A specification language for consistent model generation based on partial models

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ABSTRACT Automated graph generation has become a key component in many testing and benchmarking scenarios. For example, modeling tool qualification can be effectively supported by the direct synthesis of well-formed graph models as test inputs, systematic testing of cyber-physical systems requires different test environment models, and different optimization and design-space exploration approaches require the best models with respect to an objective function.

In this paper, we propose a novel specification language for partial models used in consistent graph model generation. The language includes constructs to uniformly capture initial, intermediate and final results of the generation by combining partial models, graph predicates and model metrics with mutual dependencies between them. The formal semantics of the language is defined by using 4-valued Belnap-Dunn logic that explicitly marks inconsistent model elements as part of the partial model. The use of our language is illustrated in the context of a complex case study defined by NASA researchers.

KEYWORDS Partial models, Model generation, 4-valued logic.

1. Introduction

Quality assurance of critical software-intensive systems frequently relies on the automated synthesis of test data to reduce conceptual gaps in the test cases. When testing domain-specific modeling tools, or autonomous cyber-physical systems in model-based systems and software engineering scenarios, test data takes the form of typed and attributed graph models. Automated model generators are key technologies to address the needs of such testing scenarios.

A model generator needs to derive consistent models where each model needs to satisfy (or deliberately violate) a set of constraints captured in the form of OCL constraints or graph predicates. Logic solvers (like SMT-solvers, SAT-solvers, CSP-solvers) have been frequently providing precise foundations for such model generators in tools like Alloy, USE, UMLtoCSP, Formula, etc. However, a recently emerging family of model generators, like PLEDGE (Soltana et al. 2020) or the VIATRA Solver (Semeráth et al. 2018), addresses the consistent model generation challenge directly on the level of graph models by sophisticated search strategies (like multi-objective optimization or SAT-solving algorithms) and powerful abstractions provided by 3-valued partial models and partial model refinement (Varró et al. 2018). To fine-tune the model generation process, iterative and incremental approaches (Semeráth et al. 2016) are proposed where models obtained as output in a previous run can be used as inputs (e.g. required or forbidden model fragments) in subsequent runs.

The vast majority of existing model generators adapted popular industrial languages and technologies (like EMF, OCL) as their input specification. On the other hand, custom domain-specific specification languages like Alloy have also become popular due to its precise semantics. However, when genera-
ing models in an iterative and incremental way along partial model refinement (Varró et al. 2018; Semeráth et al. 2018), this enforces a black-box view on model generation where the input and output models are concrete instance models, while all intermediate steps of the generation operate on abstract partial models as candidate solutions (some of which cannot be represented as regular instance models in EMF). Similarly, complex rewriting techniques are necessitated to approximately evaluate constraints over partial models (Semeráth & Varró 2017). Furthermore, the development of novel algorithms became very complicated due to the conceptual mismatch between concrete models and partial models.

The main objective of the current paper is to provide foundations for a more grey-box view of model generation by providing a high-level language for specifying model generation problems. The main purpose of this language is three-fold.

1. On the one hand, it should be modeling technology independent to capture graph-based model generation problems that arise for non-EMF models (e.g. OWL ontologies, graph databases, GraphQL and JSON data).

2. On the conceptual level, it should enable to capture partial models as first-class citizens as inputs, outputs, and the intermediate state of model generation. As such, any intermediate result of any model generation step can be used as-is for a subsequent model generation run, thus the internal states of model generation can become transparent by serializing the partial models in that format.

3. Finally, we wish to provide a mathematically precise intermediate language for describing well-formedness constraints and model metrics, again in a technology-independent way. For that purpose, we propose a combined notation influenced by logic programming and functional programming.

The main contribution of the paper is to propose a specification language for model generation problems that uses partial models, graph predicates and metrics as core language elements. A precise semantics of this core language is defined by a combination of 4-valued logic (for structural domain elements) and abstract interpretation over intervals (for attributes). Moreover, we also define some higher-level language elements as syntactic sugars which are semantically mapped back to core elements. We introduce this language on a sequence of examples for a model generation problem introduced by researchers from the NASA JPL in Herzig et al. (2017).

2. Case study

The design synthesis of interferometry mission (IM) architectures has been introduced in (Herzig et al. 2017) as a complex challenge for early mission planning for space missions of NASA where a target architecture consists of collaborating satellites (of different size and capabilities) and radio communication links between them. Each mission architecture contains multiple spacecrafts that impose a challenging design task. Constraints and mission objectives used in the current paper were defined by Herzig et al. (2017).

A metamodel for interferometry constellation missions is shown in Figure 1 in an EMF notation. An InterferometryMission consists of communicating CommElements, which are equipped with CommSubsystem subsystems (i.e. antennas with different communication frequencies) via their commSubsystem references for KaComm, UHFComm and XComm bands. Spacecraft of different sizes, including cube satellites CubeSat3U and CubeSat6U, as well as small satellites SmallSat, may carry interferometry Payloads (photos sensors), and must be able to reach the GroundStationNetwork via radio links (to send interferometry sensor data) denoted by the target references.

The constraints of a consistent interferometry mission architecture capture that

- each Spacecraft may only have a single transmitting commSubsystem (the other commSubsystem, if present, may only receive);
- all Spacecraft must have a communication path to the GroundStationNetwork;
- there must not be any communications loops (communication paths from a CommunicatingElement to itself);
- CommunicatingElement instances may only communicate if they share the same radio band (KaComm, XComm or UHFComm) and frequency; and
- CommSubsystem instances of certain radio bands require specific Satellite types (e.g. only SmallSat instances can use the KaComm band).

A set of objective functions is defined in order to quantitatively assess each IM architecture. Such objectives include

- the cost of equipment in the constellation, including Spacecraft, CommSubsystems and Payloads, which is subject to economies of scale regarding multiple pieces of equipment of the same type;
- the duration of the mission, which is characterized by the observation time and the time required to downlink the gathered images to the GroundStationNetwork;
- the coverage of the observation area achieved by the interferometry Payloads during the observation time.
3. Syntax of partial models

The goal of partial models is to explicitly represent uncertainty in models by introducing a 4-valued logic representation where a single partial model covers a set of concrete instance models. Informally, a specific structural model element (i.e., a domain object or link) can surely exist (a truth value of true), cannot exist (a truth value of false), or it may be uncertain if the model element exists (specified or unknown). The latter is signified by the unknown truth value (Kleene et al. 1952) in accordance with Reps et al. (2004); Varró et al. (2018); Semeráth & Varró (2017). If a model element is inconsistent, then a special error value is assigned to it (Marussy et al. 2018). Moreover, data objects can be used as attribute values to represent quantitative information. The core constructs of our specification language enable capturing such partial models.

Informally, our specification language allows to define graph-based partial models by using assertions where an assertion may state what is true, what is false or what is unknown related to the model. Well-formedness constraints of a domain can be defined by predicates in a Prolog-style language. Furthermore, graph metrics (e.g., objective functions, model metrics) can also be defined in combination with predicates in the style of functional programming languages (e.g., Erlang or ML).

3.1. Core syntax for partial models

3.1.1. Language literals

The partial modeling language includes the following literals: <id> defines identifiers in the model, <integer> and <real> defines numbers, and <string> defines strings of characters with escape symbol. In general, this follows a notation similar to programming languages like Java.

\[
(id\text{-}fragment) ::= (a \cdots z | A \cdots Z | \_ )
\]

\[
(id) ::= (id\text{-}fragment) (::(id\text{-}fragment)) *
\]

\[
(integer) ::= (-)? (0 \cdots 9)+
\]

\[
(real) ::= (-)? (0 \cdots 9)+ (. (0 \cdots 9)+)?
\]

Example 3.1. Spacecraft and Spacecraft::new are identifiers, 3000 is an integer and 2.71828 is a real number.

