hierarchical structural breakdown. Several comments are in order. First, `Elements`, like many other of the nonterminals to be described below, have unique identifiers. This enables dependencies among the Structure, Behavior, and Function submodels to be described. Second, `Properties` are used throughout SBF models to denote named, typed values. Here is the syntactic definition of a Property.

```
Component := INTEGER                  // Function Id
             {ConnectingPoint}*
ConnectingPoint := INTEGER            // ConnectingPoint Id
                   STRING             // ConnectingPoint name
Substance :=
```

`Properties` have `Types` in the sense of belong to a set of related values, such as `integers` or `strings`. Moreover, because SBF models physical devices, it supports the use of scientific `Units` to further describe the `Values`. The nonterminal `Constantp` is intended to enable the modeler to indicate that the `Value` of a particular `Property` is not subject to change.

```
Property := STRING                    // Property name
            Type
            Value
            Unit
            Constantp

Type := 'boolean' | 'integer' | 'real' | 'string' | ...
Value := STRING
Unit := 'gram' | 'meter' | 'second' | ...
Constantp := 'true' | 'false'
```

`Components` are `Elements` that can be connected with other `Components`. They should be distinguished from `Substances`, which are used to model fluids and forces.

```
Connection := INTEGER                 // Connection Id
              STRING                  // Name of connection
              Mechanism
              INTEGER                 // Id of first connected Component
              INTEGER                 // First ConnectingPoint Id
              INTEGER                 // Id of second connected Component
              INTEGER                 // Second ConnectingPoint Id
Mechanism := 'parallel' | 'series' | 'touching' | 'adjoining'
           | 'bolted' | 'fused' | 'hinged' | 'jointed'
           | 'tied' | 'telescoped' | 'threaded'
           | 'frictionallyEmbedded' |'sewn' |'nailed'
           | 'clipped' | 'ballAndSocket' | 'glued'
```

`Connections` are binary, associating named `ConnectingPoints` in `Components`. Moreover, `Connections` are partitioned into categories based on the way in which force is transferred between the corresponding `Components`.

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4.3. Behavior model

Behavior is modeled in SBF with deterministic finite state machines. In fact, a `Behavior` is nothing more than a set of `States` and related `Transitions`.

```
BehaviorModel := {Behavior}+
Behavior := INTEGER              // BehaviorId
            STRING               // Behavior name
            [STRING]             // Description
            {State}+
            {Transition}+

State := INTEGER                 // StateId
         STRING                  // State name
         [STRING]                // Description
         {Content}+
Content := INTEGER               // ElementId
           {Property}*
           ConnectingPointId
```

`States` constrain the values of `Properties` at `ConnectingPoints`.

```
Transition := INTEGER                // TransitionId
              STRING                 // Transition name
              [STRING]               // Description
              INTEGER                // SourceStateId
              INTEGER                // TargetStateId
              {CausalExplanation}*
CasualExplanation := INTEGER         // FunctionId
                   | INTEGER         // StateId in another Behavior
                   | INTEGER         // TransitionId
                   | INTEGER         // Structure connection
                   | INTEGER         // StimulusId
                   | STRING          // Domain principle
                   | INTEGER         // BehaviorId
```

`Transitions` are directed binary associations between `States`. Moreover, each `Transition` might have a number of `CausalExplanations` motivating the change of `State`.

4.4. Function model

`Functions` in SBF describe the role that an `Element` plays in the overall operation of a device. They express the purpose or goal of the `Element`, whereas the `Behavior` describes how the purpose is accomplished. Each `Element` in an `SBFModel` has a `Function`, and each `Function` has a corresponding `Behavior`.

```
FunctionModel := {Function}+
Function := INTEGER                  // FunctionId
            Name                     // Function name
            [STRING]                 // Description
            FunctionType
            [INTEGER]                // PrimaryArgument (ElementId)
            INTEGER                  // BehaviorId
            STRING                   // Provided
            INTEGER                  // Prerequisite StateId
            {INTEGER}+               // Resultant StateIds
            {INTEGER}*               // StimulusIds
Name := Primitive
      | NonPrimitive
Primitive := 'create' | 'destroy' | 'expel' | 'allow' | 'pump'
           | 'move'
NonPrimitive := STRING
```

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The SBF ontology comprises a number of primitive Functions useful for describing devices. Moreover, modelers can define new (NonPrimitive) ones.

```
FunctionType := 'achievement' | 'maintenance' | 'prevention'
              |'negation'
```

The SBF ontology also distinguishes categories of Functions, partitioned according to how the corresponding Behavior affects the Element's State.

