Types of Models

From SEBoK
Types of Models

There are many different types of models and associated expressions in a diverse array of modeling languages and tool sets. This article offers a taxonomy of model types and highlights how different models must work together to support broader efforts.

Contents

1 Model Classification
   1.1 Formal versus Informal Models
   1.2 Physical Models versus Abstract Models
   1.3 Descriptive Models
   1.4 Analytical Models
   1.5 Hybrid Descriptive and Analytical Models
   1.6 Domain-specific Models
   1.7 System Models
   1.8 Simulation versus Model
   1.9 Visualization
2 Integration of Models
3 References
   3.1 Works Cited
   3.2 Primary References
   3.3 Additional References

Model Classification

There are many different types of models and associated expressions to address different aspects of a system and different types of systems. Since different models serve different purposes, a classification of models can be useful for selecting the right type of model for the intended purpose and.

Formal versus Informal Models

Since a model is a representation of a system, many different expressions that vary in degrees of formalism could be considered models. In particular, one could draw a picture of a system and consider it a model. Similarly, one could write a description of a system in text, and refer to that as a model. Both examples are representations of a system. However, unless there is some meaning to the terms, there is a potential lack of precision and the possibility of ambiguity in the representation.

The primary focus of system modeling is to use models supported by a well-defined modeling language. While less formal representations can be useful, a model must meet certain expectations for it to be considered within the scope of. In particular, the initial classification distinguishes between informal and formal models as supported by a modeling language with a defined syntax and
Physical Models versus Abstract Models

The United States “Department of Defense Modeling and Simulation (M&S) Glossary” asserts that “a model can be [a] physical, mathematical, or otherwise logical representation of a system” (1998). This definition provides a starting point for a high level model classification. A model is a concrete representation that is distinguished from the mathematical and logical models, both of which are more abstract representations of the system. The physical model can be further classified as descriptive (similar to logical) or analytical (similar to mathematical). Some example models are shown in Figure 1.

Descriptive Models

A descriptive model describes logical relationships, such as the system's whole-part relationship that defines its parts tree, the interconnection between its parts, the behavior that its parts perform, or the test cases that are used to verify the system. Typical descriptive models may include those that describe the functional or physical aspects of a system, or the three dimensional geometric representation of a system.

Analytical Models

An analytical model describes mathematical relationships, such as differential equations that support quantifiable analysis about the system parameters. Analytical models can be further classified into dynamic and static models. Dynamic models describe the time-varying state of a system, whereas static models perform computations that do not represent the time-varying state of a system. A dynamic model may represent the performance of a system, such as the aircraft position, velocity, acceleration, and fuel consumption over time. A static model may represent the mass properties estimate or
prediction of a system or component.

**Hybrid Descriptive and Analytical Models**

A particular model may include descriptive and analytical aspects as described above, but models may favor one aspect or the other. The logical relationships of a descriptive model can also be analyzed, and inferences can be made to reason about the system. Nevertheless, logical analysis provides different insights than a quantitative analysis of system parameters.

**Domain-specific Models**

Both descriptive and analytical models can be further classified according to the domain that they represent. The following classifications are partially derived from the presentation on OWL, Ontologies and SysML Profiles: Knowledge Representation and Modeling (Web Ontology Language (OWL) & Systems Modeling Language (SysML)) (Jenkins 2010):

- properties of the system, such as performance, reliability, mass properties, power, structural, or thermal models;
- and technology implementations, such as electrical, mechanical, and design models;
- subsystems and , such as communications, fault management, or power distribution models; and
- system applications, such as information systems, automotive systems, aerospace systems, or medical device models.

The model classification, terminology and approach is often adapted to a particular application domain. For example, when modeling or , the model may be referred to as workflow or model, and the performance modeling may refer to the and schedule performance associated with the organization or business process.

A single model may include multiple domain categories from the above list. For example, a reliability, thermal, and/or power model may be defined for an electrical design of a communications subsystem for an aerospace system, such as an aircraft or satellite.

**System Models**

System models can be hybrid models that are both descriptive and analytical. They often span several modeling domains that must be to ensure a consistent and system representation. As such, the system model must provide both general-purpose system constructs and domain-specific constructs that are shared across modeling domains. A system model may comprise multiple views to support planning, requirements, design, analysis, and .