3.1.2. Objects: domain/data, named/unnamed

A partial model describes partial models with objects and relations between them. Objects include primitive data objects (e.g. integer and real numbers) or non-primitive domain objects. In either case, an object is either named or unnamed. A named object can be differentiated from the other named objects by using an identifier, while unnamed objects do not have designated identities. As such, replacing a named object with another object results in a different partial model, while replacing an unnamed object with another (otherwise identical) unnamed object will not change the partial model.

\[
(object\text{-}id) ::= <\text{named-obj}\text{-}id> | <\text{unnamed-obj}\text{-}id>
\]

\[
(named\text{-}obj\text{-}id) ::= \text{char}+*
\]

\[
(unnamed\text{-}obj\text{-}id) ::= <id>
\]

Figure 2 Graphical representation of a partial model.

Example 3.2. Figure 2 shows a partial model using graphical notation. Types (unary predicates) are written inside the boxes for the objects (? stands for uncertain types). Edges (binary predicates) are depicted as arrows (dashed arrows stand for uncertain edges and thick arrows stand for containment relations). The special ~ edge denotes object equality.

The named object 'gsc' represents a special antenna from a GroundStation named 'gs', and o3 is an unnamed object representing another antenna. Here, a named object is used for 'gsc' to differentiate its role from other antennas. The partial model lists three possible CommunicationType constants (based on the enumerated type in Figure 1) with named objects 'KaComm', 'UHFComm', and 'XComm'. Here, we used named objects for all enumeration constants as we want to keep the identity of each type (i.e. it matters whether 'gsc' has type 'KaComm' or 'UHFComm'). Figure 2 also depicts two data objects i1 and i2 that represent in the model (describing frequencies). Finally, the object CommSubsystem::new by itself represents all the potential new CommSubsystem objects and int::new represents all the integer data objects that may be added to the model.

3.1.3. Relation assertions

Using the core syntax, a partial model is defined with relation and value assertions over the objects. Relation assertion defines a 4-valued truth-value (one of true, false, unknown and error) for an n-ary relation for a tuple of objects. Besides the usual truth values true and false, unknown represents that the given information is not specified in the model (thus it can be either true or false in generated models), while error represent inconsistent information (i.e. true and false at the same time). From a model generation perspective, a partial model with any unknown values means that the model is not yet finished (i.e. it is not yet a regular concrete model), while any error value makes the model inconsistent, thus no consistent instance models can be generated by model refinement from the partial model.
\[ \text{relation assertion} \] = \{ \text{ground-rel-use} : \{ \text{logic-value} \} \].
\[ \text{ground-rel-use} \] = \{ \text{relation-id} (\langle (\text{ground-term}(\text{, } (\text{ground-term}))\rangle)?) \}. 
\[ \text{relation atom with ground arguments} \]
\[ \text{partial model node} \]
\[ \text{object-id} \]
\[ \text{truth value} \]
\[ \text{4-valued logic values} \]

Combined with named and unnamed objects, unary relations are sufficiently expressive to uniformly represent simple node labels, more complex type systems, enumerated types, and primitive types (like integers).

**Example 3.3.** The following code excerpt shows some unary relation assertions from the partial model in Figure 2.

\begin{verbatim}
CommSubsystem('gsc'): true.
Spacecraft('gsc'): false. % ...
CubeSat3U(o3): true.
CubeSat(o3): true.
Spacecraft(o3): true.
CommunicatingElement(o3): true.
CommType('KaComm'): true.
CommType('UHFComm'): true.
CommType('XComm'): true.
int(i1): true. int(i2): true.
int(i3): true.
int::new: true.
\end{verbatim}

Lines 1–3 of the example denote that it is true that the object 'gsc' has the type CommSubsystem, but it does not have other types like Spacecraft and Payload. Complex type systems can be formalized by listing the combination of all type predicates. For example, the types of the object o3 are listed in lines 4–7. An enumerated type can be formalized by listing all instances (like in lines 8–10), and specifying that all other objects does not have that type. Finally, primitive data objects are also represented by type predicates, like in lines 11–12.

Binary relation assertions can represent a wide range of structures like references (links, edges) and attributes.

**Example 3.4.** In Figure 2, communication paths between the satellites are represented by the \text{target} relation.

\begin{verbatim}
\text{target} ('gsc', o3): true.
\text{target} (o3, 'gsc'): unknown.
\text{target} ('gsc', 'gsc'): false.
\end{verbatim}

In the partial model, it is known that object 'gsc' is communicating with object o3 (denoted with solid edges in Figure 2 and \text{target} ('gsc', o3): true in line 1). Potential communication relations are denoted with \text{unknown} values, e.g. in line 2, the \text{target} link between o3 and 'gsc' is \text{unknown}. Finally, a partial model can represent the absence of a relation with the truth value false, e.g. in line 3.

**3.1.4. Exists and equals** A partial model uses two special relations: the unary relation \text{exists} assigns a truth-value for the existence of an object, while the binary relation \text{equals} assigns a truth-value for the equivalence of two objects. These special relations can also take \text{unknown} truth values to represent abstract graph structures.

**Example 3.5.** The following lines illustrate different combinations of \text{exists} and \text{equals} from Figure 2.

\begin{verbatim}
\text{exists}(o3): true. \text{equals}(o3, o3): true.
\text{equals}(o1, o3): false. \text{equals}(o2, o3): false.
\text{exists} (CommSubsystem::new): unknown.
\text{equals} (CommSubsystem::new, CommSubsystem::new): unknown.
\text{equals}(i1, i2): unknown. \text{equals}(i2, i3): unknown.
\text{equals}(i1, i3): unknown.
\end{verbatim}

Lines 1–2 show a concrete object o3 which is existing, and it is equal to itself and different from others. Next, lines 3–4 show a multi-object CommSubsystem::new that may or may not exist (i.e. 0.. multiplicity) and its unknown equivalence with itself denotes that the object may represent multiple different objects (i.e. ..* multiplicity). Finally, lines 5–6 represent the uncertain equivalence of the three integers i1, i2 and i3. These data objects may possibly be merged with each other.

**3.1.5. Value assertions** In a concrete model, each data object (e.g. integer) has a concrete value defined by a single assignment. However, in a partial model, the value of a data object is defined by one or more closed interval assertions, where an interval can possibly be empty or infinite (\text{inf}). If several of such assertions exist, then the intersection (conjunction) of such intervals is taken.

\begin{verbatim}
\text{value assertion} \]
\langle \text{assertion} \] ::= ⋯ | \{ \text{ground-term} \}: \{ \text{interval} \}. 
\[ \text{interval} \] ::= \{ (\text{lower-bound}), (\text{upper-bound}) \}. 
\[ \text{non-empty interval literal} \]
\[ \text{empty interval literal} \]
\[ \text{finite or infinite lower bound} \]
\[ \text{finite or infinite upper bound} \]
\end{verbatim}

**Example 3.6.** The partial model in Figure 2 illustrates three \text{int} objects. The value of i1 is set to 3000, which is defined by the interval [3000, 3000]. For i2 we define only that the value is greater than or equal to 3000, thus we use interval [3000, +\text{inf}]. Finally, by setting an interval of \text{int::new} we define the potential range of all the new integers.

\begin{verbatim}
i1: [3000, 3000]. i2: [3000, +\text{inf}].
\text{int::new}: [-\text{inf}, +\text{inf}].
\end{verbatim}

**3.2. Syntactic sugar for partial models**

Since partial models can be defined by the users as an initial input (seed) model, we also provide a simplified notation with syntactic shortcuts for the most frequently used constructs.

**3.2.1. Simplified assertions** First, relation assertions can be simplified with the prefixes ? and !, where ? stands for the truth-value \text{unknown}, ! stands for the truth-value \text{false} and the lack of a prefix stands for \text{true}. Shortcuts can also be introduced for data objects by referring to them with their value. As a result, relations in partial models can be defined by Prolog-style clauses while still keeping the option of using many-valued logic and abstract domains.
4. Graph predicates and metrics

Building on the core notions for describing partial models from section 3, in this section, we will introduce graph predicates and metrics. Graph predicates capture derived relations and well-formedness constraints, while metrics allow numerical computations, including constraints on numerical values, as well as objective functions such as model size and cost.

4.1. Graph predicates

We partition relation symbols (i.e. \( \text{relation-id} \)) into base relation symbols and defined predicate symbols.

