Challenges in Semantics for Computer-Aided Designs: A Position Paper

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Abstract
This paper presents a brief summary of a number of different approaches to the semantic representation and automated interpretation of engineering data. In this context, engineering data is represented as Computer-Aided Design (CAD) files, 3D models or assemblies. Representing and reasoning about these objects is a highly interdisciplinary problem, requiring techniques that can handle the complex interactions and data types that occur in the engineering domain. This paper presents several examples, taken from different problem areas that have occupied engineering and computer science researchers over the past 15 years. Many of the issues raised by these problems remain open, and the experience of past efforts can serve to identify fertile opportunities for investigation today.

Introduction
The relationships among shape and form, structure and function, and behavior and semantics are among the most fundamental questions studied by science and engineering—and it is precisely these relationships that must be captured and preserved by digital engineering artifacts. Engineering artifacts, such as buildings, aerospace and automotive products, and consumer goods are nearly ubiquitously represented in digital form as 3D Computer-Aided Design (CAD) models in commercial CAD systems. In spite of the fact that nearly every object we use in modern life has its genesis as a digital representation in a CAD system, we understand surprisingly little about the explicit semantics of these objects. As a result, even after over 40 years of development of digital engineering tools, their use remains largely as a proxy for paper drawings and the semantic content of the objects they create remains locked in the cognitive processes of engineering team.

This paper presents a brief introduction to five separate, but highly related, problems in representation and capture of the semantics of engineering artifacts produced by Computer-Aided Design. Specifically,

- Content-based retrieval of 3D CAD models based on their manufacturing processes, functional attributes and shape semantics.
- Representation and indexing of the structure, function and behavior of CAD assemblies (entire products or electromechanical systems).
- The representation and extraction of design process and rationale.
- Representation of digital engineering design information for long-term archival.

Examples from the authors’ body of work is provided in each case to illustrate some of the fundamental problems and to provide a context for describing the current open challenges in the semantic representation and automated interpretation of 3D Computer-Aided Design representations of engineering data.

Figure 1: Examples of 3D Computer-Aided Designs for a variety of discrete mechanical parts.
Some Challenge Areas for Engineering Semantics

Automated Feature Extraction

Feature recognition is a common algorithm in a wide variety of pattern recognition applications (i.e., vision, machining learning, etc). In the context of Computer-Aided Design and Computer-Aided Manufacturing (CAD/CAM), feature recognition provides the communication medium between design data and manufacturing applications. Features also form a basic component of tools for design analysis and feedback, as well as for systems that automatically generate process plans and drive manufacturing processes.

What is particularly devilish for engineering is that there are a multitude of different feature representations depending on what point one it at in the design lifecycle. For example, a designer may work with design features (shown in Figure 2 (a.a)). Design features typically are operations on the 3D models to create shape attributes with functional or behavioral properties. In the example, a “socket” part has “holes” that serve as mating features to attach the part to another part in an assembly model. The holes are modeled geometrically as a subtractive operation or a swept contour; they may have associated with them attributes such as manufacturing tolerances (i.e., cocentricity or circularity) that describe how tightly the final artifact must match the mathematical shape.

In contrast, manufacturing features are specific to the manufacturing process to be used to create the artifact. In the case of the “socket”, the object is to be machined from a cylindrical block of metal (Figure 2 (a.c)) and the “machining features” correspond to the metal cutting operations needed to remove the material in the delta-volume (Figure 2 (a.d)). These features are shown in Figure 2 (b) and are vastly different from those used in design. For example, machining features must take into account the geometry of the cutting tool, as well as the robotic motion that the machining center is capable of. These features correspond to the possible volumes swept by a rotating cutting tool undergoing 2.5 axis motion. For the socket, there are 22 possible machining features—these can be used to produce up to 512 possible process plans (i.e., sequences of cutting operations). While only a few of these process plans make manufacturing sense, the interpretation of the shape in terms of all of its possible manufacturing steps is a key challenge in developing automated manufacturing planning tools.

The Semantics Problems. Recognizing features from a design is the means of providing a level of design understanding to manufacturing software systems. Beyond just manufacturing features, the problem of feature recognition is really one of mapping between different interpretations of a shape in different contexts. For example, feature recognition is a key algorithmic element in CAD translation software—in which CAD data is transformed between different file formats and internal representations. Unlike a word processing document (whose principal internal structure is some form of text markup), CAD representations are often generative. A generative representation uses a set of features to create a procedural definition of a shape. Further complicating matters, each CAD system uses a different set of design features and operations, thus the generative representations vary widely across commercial systems.

Many researchers have looked into these problems. They are a key part of efforts in the ISO 10303 STEP standard. However the pace of change in the commercial world greatly outstrips standard’s progress and we find ourselves perpetually in a situation where substantial human effort is required to extract the design intention from one model and map it into alternative representations. There currently exist no common ontologies of features, nor are there any suitable “corpa” for use in developing feature taxonomies.