Wayne Wymore is credited with one of the early efforts to formally define a system model using a mathematical framework in A Mathematical Theory of Systems Engineering: The Elements (Wymore 1967). Wymore established a rigorous mathematical framework for designing systems in a model-based context. A summary of his work can be found in A Survey of Model-Based Systems Engineering (MBSE) Methodologies.

**Simulation versus Model**

The term , or more specifically, refers to a method for implementing a model over time (DoD 1998). The computer simulation includes the analytical model which is represented in executable code, the conditions and other input data, and the computing infrastructure. The computing infrastructure includes the computational engine needed to execute the model, as well as input and devices. The great variety of approaches to computer simulation is apparent from the choices that the designer of computer simulation must make, which include

- stochastic or deterministic;
- steady-state or dynamic;
• continuous or discrete; and
• local or distributed.

Other classifications of a simulation may depend on the type of model that is being simulated. One example is an agent-based simulation that simulates the interaction among autonomous agents to predict behavior (Barry 2009). They are many other types of models that could be used to further classify simulations. In general, simulations provide a means for analyzing complex dynamic behavior of systems, software, hardware, people, and physical phenomena.

Simulations are often integrated with the actual hardware, software, and operators of the system to evaluate how actual components and users of the system perform in a simulated environment. Within the United States defense community, it is common to refer to simulations as live, virtual, or constructive, where live simulation refers to live operators operating real systems, virtual simulation refers to live operators operating simulated systems, and constructive simulations refers to simulated operators operating with simulated systems. The virtual and constructive simulations may also include actual system hardware and software in the loop as well as stimulus from a real systems environment.

In addition to representing the system and its environment, the simulation must provide efficient computational methods for solving the equations. Simulations may be required to operate in real time, particularly if there is an operator in the loop. Other simulations may be required to operate much faster than real time and perform thousands of simulation runs to provide statistically valid simulation results. Several computational and other simulation methods are described in Simulation Modeling and Analysis (Law 2007).

Visualization

Computer simulation results and other analytical results often need to be processed so they can be presented to the users in a meaningful way. Visualization techniques and tools are used to display the results in various visual forms, such as a simple plot of the state of the system versus time to display a parametric relationship. Another example of this occurs when the input and output values from several simulation executions are displayed on a response surface showing the sensitivity of the output to the input. Additional statistical analysis of the results may be performed to provide probability distributions for selected parameter values. Animation is often used to provide a virtual representation of the system and its dynamic behavior. For example, animation can display an aircraft’s three-dimensional position and orientation as a function of time, as well as project the aircraft’s path on the surface of the Earth as represented by detailed terrain maps.

Integration of Models

Many different types of models may be developed as artifacts of a MBSE effort. Many other domain-specific models are created for component design and analysis. The different descriptive and analytical models must be integrated in order to fully realize the benefits of a model-based approach. The role of MBSE as the models integrate across multiple domains is a primary theme in the International Council on Systems Engineering (INCOSE) INCOSE Systems Engineering Vision 2020 (INCOSE 2007).

As an example, system models can be used to specify the components of the system. The descriptive model of the system architecture may be used to identify and partition the components of the system and define their interconnection or other relationships. Analytical models for performance, physical, and other quality characteristics, such as reliability, may be employed to determine the required values for specific component properties to satisfy the system requirements. An that represents the interaction of the system components may be used to validate that the component requirements can satisfy the system behavioral requirements. The descriptive, analytical, and executable system model each represent different facets of the same system.

The component designs must satisfy the component requirements that are specified by the system models. As a result, the component design and analysis models must have some level of to ensure
that the design model is traceable to the requirements model. The different design disciplines for electrical, mechanical, and software each create their own models representing different facets of the same system. It is evident that the different models must be sufficiently integrated to ensure a cohesive system solution.

To support the integration, the models must establish to ensure that a construct in one model has the same meaning as a corresponding construct in another model. This information must also be exchanged between modeling tools.

One approach to semantic interoperability is to use between different models. Transformations are defined which establish correspondence between the concepts in one model and the concepts in another. In addition to establishing correspondence, the tools must have a means to exchange the model data and share the transformation information. There are multiple means for exchanging data between tools, including file exchange, use of application program interfaces (API), and a shared repository.

The use of modeling standards for modeling languages, model transformations, and data exchange is an important enabler of integration across modeling domains.

References

Works Cited


Primary References


Additional References