The base relations include built-in relations, such as \text{int} and \text{real}, as well as any types, references and attributes that comprise the model. Any relation symbol appearing in the partial model without an associated predicate definition is considered to be a base relation symbol. The truth values of these relations can be specified arbitrarily by assertions.

In contrast, one may create defined predicate symbols by predicate definitions. The body of the definition may be evaluated on the model to compute a logic values. Thus assertions involving defined predicate symbols are constraints on the computed logic values. In the following, we will discuss the syntax of predicate definitions.

### 4.1.1. Predicate definitions

Predicate definitions specify queries (or constraints) as Datalog-like expressions over relation symbols. The name and parameters of the predicate are separated by the \( := \) symbol from the body of the predicate definition. The body is specified in a disjunctive normal form. Literals in the body of predicate definition may refer to \( (n\text{-ary}) \) relation symbols, including both defined (interpreted) predicate symbols and base (uninterpreted) relation symbols, metric checking literals (to check whether the value of a metric lies within an interval using the \text{in} operator), the transitive closures of binary relations, as well as negations thereof. However, as a specific limitation, recursive definitions (when a predicate definition refers to itself directly or indirectly via relations and metric use) are disallowed.

When using relations (and metrics) in predicate definitions, only variables (e.g. \( x \)) are allowed as arguments. Variables in a body of a predicate that do not appear in the parameter list of the predicate definition are implicitly considered to be existentially quantified.

### 3.2.2. Any and default

The complete definition of a partial model needs to assert truth values for all relations. This would require a large number of assertions, which is impractical. Therefore, we introduce two mechanisms to assert truth-values with the assumption that everything unspecified is unknown (i.e. open-world semantics). For example, we can avoid explicitly enumerating the types an object does not have by setting the default logic value for each type to false (i.e. closed-world semantics). Conversely, setting default to error means every possible value of the relation must be explicitly enumerated, even unknown values.

- \((\text{ground-term}) := \cdots | \ast \quad \triangleright \text{all partial model nodes}\)
- \((\text{default-assertion}) := \text{default} (\text{assertion})\)

Table 2 shows the translations for these constructs.

Example 3.8. By defining the default value of \text{CommSubsystem} as false in line 1, we may omit the statement \text{CommSubsystem}(01): false. However, logic values differing from false must be state explicitly.
predicates can serve as conditions in aggregation expressions (e.g. to count the number of matches of a predicate) and in conditions.

4.1.3. Type annotations In addition to the core syntax above, several language constructs help define predicates as syntactic sugar. Firstly, unary predicates may appear as type annotations in parameter. For each type annotation, a call to the predicate used as type is added to each disjunctive case of the predicate body, which results in a notation reminiscent of typed object-oriented programming languages.

\[
\text{(predicate-def)} ::= \text{(pred-def-core)} \quad \triangleright \text{basic form of predicate definition without any modifiers}
\]

\[
\text{(pred-def-core)} ::= \text{(predicate-id)} \text{(! (param)}, \text{, (param))*?)} :\text{ - (disjunction)}.
\]

\[
\text{(param)} ::= \text{(var-id)} \quad \triangleright \text{parameter variable}
\]

\[
\text{(disjunction)} ::= \text{(conjunction)} \text{;} \text{ (conjunction)*}
\]

\[
\text{(conjunction)} ::= \text{(literal)} \text{, (literal)*}
\]

\[
\text{(literal)} ::= \text{(atom)} \quad \triangleright \text{positive atom}
| \text{! (atom)} \quad \triangleright \text{negative atom}
\]

\[
\text{(atom)} ::= \text{logic-value} \quad \triangleright \text{4-valued logic constant}
| \text{relation-use} \quad \triangleright \text{relation application}
| \text{(relation-id)*}(\text{term}, \text{term}) \quad \triangleright \text{transitive closure}
| \text{(relation-id)*}(\text{term}, \text{term}) \quad \triangleright \text{reflexive transitive closure}
| \text{metric-use} \text{ in } \text{interval} \quad \triangleright \text{metric check}
| \text{term} \text{ == } \text{(named-object-id)} \text{ find by name}
| \text{(metric-use)} \text{ (relation-id)} \text{((term), (term))*?}
| \text{(metric-use)} \text{ (relation-id)} \text{((term), (term))*?}
\]

Example 4.1. Consider the following two predicate definitions:

\[
\text{directCommunicationLink(from, to) :}
\]

\[
\text{Spacecraft(from), CommunicatingElement(to),}
\text{commSubsystem(from, fromComm),}
\text{target(fromComm, toComm),}
\text{noLinkToGroundStation(s) :}
\]

\[
\text{Spacecraft(s), g == 'gs',}
\text{!directCommunicationLink+(s, g).}
\]

The binary predicate directCommunicationLink matches pairs of Spacecraft objects from from and CommunicatingElement objects to such that there is a direct communication path between the antennas of the two elements. The variables fromComm and toComm of the antennas are (implicitly) existentially quantified.

The noLinkToGroundStation selects Spacecraft objects where there is no indirect communication path to the named 'gs' object, where indirect paths are defined as the transitive closure of the directCommunicationLink relation.

4.1.2. Combining metrics and predicates As a special feature of our language, metrics (see subsection 4.2) and graph predicates can mutually depend on each other. On the one hand, a predicate can check if the value calculated by a metric is within a specific interval (metric value check atom). Reversely, predicates can serve as conditions in aggregation expressions (e.g. to count the number of matches of a predicate) and in conditionals.

Example 4.2. Predicate definition directCommunicationLink from the previous example can be more succinctly written as follows (where Spacecraft and CommunicatingElement are introduced as types):

\[
\text{directCommunicationLink}(\text{Spacecraft from, CommunicatingElement to}) :\text{=}
\]

\[
\text{commSubsystem(from, fromComm), target(fromComm, toComm),}
\text{commSubsystem(to, toComm).}
\]

4.1.4. Error predicates The error keyword may be used to introduce error predicates, which should never hold in a consistent model. This is equivalent to asserting that the predicate is false everywhere in the model.

\[
\text{(predicate-def)} ::= \text{... | error (pred-def-core)} \quad \triangleright \text{error predicate definition}
\]

Example 4.3. To designate noLinkToGroundStation as an error predicate, one may define it as

\[
\text{error noLinkToGroundStation(s) :}
\]

\[
\text{Spacecraft(s), g == 'gs',}
\text{!directCommunicationLink+(s, g).}
\]

which is equivalent to define noLinkToGroundStation without the error keyword and asserting that

\[
\text{noLinkToGroundStation(*) : false.}
\]

For convenience, we also allow error predicates without a predicate name. These definitions are filled in with a newly generated, unique name before translation into assertions.

\[
\text{(unnamed-error)} ::= \text{error(! (param), (param))*?)} :\text{ - (disjunction)}.
\]

Example 4.4. The unnamed error predicate definition

\[
\text{error(s) :}
\]

\[
\text{Spacecraft(s), g == 'gs', !directCommunicationLink+(s, g).}
\]

is equivalent to a definition with a generated unique name

\[
\text{error unnamed0001(s) :}
\]

\[
\text{Spacecraft(s), g == 'gs', !directCommunicationLink+(s, g).}
\]

4.1.5. Functional predicates We introduce the functional keyword to express that the last parameter of an n-ary relation (n ≥ 2) is functionally dependent on its previous parameters, i.e. for any possible binding of the first n − 1 parameters, there is at most only a single possible value for the rth parameter for which the predicate evaluates to true in a consistent model. This is achieved by automatically
Metric definitions are structured similarly to predicate definitions (using the := operator instead of the := operator). The body of the metric definition is a metric expression. Basic elements of metric expressions include numerical constants, terms referencing parameters, other variables or named objects, applications of other metrics, as well as elementary algebraic operators. While we currently only specify a small set of elementary operators (addition, subtraction, multiplication, division, exponentiation and the floor function), the language is easily extensible with additional operators (e.g. logarithms). Similarly to predicate definitions, direct or indirect recursion between metric definitions is not allowed.