4.5. Stimuli

The final constituent of an SBFModel describes the environmental Stimuli that can affect its Behavior.

```
Stimulus := INTEGER          // StimulusId
            STRING           // Stimulus name
            [STRING]         // Description
            [Property]
```

A Stimulus may have an associated Typed Value, describing its amplitude.

5. STATIC SEMANTICS OF THE SBF SPECIFICATION LANGUAGE

The abstract syntax given above is useful for describing the appearance of SBF models, but not how to interpret them. In particular, some syntactically correct models may be incomplete or have inconsistencies. Here we present a series of rules that distinguish valid from invalid SBFModels.

5.1. Uniqueness constraints

The first category of rules merely specifies that the identifiers used to label different model constituents are themselves different; for example, each ElementId in Element is unique. Additional rules in this category include the following:

- There is only one (directed) Transition between any two States in the same Behavior.
- No two CausalExplanations on any Transition may be the same.
- No NonPrimitive-Function Name can be the same as any Primitive-Function Name.
- For any given Connection, the FirstConnectingPointId must differ from the SecondConnectingPointId.

5.2. Referential integrity constraints

Another category of model rules has to do with *referential integrity*. This term refers to situations in which an identifier in one place in a model refers to (is identical with) an identifier in another place. In general, there should be no missing targets to any reference.

- Each StartStateId and StopStateId in Function correspond to actual States existing in the Behavior referred to by BehaviorId in that same Function.
- Each ElementId in Content refers to a Component that has a Function with the Behavior containing the State that has that Content.
- Each ConnectingPointId in Content refers to a ConnectingPoint in a Component that has a Function with the Behavior containing the State that has that Content.
- Each SourceStateId and TargetStateId in Transition refers to States that exist in the Behavior containing the Transition.

5.3. Organizational rules

Several other rules are necessary to further constrain valid SBFModels.

- A StructureModel contains exactly one Element that is not referred to by a SubElementId. That Element must be a Component and not a Substance. That is, the device itself being modeled is considered as the outermost Element.
- Each SubElementId in any Element refers only to an Element below itself in the containment hierarchy. That is, the Element containment hierarchy is a strict tree structure.
- Each FunctionId in the CausalExplanation of a Transition in a Behavior corresponds to a Function that does not have that Behavior.
- Each StateId referred to in a CausalExplanation of a Transition in a Behavior corresponds to a State in a different Behavior.
- Each TransitionId referred to in a CausalExplanation of a Transition in a Behavior corresponds to a Transition in a different Behavior.

6. DYNAMIC SEMANTICS OF THE SBF SPECIFICATION LANGUAGE

SBF models of complex systems enable computer programs to draw inferences about the systems. The inference rules for a particular program provide an interpretation of an SBF model. The set of allowed interpretations for a program

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provide a specific dynamic semantics to the SBF language. Here are several examples.

6.1. Kritik

Kritik (Goel & Chandrasekaran 1989, 1992; Goel, 1992; Goel et al., 1997) was an early automated design system that integrated case-based and model-based reasoning for generating conceptual designs of simple engineering devices. It indexed known designs by their functions and used SBF models of the known designs to guide the process of design adaptation. Given the functional specification of a new problem, Kritik retrieved designs that delivered similar functions, viewed a retrieved design as having failed to achieve the given desired function, and used the SBF model of the design to diagnose the causes for the failure and propose modifications to the design. Thus, the SBF models in Kritik enabled both design retrieval and design adaptation.