Content-based Retrieval of CAD Objects

Since the early 1970s, researchers have been interested in automating the indexing and retrieval of engineering objects stored in databases. There are two basic types of approach for matching and retrieval of 3D CAD data. (1)
feature-based techniques and (2) shape-based techniques. The feature-based techniques are the oldest. They are based on extracting engineering features e.g., machining features, and form features, from a solid model of a mechanical part for use in database storage, and automated coding. These techniques often work off of the native CAD or solid model. These models are mathematically precise, topologically watertight, and algorithmically and numerically complex. For example, solid model might consist of a data structure describing the boundary of a 3D model in terms of a set of NURBS patches. Interrogating and reasoning about the information in this model requires understanding the semantics of the underlying geometric and topological representations.

The shape-based techniques are more recent, owing to research contributions from computational geometry, computer vision, and computer graphics. A shape-based approach works as the representational “lowest common denominator”: polygon mesh available from faceting solid models, in the form of VRML or STL. From the polygon mesh, different transformation invariant attributed can be extracted as the means of similarity among 3D models. In contrast to the CAD and solid model techniques, the semantics of VRML and STL and other mesh formats is trivial. However, along with this simplicity comes a severe loss of the design knowledge present in the native geometry as well as loss in the accuracy of the shape models themselves.

Shape matching techniques are generally robust under model degradation, but it is a rigid technique and is a poor discriminator among model classes, because it usually emphasizes gross model shape, rather than the discriminatory features that are common in CAD/CAM data.

The Semantics Problems. In the CAD/CAM domain, engineering artifacts can have multiple classifications. For example, discrete machined parts can be classified in to different categories according to different classification criteria, such a functionality (e.g., brackets or fasteners), manufacturing cost and manufacturing process (e.g., casting, machining, forging, molding). Figure 3 (a) shows four CAD models under two different, but perfectly reasonably, classification schemas. The first classification is based on the manufacturing process, where parts are separated into either “3-axis machining” or “casting” processes. In machining, rotating cutting tools remove material based on swept volumes; these sweeps are limited to those on a 3-axis vertical machining center. The second, orthogonal classification, is based on mechanical function. Figure 3 (a) also shows a decomposition into parts that function as “brackets” or as “housings”.

Representation of Structure, Function and Behavior

In order to tackle the many abstract concepts and underlying structures involved, effective reasoning on mechanical assemblies requires knowledge-based mechanisms for representing the design intent for the assembly. Assemblies are primarily defined by their intended function—goals achieved and tasks performed—and the decomposition of that function into sub-functions. These in turn define form, structure, and behavior. It is therefore the representation and reasoning of function to which this work has been initially scoped. Reasoning at this abstract, design rationale-oriented level is best accomplished through the application of knowledge representation and reasoning techniques. Doing so under a framework with clear semantics, desirable for auto-

Figure 3: An example of CAD artifacts (a) with different manufacturing interpretations and (b) a shape-based discriminator for distinguishing among these manufacturing processes.
The example shown in Figure 4 gives a representation based on function and flow—the materials, energies, and signals on which functions operate. The representation is broken up into two parts: a core ontology, defining the basic structure of a function-based mechanism description language, and vocabulary extensions, providing terminologies with which rich descriptions can be written in the language. The core ontology defines the universe of objects as consisting of assemblies, components, functions, and flows. Several relations are also given, for example to associate assemblies with functions and with their input and output flows. The vocabulary extensions provide taxonomies of functions and flows derived from those developed in prior work by the authors and colleagues at NIST.

Beyond the representation of structure, function and behavior is design rationale. Design rationale is an explanation of why an artifact, or some part of it, is designed the way it is. Design rationale includes all the background knowledge such as deliberating, reasoning, trade-off and decision-making in the design process of an artifact—information that can be valuable, even critical, to various people who deal with the artifact. The research has ranged from basic observations about the design process to different approaches to capturing design rationale.

The Semantics Problems. Current representations for function and behavior are still largely ad hoc. It is very difficult to have a representation that rigorously and unambiguously captures the semantics of a mechanism. Complicating matters is the fact that there can be multiple interpretations of the function and behavior. In some ways this is not unlike the problems faced in the intelligence, machine vision or plan recognition domain, where various low-level events can be consistently described in several ways. Current approaches are to provide languages expressive enough to describe and distinguish devices while maintaining efficiency and computability. It is neither so formal as to prevent practical computing, nor so informal as to prohibit automated reasoning. The following section outlines the use of such reasoning in a design repository.

A second challenge is the integration of these languages into existing commercial design and manufacturing software systems and workflows. Current CAD/CAM tools are still predominantly geometry-centric. While various research and even commercial systems have attempted to create “knowledge-based CAD systems”, these have focused usually on formalizing design rules and less on the representation and capture of design rationale. Automating the generation of these representations is also difficult.