Evaluation of the metric expression may fail with a non-number result if an illegal operation, e.g. division by zero is performed. Otherwise, the result of the expression evaluation will be the value of the defined metric. When a term (variable or named object) appears in a metric expression, it must refer to a data node in the graphs and will evaluate to the numeric value bound to the data node. Plain terms not referring to data nodes cause expression evaluation to fail. In contrast with predicate definitions, metric definitions do not quantify over their free variable existentially.

### Example 4.5. When one designates directCommunicationLink as functional by writing

| functional directCommunicationLink
| Spacecraft from, CommunicatingElement to :=
| commSubsystem(from, fromComm),
| target(fromComm, toComm),
| commSubsystem(to, toComm). |

then each from object may have at most one to object associated with it. Thus the following error definition is generated:

| error directCommunicationLinkNotFunctional
| Spacecraft from :=
| equals(to1, to2),
| directCommunicationLink(from, to1),
| directCommunicationLink(from, to2). |

### 4.2. Model metrics

Analogously to predicate definitions that allow computing logic values from (partial) models, we introduce metric definitions to compute numerical values.
In lines 5–7, spacecraftCost calculates the costs associated with a satellite $s$, which is comprised of the unit cost, the cost of the payload and the cost of the associated communications subsystems. Thanks to economies of scale, the unit cost decreases as the number of spacecraft of the same kind is increased. Subsequent calculations except kindCount were omitted from the code example for brevity.

In line 8–11, kindCount calculates the number of spacecraft of the same kind as $s$. This is achieved by a switch expression, which finds the type of the spacecraft, and aggregation expressions, which count spacecraft of a given type.

Lastly, line 12 contains an assertion, which restricts the InterferometryMission o1 to be no more expensive than $50,000,000.

4.2.5. Syntactic sugar for defining metrics Several syntactic shortcuts are provided for defining metrics. First, unary relations can be placed before parameter names to serve as type predicates, similarly to predicate definitions. This is equivalent to introducing a switch expression with a single case, and ensures that the evaluation of the metric fails if the parameter does not satisfy the given relation.

Example 4.7. To restrict the domain of spacecraftCost to Spacecraft objects, we may write

```plaintext
spacecraftCost(Spacecraft s) :=
  basePrice(s) * (kindCount(s) ^ (-0.25)) +
  payloadCost(s) + commSubsysCost(s).
```

which is equivalent to

```plaintext
spacecraftCost(s) := spacecraftCost(s) ->
  basePrice(s) * (kindCount(s) ^ (-0.25)) +
  payloadCost(s) + commSubsysCost(s).
```

Moreover, we introduce a dedicated operator count to count objects satisfying some relations. This replaces the summation of constant 1 values to enhance readability.

```plaintext
(count expresion) ::= ··· | count { (relation-use) }
```

Example 4.8. With the count operator, kindCount can be more concisely written as

```plaintext
kindCount(s) :=
  CubeSat3U(s) -> count { CubeSat3U(_s2) };
  CubeSat6U(s) -> count { CubeSat6U(_s2) };
  SmallSat(s) -> count { SmallSat(_s2) }.
```

Lastly, there is a shorthand for the joint use of functional relations and the single aggregation operator. Relation symbols that are functional can be used as if they were metrics by omitting their last argument. The value of the expression is the value bound to the data object appearing as the single possible value of the omitted argument such that the relation evaluates to true. In other words, we may write

```plaintext
single { m_o | (relation-id) (m_1, · · · , m_{n-1}, m_n) } as (relation-id) (m_1, · · · , m_{n-1}) instead. This allows mimicking object-oriented programming languages more closely, where values of attributes (here represented as functional binary relations) can be accessed directly from objects.
```

```
(metric-expr) ::= ··· | { (relation-use) -> (metric-expr) }*

```
Example 4.9. Because observationTime is a functional relation, the missionCost metric can access its value directly:

```
missionCost(m) ::= 
  sum { spacecraftCost(s) | spacecraft(m, s) } + 100000.0 * observationTime(m).
```

4.3. Scopes

Scope constraints are used to restrict the number of objects represented by multi-objects in a model. While scopes can be expressed with metrics using the count operator, we provide a dedicated facility for scope definitions to separate this concern from the rest of the partial model.

Scope definitions refer to an unary relation acting as a type predicate. They may specify lower or upper bounds or the exact number of objects satisfying the type predicate. In contrast with other model generators, such as Alloy (D. Jackson 2002), there are no restrictions on the unary relations for which scopes can be defined. In particular, they may be arbitrarily overlapping. However, contradictory scope constants, just like contradictory error predicates, lead to partials models without any corresponding consistent concrete model.

```
(scope-decl) ::= scope {relation-id} {comp-op} {int}.
```

Example 4.10. The following scope constraints specify that the model should contain between 16 and 32 domain objects, while exactly 12 objects should be Spacecraft instances.

```
scope domain => 16.
scope domain <= 32.
scope Spacecraft := 12.
```

Without using the notation for scopes, this could have alternatively been written as

```
numberOfDomainObjects() := count { domain(o) }.
numberOfDomainObjects() := [16, 32].
numberOfSpacecraft() := count { Spacecraft(_) }.
numberOfSpacecraft() := [12, 12].
```

4.4. Containment hierarchy

Containment constraints often occur in industrial modeling environments such as UML (Rumbaugh et al. 2004), SysML (Friedenthal et al. 2008) and EMF (“Eclipse Modeling Framework” 2019). While containment constraints can be specified as error predicates or as assertions, we provide a language-level facility to enable an easy definition. The advantages of this approach are twofold:

- Handwritten containment constraints as predicate definition can frequently grow very large due to the need to specify all containment relations in one place.
- Model generators can leverage explicitly signaled containment information to increase their efficiency.

Binary base relations can be designated as containment relations using the containment keyword.Unary base relations can be designated as containment roots using the root keyword.

```
(containment-decl) ::= containment {base-relation-id}.
```

To satisfy the containment hierarchy constraint in a concrete model, for every object o exactly one of the following constraints must hold:

- o is a data object.
- o has a unary base predicate marked as a root that evaluates to true (i.e. it is a root object).
- relation(c, o) evaluates to true for exactly one containment relation relation and object c.

Additionally, the containment relations must span a forest, i.e. there may be no loops along containment relations.

Example 4.11. Consider the root and containment declarations

```
root ConstellationMission.
containment spacecraft.
containment commSubsystem.
```

Each object must either be a data object, a root object, or be connected to some other (containing) object with a spacecraft or a commSubsystem reference. In addition, spacecraft and commSubsystem references jointly form a forest. We can alternatively write this as error patterns as follows:

```
containmentRelation(c, o) :-
  spacecraft(c, o); commSubsystem(c, o).
error cyclicContainment(o) :-
  containmentRelation*(o, o).
numberOfContainers(o) :=
  count { spacecraft(_, o) } +
  count { commSubsystem(_, o) }.
rootRelation(o) :- ConstellationMission(o).
containedObject(o) :-
  domain(o), !rootRelation(o).
error wrongNumberOfContainers(o) :-
  containedObject(o),
  numberOfContainers(o) != 1 ;
  containedObject(o),
  containmentRelation(_, o).
```

Lines 1–2 define the predicate containmentRelation, which matches pairs of objects in any containment relations. The error predicate cyclicContainment in lines 3–4 asserts that cycles in this relation are an error, i.e. containment edges form a forest. Lines 5–7 count a the number of containers than an object o has. Note that instead of counting containmentRelation edges, we sum the counts of the different containment relations. This allows detecting parallel edges (of different containment relations) multiple times, which are erroneous. In line 9–10, the containedObject predicate matches the objects which need exactly one adjacent containment edge. Any other objects must not have a container. This constraint is formalized by the wrongNumberOfContainers error predicate in lines 11–15.

5. Mapping from domain-specific modeling technologies

Most existing model generators rely upon popular modeling technologies to define the core domain concepts in the form of a metamodel. To illustrate the power of abstractions in our
specification language, we demonstrate how the concrete syntax of Xcore (“Xcore” 2020) can be mapped into partial models.

Xcore introduces a type system for models described with a class hierarchy, denoting the generalization relation with the extends keyword and selecting classes that do not have direct instances with the abstract keyword. The class hierarchy is extended by enumerated types with predefined instances (with the enum keyword). References and attributes are defined over the strict type hierarchy with multiplicity constraints (denoted with [lower, upper]) and containment notation (denoted with the contains keyword).