6.2. Ideal

Ideal (Bhatta & Goel, 1994, 1997; Goel & Bhatta, 2004) was an early automated design system that used cross-domain analogies for generating designs of simple engineering systems. Like Kritik, it indexed known designs by their functions and used SBF models of known designs to guide the process of design adaptation. In addition, it used behavior–function (BF) representations of abstract teleological mechanisms (such as "feedback") for cross-domain transfer. Given the functional specification of a new problem, Ideal retrieved a design that delivered a similar function, viewed the retrieved design as having failed to achieve the given desired function, used the SBF model of the design to diagnose the causes for the failure, and proposed modifications to the design. If and when this adaptation process failed, however, Ideal abstracted the functional specification of the problem, retrieved an abstract teleological solution relevant to the abstract functional problem, and instantiated the mechanism in the context of the design retrieved earlier. It also learned abstract teleological mechanisms from SBF models of known designs, Thus, SBF models in Ideal enabled abstraction, transfer, and instantiation of abstract teleological mechanisms.

6.3. Torque

Torque (Griffith et al., 2000) modeled verbal protocols (Clement, 1989) of physicists addressing problems pertaining to spring systems. In Torque, retrieval of an analog was based on structural similarity instead of functional similarity. The SBF model of the retrieved analog enabled the transfer of qualitative mathematical knowledge across domains (from the domain of bending beams to that of spring systems).

6.4. Archytas

Archytas (Yaner & Goel, 2007, 2008) is a recent automated system that constructs SBF models from design drawings by analogical transfer of SBF models of similar drawings. Given a target design drawing, and given also a known design drawing and its SBF model, Archytas maps, transfers, and adapts the SBF model of the known drawing to the target drawing. To do this, Archytas extends SBF models into DSSBF models (for drawing–shape–structure–behavior–function), thereby incorporating shapes and drawings into SBF models. In Archytas, analogical mappings at one level of abstraction (e.g., shape) enable transfer at the next higher level (e.g., structure).

7. FROM AUTOMATED SYSTEMS TO INTERACTIVE ENVIRONMENTS

Kritik, Ideal, Torque, and Archytas are automated systems. InteractiveKritik (Goel et al., 1996; Goel & Murdock, 1996) enabled a human to browse the designs in Kritik's design library, the SBF models of the designs Kritik generated, and Kritik's processing in generating a design. InteractiveKritik, however, did not allow the user to interactively construct SBF models.

Interactive model construction environments provide external representations, which can be powerful tools to support designing and learning. Clement (2000) has argued that learning is fundamentally a process of model construction and revision. Buckley (2000) has shown that interactive environments can help students learn about models of complex systems at multiple levels of organization. Hmelo-Silver et al. (2000) have suggested that although experts model a complex system in terms of its interrelated structure, behaviors, and functions, novices primarily express its structure, demonstrate limited understanding of its functions, and largely miss its behaviors. Hmelo-Silver and Pfeffer (2004) have recommended use of interactive environments that enable construction of SBF models to support learning about complex systems. We have constructed an interactive SBF model construction tool, called SBFAuthor, which will be used in several interactive environments. Here are a few examples.

7.1. ACT

SBF models are being used as the foundation for the Aquarium Construction Toolkit (ACT), an interactive learning environment for modeling aquaria as a complex system (Hmelo-Silver et al., 2008). The idea is that middle school students may be better able to understand a complex system such as an aquarium if they build a model of the system that includes not only its structural descriptions, but also its functions and behaviors. SBFAuthor is the interactive tool that will be used to construct these models. ACT also enables the students to simulate their models using NetLogo (Wilensky, 1999). Hence, SBF models must be executable by NetLogo, and NetLogo provides an operational semantics for SBF. Thus far, we have been able to build useful interpretations for a subclass of SBF models, and we are actively working on others.

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7.2. DANTE

SBF models are also the foundation for an interactive environment, called DANTE, for supporting biologically inspired design (Vattam et al., 2008). The idea is that representations of biological systems such as SBF models may facilitate analogical transfer of biological principles to engineering problems. DANTE provides another semantic interpretation of SBF models. It does this with a feature called a *critic*. Critics are reactive checkers of a user's models. Consequently, DANTE must treat SBF models as potentially being incorrect. The semantics of an incorrect model is the appropriate response (diagnosis and suggestion) from the relevant critic to the modeler.

