

Ontological Support for Teaching Strategy in Intelligent Visual Reasoning Tutor

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Abstract. IVRT is an ITS for visual reasoning, using the missing view problem. It combines an ITS framework with a solid modeling kernel that supports hint-generating rules using geometric reasoning. We develop an ontology for IVRT's hint generation rules, and a separate ontology for IVRT's teaching strategy. Teaching strategy rules are stored in a custom text format, with compilation to Jess.

The ability to visualize and reason about geometric aspects of 3D objects is critical for success in many disciplines in engineering and architecture. The missing view problem [1], shown in Figure 1, is typically used in visual reasoning instruction. Two consistent, principal orthographic views are given, and the learner must provide the third view corresponding to a valid 3D solid object.

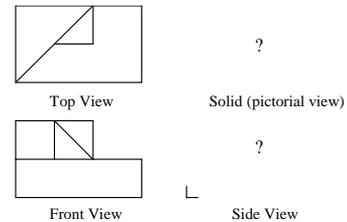


Figure 1. Missing view problem

We have developed an instructional software system called Intelligent Visual Reasoning Tutor (IVRT). IVRT's missing view problem solving module [2][3] implements a solid modeling kernel that can represent and display a partially-constructed 3D solid. It provides interactive sweeping operations [4] to incrementally construct faces of a solution solid.

The missing view module's hint generation knowledge is implemented as a set of rules in CLIPS [5], shown in Table 1. We develop an ontology to represent the knowledge behind these rules, shown in Figure 2, and instantiate a model of these rules, shown in Figure 3. This model explicitly represents the kinds of knowledge used by each rule, and identifies the major programming patterns involved in their implementation.

1. Create visible faces first (top and front faces).
2. Create hidden faces next (bottom and back faces).
3. Create faces using <i>construct</i> command last (right faces, then left faces).
4. Prefer the face that is adjacent to the most correct faces.
5. Prefer the face with the most incident edges.
6. Prefer the face whose normal vector contains the most zero components.

Table 1. Hint-generating rules

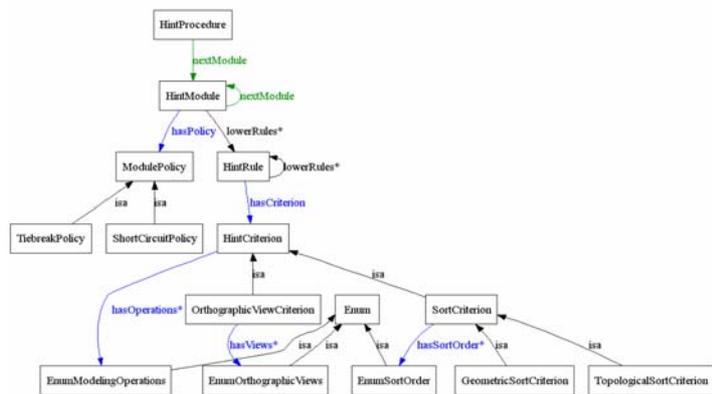


Figure 2. Ontology classes for hint-generating rules