Example 5.1. The following example captures a fragment of the metamodel in Figure 1.

abstract class CommunicatingElement {
  contains CommSubsystem[1, 2] commSubsystem
}
class CubeSat3U extends CommunicatingElement {
  class CommSubsystem {
    refers CommSubsystem[0, 1] target
    CommunicationType[1, 1] type
    real [1, 1] frequency
  }
}
enum CommunicationType { KaComm, UHFCComm, XComm }

CommunicatingElement is defined as an abstract class, which has a concrete subclass CubeSat3U. Each CommunicatingElement contains one or two CommSubsystem instances, which has a non-containment reference to another CommSubsystem, a reference to an enum CommunicationType, and a real frequency value.

In order to map Xcore to partial models, each non-abstract class can be mapped to partial model object ⟨relation-id⟩::new, which represents all potential newly generated instances, and defines its valid type combination (based on the inheritance relations). With this mapping, abstract classes are also represented correctly.

Enumerated types with enum literals ⟨id1⟩, ···, ⟨idn⟩ can mapped to partial model objects by

- defining a named-object for each literal ‘⟨id⟩’,
- setting the type of each object ‘⟨id⟩’ to true for the enum type and false for each other types and
- specifying that no other object has the enum type.

These translations are illustrated in Table 4. The type hierarchy imposes global structural well-formedness constraints over all objects of the partial model. This is enforced by the following structural predicates:

- If an object is an instance of class C then it is an instance of all supertypes Sup_1, ···, Sup_n.
- If an object is an instance of class C then it is not an instance of any incompatible class I_1, ···, I_n that is not connected to C via a directed extends path.
- If an object is an instance of an abstract type A, then it must be an instance of a concrete subtype Sub_1, ···, Sub_n.

Each reference and attribute R is mapped to a relation with default value unknown between the objects, and false between objects with incompatible types I_1 and I_2 (including the ::new objects derived from the classes).

References and attributes also impose extra global structural constraints as follows.

- Type of a relation R between classes S and T is enforced by the following constraint:
- If a reference R has a contains label, then it is a containment relation.
- The multiplicity constraint [lower, upper] is translated into two error predicates that match when the multiplicity is outside of the given range.
The second partial order is implication order, which defined as

\[ (X \leq Y) \iff [(X = \text{false}) \lor (X = Y) \lor (Y = \text{error})] \]

and serves as a generalization of logical implication. We will write \( X \sqsubseteq Y \) and (resp. \( X < Y \)) when \( X \sqsubseteq Y \) (resp. \( X \leq Y \)) and \( X \neq Y \) hold.

The information merge operator \( \oplus \) merges 4-valued truth values where contradictory information results in \text{error}. Other operations on 4-valued truth values \( \neg^4, \lor^4, \land^4 \) are extensions of the usual logic operators \( \neg, \lor, \land \). Their truth tables (see Figure 3b) correspond with their classical counterparts for \( \{\text{false}, \text{true}\} \) inputs.

Semantically, \text{unknown} truth value represents potential \text{true} or \text{false} (or \text{error}) values, and the semantic is chosen to cover all of those options. On the other hand, \text{error} is often unintuitive, but it allows the precise and explicit localization of inconsistencies within models (Belnap 1977; Chechik et al. 2011). For example, we may see that if \( X = \text{error} \) and \( Y = \text{unknown} \), then \( X \lor^4 Y = \text{true} \), because the only way for our logical inference to result in a consistent truth value is to eventually learn that \( Y \) is \text{true}. Should \( Y \) turn out to be \text{false}, the inconsistent \text{error} value will be propagated.

### 6.1.1. Four-valued logic

In this paper, we utilize 4-valued logic to explicitly represent unfinished, partial (paraconsistent) models, as well as errors and inconsistencies (paraconsistency) arising during the evaluation of computations over such models. This section provides semantic foundations for our specification language based on the inconsistency-tolerant Belnap-Dunn 4-valued logic (Belnap 1977; Kamide & Omori 2017), which can reason about runtime errors caused by undefined arithmetic operations, such as division by zero (McKubre-Jordens & Weber 2012).

Belnap-Dunn 4-valued logic contains the usual false \text{false} and true \text{true} truth values, the unknown \text{unknown} value introduced for unspecified or unknown properties, and the inconsistent \text{error} value that signals inconsistencies and computation failures. The subset \( \{\text{false}, \text{true}, \text{unknown}\} \) of logic values can express partial, but consistent information. Conversely, the subset \( \{\text{false}, \text{true}, \text{error}\} \) expresses possibly inconsistent, but complete information.

Two partial orders can be defined over 4-valued logic values (Figure 3a). Information order (denoted by \( \sqsubseteq \)) expresses the gathering of information as new facts are learned during the refinement of partial models. Facts with \text{unknown} logical value can be set to either \text{true} or \text{false}, while a change to \text{error} signifies an inconsistency or failure. This order is defined as

\[ (X \sqsubseteq Y) \iff [(X = \text{unknown}) \lor (X = Y) \lor (Y = \text{error})] \]

Two partial orders can be defined over 4-valued logic values (Figure 3a). Information order (denoted by \( \sqsubseteq \)) expresses the gathering of information as new facts are learned during the refinement of partial models. Facts with \text{unknown} logical value can be set to either \text{true} or \text{false}, while a change to \text{error} signifies an inconsistency or failure. This order is defined as

\[ (X \sqsubseteq Y) \iff [(X = \text{unknown}) \lor (X = Y) \lor (Y = \text{error})] \]

and serves as a generalization of logical implication. We will write \( X \sqsubseteq Y \) and (resp. \( X < Y \)) when \( X \sqsubseteq Y \) (resp. \( X \leq Y \)) and \( X \neq Y \) hold.

The information merge operator \( \oplus \) merges 4-valued truth values where contradictory information results in \text{error}. Other operations on 4-valued truth values \( \neg^4, \lor^4, \land^4 \) are extensions of the usual logic operators \( \neg, \lor, \land \). Their truth tables (see Figure 3b) correspond with their classical counterparts for \( \{\text{false}, \text{true}\} \) inputs.

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#### 6.1.2. Interval arithmetic

We use interval arithmetic to represent unfinished or inconsistent numerical values. A closed, possibly infinite interval \( iv \in IV \subseteq 2^R \) of real numbers denotes a set of possible numerical values. The empty interval \( \emptyset \in IV \) denotes a missing value or a result of failed computation.

The operators \( +^\sharp, -^\sharp, \cdot^\sharp, /^\sharp, \uparrow^\sharp, \Sigma^\sharp, \min^\sharp \) and \( \max^\sharp \) refer to the interval arithmetic (Kulisch 2009) versions of the usual \(+\) (addition), \( -\) (subtraction), \( \cdot\) (multiplication), \( /\) (division), \( \uparrow\) (exponentiation), \( \Sigma\) (summation), \( \min \) and \( \max \) operations over real numbers, respectively. The \( \cup \) symbol denotes the join (smallest interval containing both intervals) of two intervals, while \( \cap \) denotes interval intersection. Additionally, we define a special (directed) multiplication operator, which properly propagates the number of errors when multiplying a value with a number of matches.

\[ iv_1 \uparrow iv_2 = \begin{cases} iv_1 \cdot iv_2 & \text{if } iv_2 \neq \emptyset, \\ [0,0] & \text{if } 0 \in iv_1 \text{ and } iv_2 = \emptyset, \\ \emptyset & \text{if } 0 \notin iv_1 \text{ and } iv_2 = \emptyset. \end{cases} \]

### 6.2. Partial models

First, we provide an algebraic definition for partial models. For that purpose, one needs to establish a signature and a logic structure defined over it.

**Definition 1.** A signature \( \langle \Sigma, \alpha \rangle \) is collection of relation and metric symbols \( \Sigma = \Sigma_R \cup \Sigma_N \) and an arity function \( \alpha : \Sigma \to \mathbb{N} \). \( \Sigma_R \) is the finite set of relation symbols, which includes

\[ \{R_1, \ldots, R_k\}, \text{ which are the base relation symbols}; \]
where all structural predicates are resolved to unknown intervals representing data values uniquely contain a single

A partial model

Definition 3. A partial model \( P \) is regular if it is structurally regular (Definition 11), naming regular (Definition 12) and data regular (Definition 13).