8. SBFAuthor: AN INTERACTIVE SBF MODEL CONSTRUCTION TOOL

SBFAuthor is a computational tool for interactive construction of SBF models of complex systems. SBFAuthor adds a visual syntax for the SBF specification language described above. The utility of visual notation for modeling languages has been shown in practice through visual modeling paradigms such as the entity relationship (ER) model and the Unified Modeling Language (UML). Visual representations offer significant advantages over textual representations for model building and model comprehension. For example, Nosek and Roth (1990) noted that semantic networks (graphs) are more understandable than their equivalent predicate logic (textual) representation. Particular knowledge representation frameworks such as concept maps (Novak & Gowin, 1984) are based on well-defined graphical notations.

SBFAuthor is an editor that facilitates building SBF models using visual notations. Figure 4 captures the correspondence between the elements of the abstract syntax of SBF specification language and their visual counterparts.

SBFAuthor visually partitions an SBF model into three views: Structure view, Behavior view, and Function view, as shown in Figure 5. This figure depicts the SBF model of the Gyroscope follow-up shown in Figure 1.

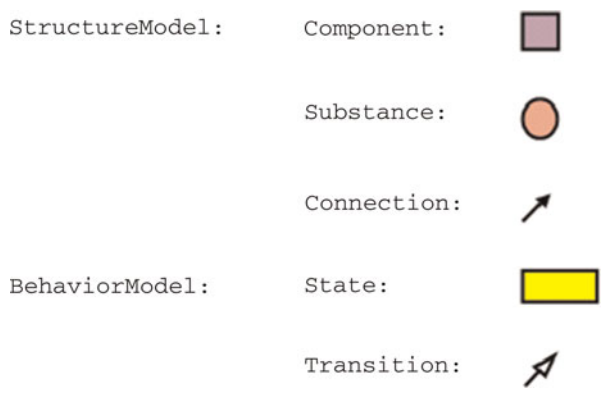


Fig. 4. The visual syntax of the SBFAuthor. [A color version of this figure can be viewed online at journals.cambridge.org/aie]

8.1. Structure view

The structure view enables users to create the `StructureModel` portion of the SBF model in terms of `Components`, `Substances` and their associated `Connections`. The Structure model is presented as a graph. For each `Component` or `Substance`, a corresponding node is created. The `Connections` are represented as labeled links between nodes in the graph. Figure 6a shows the structure view of the gyroscope follow-up presented in Figure 1. It consists of `Component` nodes like "`Gyroscope`," "`Worm Wheel`," "`Pivot`," and so forth. It also contains a `Substance` node "`Angular momentum`." Links between these nodes indicate specific kinds of `Connections` like "`contains`," "`connected`," and so forth.

`Components` and `Substances` can be described using dialog boxes (see Fig. 6b) that can be invoked to enter their feature values (e.g., "`name`," "`description`") and to include their `Properties` (e.g., "`location`" and "`magnitude`" in the case of "`Angular Momentum`"). According to the SBF specification, a `Component` can itself comprise a subsystem with its own SBF model. In such situations users can create a separate SBF model for that `Component` and include a reference to the `Function` of that model in the parent model.

8.2. Behavior view

The behavior view allows users to create the `BehaviorModel` portion of the SBF model by allowing them to create one or more `Behaviors`, which appear as different tabs in the behavior view. For each `Behavior`, users can create `States` and `Transitions`. Each `Behavior` is represented as a state-transition graph as shown in Figure 7a. Every `State` and `Transition` is associated with a dialog box. Recall from the SBF specification that `States` can constrain the values of `Component/Substance Properties`. The `State` dialog box allows the user to model `State` variables and their values by choosing them from a list of `Properties` derived from the `StructureModel`. For example, Figure 7b represents "`State 1`" of the behavior model constructed in Figure 3. The variables "`Location`" and "`Magnitude`" of "`Angular Momentum`" are assigned values "`Gyroscope`" and "`M_input`," respectively. These variables are derived from the `StructureModel`.

Recall that `Transitions` capture the `CausalExplanations` in the form of references to various model elements (like `Functions`, `Behaviors`, etc.). The `Transition` dialog box enables a user to express this information. For instance, in Figure 7c the `Transition` named "`USING FUNCTION ALLOW Angular Momentum of Linkage-AB`" contains a reference to the `Function` "`ALLOW Angular Momentum`." This captures one of the causes for the `Transition` between "`State 1`" and "`State 2`," reflecting the `Function` of a particular linkage to allow angular momentum to move from the gyroscope to the pivot.