Long-Term Archiving of Digital Engineering Artifacts

Perhaps the most significant open problem is that of extraction of engineering semantics in order to ensure long-term sustainability of engineering artifacts. For many modern industries, engineering design and manufacturing knowledge needs to be preserved over 50-to-75 year lifespans. Tradi-
tional digital data management for these organizations are usually dependent on the proprietary formats of commercial software systems and cannot guarantee the readability and utility of data over long periods. While 3D computer-aided design (CAD) modeling has become an indispensable aspect of modern engineering, the engineering part print (i.e., blueprint or 2-D drawing on paper, aperture cards, microfiche) remains as the principal method of design knowledge archival. From an archival standpoint, much of the knowledge generated during the modern engineering enterprise (i.e., by 3D CAD, simulation, etc.) is simply lost in this process. Even considering data that are archived, over a long enough product lifetime, the data files and supporting infrastructure required to access CAD product designs will be obsolete and unusable. Some artifacts (i.e., airframes, ships, bridges and other civil infrastructure) have lifetimes that extend not just over changes in CAD software packages, but across the development of CAD and computing technology itself.

One might hold out hope that these problems would be addressed by either industry or the international standards community. International standards such as STEP, ISO 103033, are certainly an integral aspect of any solution for digital preservation. However, the current STEP standard does not cover the full breadth of problems needed to ensure long-term interpretability of engineering knowledge. Issues like organizational workflow, information provenance, model evolution over time and relationships among model representations (that may not be in STEP) are all critical to ensure that future users of this preserved engineering knowledge can understand the data they have access to.

**Example: Drexel’s Bio-Robotic Snake.** Consider the capture the semantics of a complex design and analysis task for simulation of the biologically-inspired (snake-like) robot shown in Figure 5(a). In this example, one of the principal challenges is how to capture and represent the simulation/analysis workflow, shown in Figure 5(c).

**The Semantics Problems.** Archiving introduces many new problems in semantics. For example, long-term interpretability is not often a design criteria for knowledge representation structures. Specific to engineering, however, the workflows, processes and internal complexities of engineering objects all require new ideas for representation. Provenance is particularly troublesome in an engineering context: models are often passed around an organization, with many individuals creating derivative models specific for the needs of their task in the workflows. The result is a family of related models, all of which describe the same artifact. Hence, to really understand the artifact (i.e., for diagnostic or forensic purposes) one really needs to have access to this network of files and understand their relationships.

![Figure 5: CIBER-U Case Study: A Bio-Inspired Robot.](image)

![Figure 6: Two challenges with engineering objects: the diversity of related file elements and their internal representational complexities.](image)

**Discussion and Open Issues**

The engineering domain is a rich source of a semantics problems. Across all of the examples above, the complexity and diversity of engineering data is shown to pose unique
challenges. For CAD and 3D data in the engineering process, many derivative model representations must be created for the various workflows central to design, manufacturing, and lifecycle activities. For example, some geometric models contain data that specifies shape, but the shape can be specified in different ways—each with different representational power, precision and end purpose. Shown in Figure 6, Vertex-based Model data, Curve Models, Wireframe Models, Surface Models, and Solid Models may all exist in the engineering workflow. Today, the vast majority of objects produced by modern CAD software are 3D models or solid models. One of the main the derivative models from 3D solid models include meshes, (polyhedral approximations of 3D objects, usually used in graphics for rendering). However, the semantic content of a 3D mesh is far lower than that of the 3D solid model as present in a proprietary CAD system (i.e., Pro/Engineer). Understanding these trade-offs, and how to represent and capture these semantics, is largely a open problem.

Another factor contributing to the complexity of this domain: Engineering artifacts each have a physical realization. CAD models usually represent a manufacturable object (i.e., physical part, building, assembly, etc). Existing approaches to shape semantics are usually designed for recognition and classification—rarely venturing into the representations required to understand how to manufacture or assemble the structure described. For example, a part is machinable on a 3-axis machining center or it is not; a part has four symmetrically spaced holes for fastening with bolts or it does not. This is in contrast to problems in the domain of information retrieval, where datasets are pre-classified based mostly on human intuition (i.e., boats get grouped with boats; airplanes with airplanes).

Conclusions

This paper has presented a number of problems related to the representation of semantics in the context of engineering artifacts and CAD/CAM data in particular. The engineering domain is of unique global importance, as it documents our infrastructure, consumer products and defense and healthcare systems. Given its importance, it is surprising that the domain has not seen more sustained and large-scale efforts to comprehensively address the issue of semantics. In many ways the engineering domain shares considerable similarity with eScience: as engineering activity migrates to exist nearly entirely on a digital substrate, semantic representations are needed to understand and capture the interactions among data, software and human elements. These problems are likely to be perpetual, as with each subsequent generation there are new technologies, design concepts and products we require new structures to capture and represent this knowledge.

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