Concrete instance models are special form of partial models where all structural predicates are resolved to true or false values (i.e. unknown and error values are not used) and all intervals representing data values uniquely contain a single value (i.e. when an attribute value is 1 then it is represented as an interval \([1, 1]\)).

Definition 4. A regular partial model \( P \) is concrete, if

for each relation symbol \( r \in \Sigma_R \) which is either a base relation symbol \( R_1, \ldots, R_n \), an object name \( 'N_1', \ldots, 'N_n' \) or a built-in relation symbol \( \text{exists}, \text{equals}, \text{domain}, \text{data}, \text{real}, \text{int} \), the interpretation \( I_P(r) \) contains true and false values only;

for each object \( o \in \mathcal{O}_P \) satisfying \( I_P(\text{real})(o) = \text{true} \), \( \forall_P(\text{value})(o) = [x, x] \) for some real number \( x \in \mathbb{R} \);

for each object \( o \in \mathcal{O}_P \) satisfying \( I_P(\text{int})(o) = \text{true} \), \( \forall_P(\text{value})(o) = [x, x] \) for some whole number \( x \in \mathbb{Z} \).

6.3. Compatibility and inconsistency of partial models

To incorporate the definitions of graph predicates and metrics, we define a respective theory (set of axioms). As such, we can focus only on semantic interpretations of partial models which are compatible with the actual definitions. For example, if a particular metric computes a value as a result, then the interpretation of the data object stored in the partial model should actually retrieve the respective value.

Definition 5. Given a signature \( \langle \Sigma, \alpha \rangle \), a theory \( T \) is a set of axioms

\[
T = \{ P_1(v_1^{(1)}, \ldots, v_{\alpha(P)}^{(1)}) \leftrightarrow \varphi_{P_1}, \ldots, P_l(v_1^{(1)}, \ldots, v_{\alpha(P)}^{(1)}) \leftrightarrow \varphi_{P_l}, M_1(v_1^{(1)}, \ldots, v_{\alpha(M)}^{(1)}) \equiv \mu_{M_1}, \ldots, M_u(v_1^{(1)}, \ldots, v_{\alpha(M)}^{(1)}) \equiv \mu_{M_u} \},
\]

where there is a single axiom for each predicate symbol \( P_i \) and each metric symbol \( M_j \). The variables \( v_i^{(1)} \) and \( v_{\alpha}(M_j) \) correspond to the parameters of the predicate \( P_i \) or the metric \( M_j \), and are free in the predicate definition body \( \varphi_{P_i} \) or metric expression \( \mu_{M_j} \), respectively.

6.3.1. Semantics of predicates and metrics

A defined predicate \( P(v_1, \ldots, v_n) \leftrightarrow \varphi \) can be evaluated on a partial model \( P \) along a variable binding \( Z: \{ v_1, \ldots, v_n \} \rightarrow \mathcal{O}_P \) (denoted as \( \models \varphi \)), which can result in four truth values: true, false, unknown or error. The inductive rules for evaluating the semantics of a logic expression are illustrated in Table 5. We use the notation \( Z' = (Z, u \mapsto o) \) to extend the binding \( Z \) into a new binding \( Z': Z'(v) = o \) if \( v = u \), \( Z(v) \) otherwise.

A metric \( M(v_1, \ldots, v_n) \equiv \mu \) can also be evaluated on a partial model \( P \) along variable binding \( Z: \{ v_1, \ldots, v_n \} \rightarrow \mathcal{O}_P \) (denoted as \( \models \mu \)), resulting in a possibly infinite interval. The empty interval signifies that the evaluation of the metric has failed, e.g. due to division by zero or other undefined algebraic operations. The inductive rules capturing the semantics of a metric expression are illustrated in Table 6.

6.3.2. Compatibility and inconsistency

The proper alignment of the truth values and metric intervals assigned by the semantic interpretation of a partial model and the definition of predicates and metrics (in a theory) is characterized by the concept of compatibility. Informally, a partial model \( P \) is compatible with a theory, if the truth values and metric intervals assigned by the definition and the interpretation are identical.

Definition 6. Given a signature \( \langle \Sigma, \alpha \rangle \) along with a regular partial model \( P \) and a theory \( T \), \( P \) is compatible with \( T \), if

for each predicate symbols \( P \in \Sigma_R \) and for all objects \( o_1, \ldots, o_{\alpha(P)} \in \mathcal{O}_P \), \( \models \varphi_P \) \( = \models \varphi_{P}(o_1, \ldots, o_{\alpha(P)}) \), where \( Z = v_1^{(1)} \mapsto o_1, \ldots, v_{\alpha(P)}^{(1)} \mapsto o_{\alpha(P)} \) and
\[
\begin{align*}
\langle R(v_1, \ldots, v_n) \rangle^P_Z & := \mathcal{I}_P(R)(Z(v_1), \ldots, Z(v_n)) \\
\langle v = 'N' \rangle^P_Z & := \mathcal{I}_P('N')(Z(v)) \\
\langle R+(v_1, v_2) \rangle^P_Z & := \langle R(v_1, v_2) \rangle^P_Z \lor \langle R(v_1, m_1), R(m_1, v_2) \rangle^P_Z \lor \langle R(v_1, m_1), \ldots, R(m_n, v_2) \rangle^P_Z, \\
\langle R^*(v_1, v_2) \rangle^P_Z & := \mathcal{I}_P(\text{equals})(Z(v_1), Z(v_2)) \lor \langle R+(v_1, v_2) \rangle^P_Z \\
\langle \forall \psi \rangle^P_Z & := \bigwedge_{o_{1, \ldots, o_m} \in \mathcal{O}_P} (\neg \exists \langle u_1 \rangle^P_Z \lor \ldots \lor \neg \exists \langle u_m \rangle^P_Z) \lor (\exists \langle \psi \rangle^P_Z)_{o_{1, \ldots, o_m}} \lor (\exists \langle \psi \rangle^P_Z)_{o_{1, \ldots, o_m}}, \\
\langle M(v_1, \ldots, v_n) \rangle^P_Z & := \begin{cases} 
\text{unknown} & \text{if } iv' \cap iv \neq \emptyset \text{ and } iv' \subsetneq iv, \\
\text{error} & \text{if } iv' = \emptyset, \\
\text{false} & \text{if } iv' \neq \emptyset \text{ and } iv' \cap iv = \emptyset, \\
\text{true} & \text{if } iv' \neq \emptyset \text{ and } iv' \subseteq iv
\end{cases}
\end{align*}
\]

Table 5 Inductive semantics of graph predicates.

- for all metric symbols \( M \in \Sigma_M \) and for all objects \( o_1, \ldots, o_{a(M)} \in \mathcal{O}_P \), \( \langle M \rangle^P_Z = \mathcal{V}_P(M)(o_1, \ldots, o_{a(M)}) \), where \( Z = v_m^{(1)} \mapsto o_1, \ldots, v_m^{(a(M))} \mapsto o_{a(P)} \).

Using the 4-valued interpretation, inconsistencies for a domain object in a partial model are recorded in the partial model itself by using the \text{error} values. Similarly, inconsistencies for a data object are identified when the interval of possible values for the data object is the empty set. Note that an inconsistent model cannot be concrete.

**Definition 7.** A regular partial model \( P \) is \text{inconsistent} if

- for some relation symbol \( r \in \Sigma_R \) which is either a base relation symbol \( R_1, \ldots, R_q \), an object name \( 'N_1', \ldots, 'N_l' \) or a built-in relation symbol \text{exists}, \text{equals}, \text{domain}, \text{data}, \text{real} or \text{int}, the interpretation \( \mathcal{I}_P(r) \) contains at least one \text{error} value; or
- there is an object \( o \in \mathcal{O}_P \) with \( \mathcal{I}_P(\text{data})(o) \supseteq \text{true} \) and \( \mathcal{V}_P(\text{value})(o) = \emptyset \).