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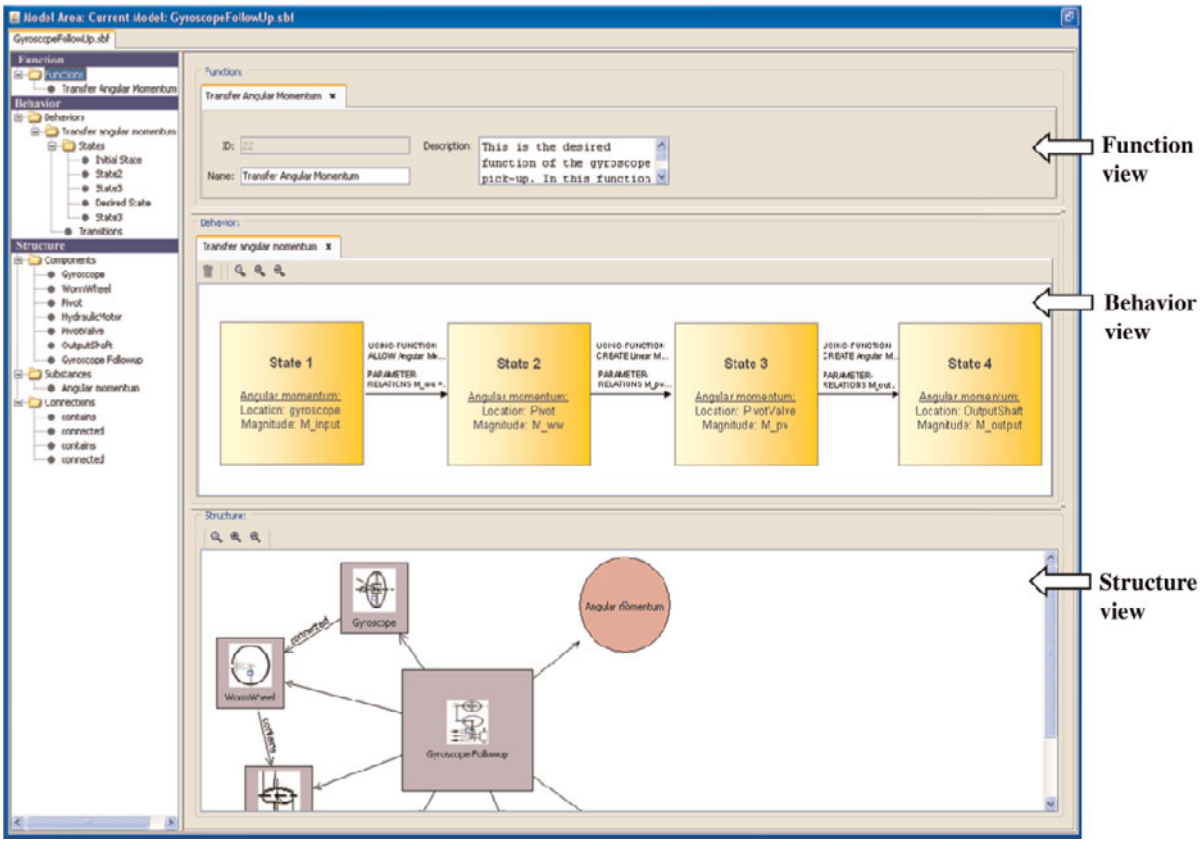


Fig. 5. A snapshot of SBFAuthor's main interface. [A color version of this figure can be viewed online at journals.cambridge.org/aie]