**6.4. Model generation and optimization by refinement and concretization**

The generation of consistent (concrete) models is driven by a series of refinement and concretization steps where the level of uncertainty in partial models is gradually reduced. When all uncertainties are resolved and there is still no inconsistency in the model then a concrete model is obtained.

**Definition 8.** A refinement from \( P \) to \( Q \) (denoted as \( P \sqsubseteq Q \)) is defined by a refinement function between the objects of the partial model \( \text{ref} : \mathcal{O}_P \rightarrow 2^{\emptyset \cap \top} \) which respect the information ordering of 4-valued logic values and intervals:

- For each relation symbol \( r \in \Sigma_R \), for each object \( p_1, \ldots, p_{a(r)} \in \mathcal{O}_P \) and for each corresponding object \( q_1 \in \text{ref}(p_1), \ldots, q_{a(r)} \in \text{ref}(p_{a(r)}) \),
  \( \mathcal{I}_P(r)(p_1, \ldots, p_{a(r)}) \subseteq \mathcal{I}_Q(r)(q_1, \ldots, q_{a(r)}) \).
- For each metric symbol \( m \in \Sigma_M \), for each object \( p_1, \ldots, p_{a(m)} \in \mathcal{O}_P \) and for each corresponding object \( q_1 \in \text{ref}(p_1), \ldots, q_{a(m)} \in \text{ref}(p_{a(m)}) \),
  \( \mathcal{V}_P(m)(p_1, \ldots, p_n) \supseteq \mathcal{V}_Q(m)(q_1, \ldots, q_n) \).
- All objects in \( Q \) are refined from an object in \( P \):
  \( \mathcal{O}_Q = \bigcup_{p \in \mathcal{O}_P} \text{ref}(p) \).
- Existing objects \( p \in \mathcal{O}_P \) cannot disappear, i.e. they must have non-empty refinements:
  \( [\mathcal{I}_P(\text{exists})(p) \supseteq \text{true}] \Rightarrow [\text{ref}(p) \neq \emptyset] \).

Next, we formally define the task of (consistent) model generation along partial models. Given a model generation task, a \text{complete model generator} outputs some model \( Q \in \text{solutions}(P, T) \) if \( \text{solutions}(P, T) \) is non-empty. Otherwise, it provides a proof of the unsatisfiability of the task.

**Definition 9.** A \text{model generation task} consists of a signature \( \langle \Sigma, \alpha \rangle \) along with a regular partial model \( P \) and a theory \( T \). The
For aggregation operators \(\text{sum}, \text{min}, \text{max}\) and \(\text{single}\) \((\text{aggregation-op}\{\mu \mid \varphi\})^p_Z\), we define the auxiliary functions:

- \(\text{matches}_\varphi(X) = \{Z' = (Z, u_1 \mapsto o_1, \ldots, u_m \mapsto o_m) \mid \mathcal{I}_p(\text{exists})(o_1) \land \cdots \land \mathcal{I}_p(\text{exists})(o_1) \land [\varphi]^p_Z = X\}\),

- \(\text{matchCount}_\varphi(Z') = \begin{cases} \text{count}(Z'(o_1)) \downarrow \cdots \downarrow \text{count}(Z'(o_m)) & \text{if } Z' \in \text{matches}_\varphi(\text{true}), \\ \text{count}(Z'(o_1)) \downarrow \cdots \downarrow \text{count}(Z'(o_m)) \cup [0, 0] & \text{otherwise} \end{cases}\),

- \(\text{count}(o) = \begin{cases} [0, 1] & \text{if } \mathcal{I}_p(\text{exists})(o) = \text{unknown} \text{ and } \mathcal{I}_p(\text{equals})(o, o) = \text{true}, \\ [1, 1] & \text{if } \mathcal{I}_p(\text{exists})(o) = \text{true} \text{ and } \mathcal{I}_p(\text{equals})(o, o) = \text{true}, \\ [1, +\infty] & \text{if } \mathcal{I}_p(\text{exists})(o) = \text{unknown} \text{ and } \mathcal{I}_p(\text{equals})(o, o) = \text{unknown}, \\ \emptyset & \text{if } \mathcal{I}_p(\text{exists})(o) = \text{error} \text{ or } \mathcal{I}_p(\text{equals})(o, o) = \text{error}, \end{cases}\)

where \(\varphi\) has free variables \(u_1, \ldots, u_m\).

\[
\langle \text{sum} \{\mu \mid \varphi\}\rangle^p_Z := \begin{cases} \emptyset & \text{if } \text{matches}_\varphi(\text{error}) \neq \emptyset, \\ \bigcup_{Z' \in \text{matches}_\varphi(\text{unknown}) \cup \text{matches}_\varphi(\text{true})} \text{matchCount}_\varphi(Z') \uplus \langle \mu \rangle^p_Z & \text{otherwise} \end{cases}
\]

\[
\langle \text{min} \{\mu \mid \varphi\}\rangle^p_Z := \begin{cases} \emptyset & \text{if } \text{matches}_\varphi(\text{error}) \neq \emptyset, \\ \min^{\uparrow} \{iv_1, iv_2\} & \text{other}\text{wise} \end{cases}
\]

where \(iv_1 = \min^{\uparrow} Z' \in \text{matches}_\varphi(\text{unknown}), \mu^p_{Z'} \neq \emptyset \langle \mu \rangle^p_Z \) and \(iv_2 = \min^{\uparrow} Z' \in \text{matches}_\varphi(\text{true}) \langle \mu \rangle^p_Z \).

\[
\langle \text{max} \{\mu \mid \varphi\}\rangle^p_Z := \begin{cases} \emptyset & \text{if } \text{matches}_\varphi(\text{error}) \neq \emptyset, \\ \max^{\uparrow} \{iv_1, iv_2\} & \text{otherwise} \end{cases}
\]

where \(iv_1 = \max^{\uparrow} Z' \in \text{matches}_\varphi(\text{unknown}), \mu^p_{Z'} \neq \emptyset \langle \mu \rangle^p_Z \) and \(iv_2 = \max^{\uparrow} Z' \in \text{matches}_\varphi(\text{true}) \langle \mu \rangle^p_Z \).

\[
\langle \text{single} \{\mu \mid \varphi\}\rangle^p_Z := \begin{cases} \emptyset & \text{if } \text{matches}_\varphi(\text{error}) \neq \emptyset \text{ or } \text{matches}_\varphi(\text{true}) \geq 2, \\ \langle \mu \rangle^p_Z \in Z' & \text{if } \text{matches}_\varphi(\text{error}) = \emptyset \text{ and } \text{matches}_\varphi(\text{true}) = \{Z'\}, \\ \bigcup_{Z' \in \text{matches}_\varphi(\text{unknown})} \langle \mu \rangle^p_Z & \text{otherwise} \end{cases}
\]

Table 6 Inductive semantics of metric expressions.
solutions of the model generation task are

\[ \text{solutions}(P, T) = \{ Q \mid P \subseteq Q \text{ and } Q \text{ is concrete and it is compatible with } T \}. \]

As a corollary, if a partial model \( P \) is inconsistent, then its refinements \( Q (P \subseteq Q) \) cannot be concrete. Hence, when searching for a concrete, consistent refinement of \( P \), model generators can abandon inconsistent partial models \( Q \) without compromising the completeness of model generation.

An optimizing model generator aims to find a consistent model \( Q \in \text{optimal}(P, T, M) \) optimal with respect to a given objective \( M \) if such a model exists. As such, conceptually, a model generator can also handle model optimization challenges with appropriate modifications of the search strategy (e.g. using a branch-and-bound strategy).

**Definition 10.** A single-objective model optimization task is a model generation task \( (P, T) \) over a signature \( (\Sigma, \alpha) \) along with a 0-ary metric symbol \( M \in \Sigma (\alpha(M) = 0) \) serving as the objective to be maximized. The optimal solutions of the task are

\[ \text{optimal}(P, T, M) = \{ Q \mid Q \in \text{solutions}(P, T) \text{ and } \forall Q' \in \text{solutions}(P, T): V_Q(M)(\cdot) \geq V_{Q'}(M)(\cdot) \}. \]

Note that the definition of a model generation task complies (but generalizes) the previous formalization of the model generation problem in Varró et al. (2018). As such, existing model generation techniques and tools (Semeráth et al. 2018, 2020) remain applicable in our context. On the other hand, these existing approaches do not support an optimizing model generator, which is to be targeted in future work.