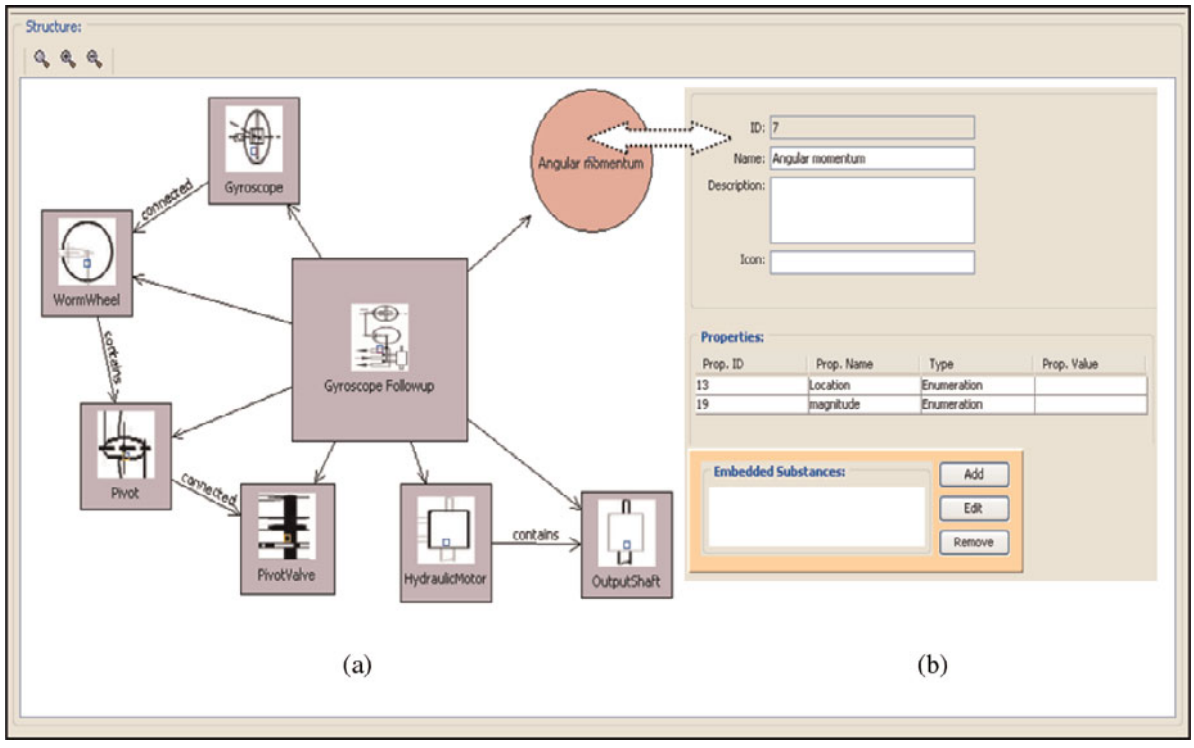


Fig. 6. The structure view: (a) a structure model of a gyroscope follow-up and (b) a substance dialog box. [A color version of this figure can be viewed online at journals.cambridge.org/aie]

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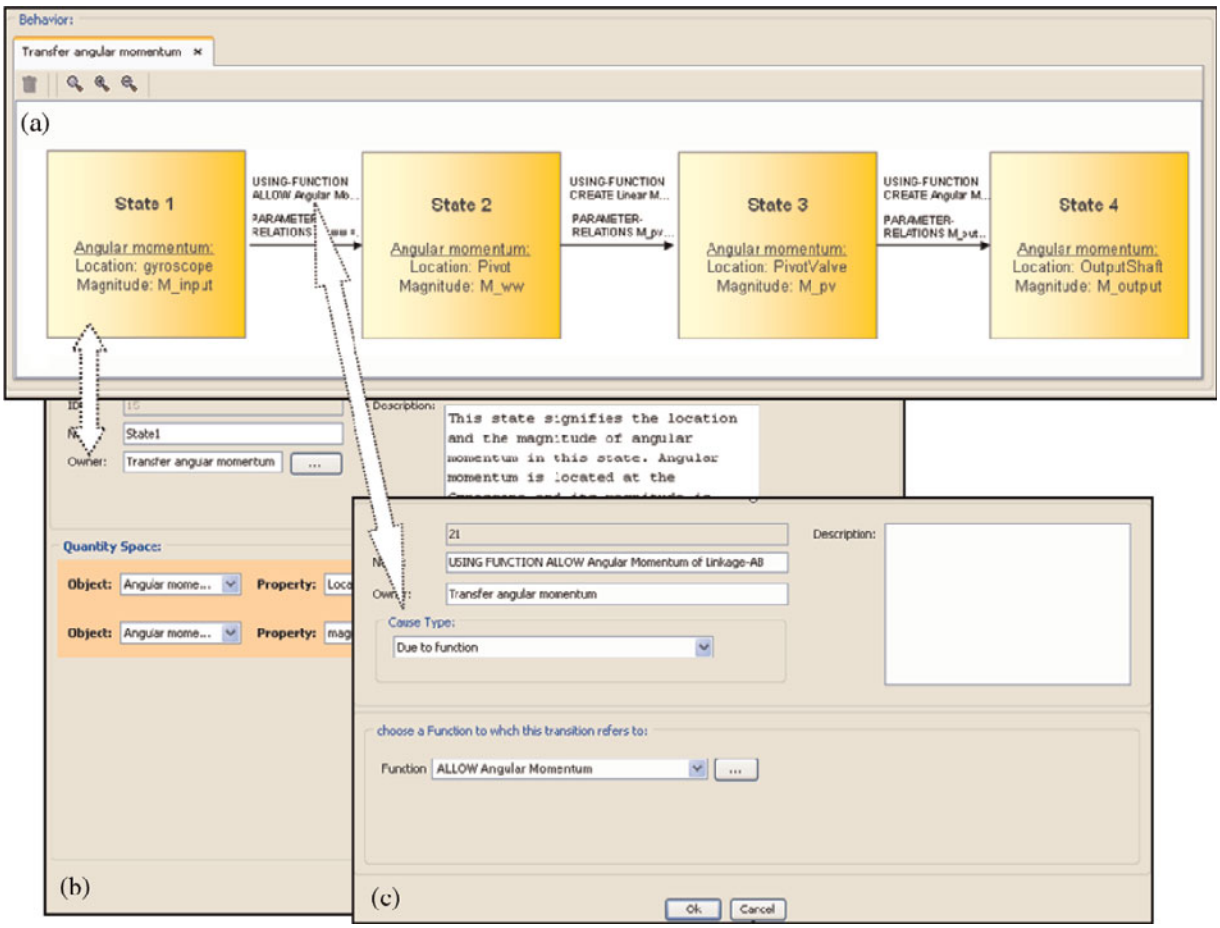


Fig. 7. The behavior view: (a) a specific behavior and (b) state and (c) transition dialog boxes. [A color version of this figure can be viewed online at journals.cambridge.org/aie]

8.3. Function view

The function view allows users to create the `Function-Model` portion of the SBF model as one or more `Functions` that appear as different tabs in the function view as shown in Figure 8. For each `Function` users can state its type (`Primitive` or `Nonprimitive`). If the function being modeled is a `Primitive Function`, the user can choose from an existing set of `Primitive Functions` that are present in the teleology. If it is a `Nonprimitive Function`, the user has to define the new `Function` and include a reference to a `Behavior` that accomplishes it. In Figure 8, for example, the `Function` "`Transfer Angular Momentum`" is accomplished by the `Behavior` "`Transfer Angular Momentum Behavior`." In addition to the reference to a `Behavior`, a user has to specify



Fig. 8. The function view. [A color version of this figure can be viewed online at journals.cambridge.org/aie]

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a StartState labeled "Initial State" (e.g., "State 1" in Fig. 8) and a StopState called "Desired State" (e.g., "State 4" in Fig. 8) associated with that Function, which the user can specify by either choosing States existing in the BehaviorModel or by creating new ones. In accordance with the SBF specification, users can also include a reference in the function view to an external Stimulus that initiates that Function.

9. CONCLUSIONS

Research on engineering design and problem solving has developed several ontologies for expressing teleology. The literature typically describes these ontologies only at a high level, often using teleological models of specific systems as illustrations. However, the development of interactive tools for constructing teleological models of complex systems requires a more precise specification of the ontology. Giving a formal definition to SBF leads to a variety of benefits including precision, consistent usage, and tool building. Each application of the SBF language requires not only the abstract syntax and static semantics but also a concrete syntax and a dynamic semantics. The precise specification potentially enables a range of additional automated capabilities such as model checking, model simulation, and interactive guides and critics for model construction. In this paper, we viewed SBF as a programming language. We specified the abstract syntax and the static semantics of the SBF language. The next step in our work is to develop the precise dynamic semantics of the language; in this paper, we merely pointed to earlier work that illustrates the dynamic semantics. The SBF language captures the expressive power of the earlier programs and provides a basis for interactive construction of SBF models. We also described an interactive model construction tool called SBFAuthor that is based on the abstract syntax and static semantics of the SBF language.

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