7. Related work

Several software engineering and verification approaches are based on the automated generation or extension of graph-based models. In the following, we are summarizing those techniques and their relation with our partial modeling language.

7.1. Model generation for domain-specific languages

Several approaches map a model modeling artifacts into a logic problem solved by an underlying backend solver. There are several approaches aiming to validate standardized engineering models enriched with OCL constraints (Gogolla et al. 2005) by relying upon different back-end logic-based approaches such as constraint logic programming (Cabot et al. 2007, 2008; Büttner & Cabot 2012), SAT-based model finders (Shah et al. 2009; Anastasakis et al. 2010; Büttner et al. 2012; Kuhlmann et al. 2011; Soeken et al. 2010; Semeráth et al. 2017, 2016), CSP solvers (González et al. 2012) first-order logic (Beckert et al. 2002), constructive query containment (Queralt et al. 2012), higher-order logic (Brucker & Wolff 2007; Grönniger et al. 2009), rewriting logic (Clavel & Egee 2008), or genetic algorithms (Soltana et al. 2017, 2020).

Partial snapshots, metamodels and a large fragment of constraints, defined either as OCL constraints (“Object Constraint Language, v2.4” 2014) or Viatra queries (Varró et al. 2016), can be uniformly represented as first order constraints (Semeráth et al. 2017). As a main conceptual difference, this approach operates directly on a mathematically precise formalism, but, for the sake of reusability, it avoids direct integration with existing modeling tools.

7.2. Partial modeling

Partial models are similar to uncertain models, which offer a rich specification language (Famelis et al. 2012; Salay & Chechik 2015) amenable to analysis. They provide a more intuitive, user-friendly language compared to our mathematical notation, but without handling additional predicates and metrics. With respect to expression power, Semeráth & Varró (2017) presented a rewriting techniques that maps uncertain models to (3-valued) partial models. Potential concrete models compliant with an uncertain model can be synthesized by the Alloy Analyzer (Salay et al. 2012), or refined by graph transformation rules (Salay et al. 2015).

In some extended approaches, like Famelis et al. (2013), it is possible to analyze predicates and executions of model transformation rules on partial models by using a SAT solver (MathSAT4) or by automated graph approximation (referred to as “lifting”), or by graph query engines (Semeráth & Varró 2017). As a key difference, our approach carries out model refinement while simultaneously evaluating predicates and metrics.

7.3. Feature modeling

Feature modeling is an engineering formalism to define the possible combination of selected features from a large selection of atomic features (Lee et al. 2002; Griss et al. 1998). While configurations (i.e. a valid combination of features) can be represented as a graph model (where each object is named, and only edges and the exists relation have unknown values), not all graph generation problems can be represented as a feature modeling problem.

Clafer (Bak et al. 2010) is a lightweight structural modeling language used for feature modeling with minimalistic syntax and rich semantics equivalent to first-order relational logic. The specification language supports structural modeling, constraints (well-formedness constraints are written in their own language, which is said to be equivalent to FOL) and also partial configurations. Partial configurations are like partial snapshots in our approach: instance models with undefined attributes and features that can be the basis of model completion. DSL specification given in Clafer are validated using the Clafer Tools (Antkiewicz et al. 2013) that supports various tasks for domain engineering, like consistency checking and instance model generation based on backend reasoners like Alloy or Choco (Liang 2012).

7.4. Symbolic approaches

Certain techniques use abstract (or symbolic) graphs for analysis purposes. A tableau-based reasoning method is proposed for graph properties (Schneider et al. 2017; Pennemann 2008; Alsibahi et al. 2016), which automatically refines solutions based on WF constraints, and handles the state space in the form of a resolution tree as opposed to a partial model. When scalability
evaluation is included, these techniques demonstrated to derive only small graphs (< 10 objects).

7.5. Shape analysis
Another group of symbolic approaches introduces sophisticated type graphs to uniformly represent objects with similar properties. Reps et al. (2004); Ferrara et al. (2012); Gopan et al. (2004) introduce predicate abstraction techniques for graphs using 3-valued logic, which is used as a theoretical basis for our model generation technique. In those approaches, concretization is used to materialize (typically small) counter-examples for designated safety properties in a graph transformation system. However, their focus is to support model checking of abstract graph transformation systems, which can evaluate complex trajectories, but do not scale in the size of the models. A similar neighborhood-based abstract interpretation technique is introduced by Rensink & Distefano (2006), where each abstract object is represents objects with similar neighborhood (up to a given range).

Handling numeric (integer or real) variables and constraints in model generation scenarios requires their abstract interpretation through numerical abstract domains (Miné 2004; Singh et al. 2018). Numerical abstract domains may be used to summarize object attributes in value analysis of heap programs (Magill et al. 2007; McCloskey et al. 2010; Ferrara et al. 2012). Summarized dimensions (Gopan et al. 2004) were introduced to succinctly represent attributes of a potentially unbounded set of objects via multi-objects. This approach enables attribute handling in three-valued partial models, and allows checking for refinements by abstract subsumption (Anand et al. 2009). But these approaches do not generate graph models.

7.6. Specification frameworks
Complete frameworks with standalone specification languages include Formula (E. K. Jackson et al. 2011), which uses the Z3 SMT-solver (de Moura & Björner 2008), Alloy (D. Jackson 2002), which relies on a similar relational logic (Toralak & Jackson 2007) and SAT-solvers like Sat4j (Le Berre & Parrain 2010), and Clafer (Bak et al. 2013), which uses reasoners like Alloy.

As a main difference, our specification language is capable of uniformly representing various model generation and optimization tasks, any initial or intermediate state of model generation, and the concrete models generated as result. As such, it is especially suitable for support iterative model generation, where the output of one run can be used as the input of the next run.

8. Conclusions and future work
In this paper, we have proposed a novel specification language for partial models to be used for consistent graph model generation. While a novel class of model generators derive consistent instance models along a refinement calculus of partial models, we intend to uniformly represent any intermediate state of the generation (not only the initial and final states). Key features of our specification language have been presented along a series of examples in the context of a complex case study proposed by NASA researchers. As a major novelty, our language seamlessly integrates partial models with graph predicates and graph metrics, which can mutually depend on each other. We define precise semantics for our language using a 4-valued interpretation based on Belnap–Dunn logic. On a practical note, a parser and textual editor has been implemented using the Xtext technology.

The proposed language can serve as a user-friendly front-end for model generation tasks used in a series of past papers (Semeráth et al. 2018, 2020) as well as in other modeling challenges. However, no support exists currently for model optimization tasks, which provides the main direction of our future work.

References


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A. Grammar of the configuration language

Below we provide a grammar for our proposed configuration language in a form that suitable for illustrative purposes, but is not context-free. For a slightly more complex form that is suitable for implementing a parser, we refer to our implementation available at https://github.com/viatra/VIATRA-Generator.

\[
\begin{align*}
(id-fragment) \ ::= & \ (a \cdots z \mid A \cdots Z \mid \_ \mid 0 \cdots 9)* \\
(id) \ ::= & \ (id-fragment) \ (\::\ (id-fragment))^* \\
(integer) \ ::= & \ (-)? \ (0 \cdots 9)+ \\
(real) \ ::= & \ (-)? \ (0 \cdots 9)+ \ (0 \cdots 9)+? \\
(object-id) \ ::= & \ (named-obj-id) \ | \ (unnamed-obj-id) \\
(char) \ ::= & \ any \ Unicode \ character \ except \ ' \\
(named-obj-id) \ ::= & \ '{char}+* \\
(unnamed-obj-id) \ ::= & \ (id) \\
(relation-id) \ ::= & \ (builtin-rel-id) \ | \ (base-rel-id) \ | \ (predicate-id) \\
(predicate-def) \ ::= & \ (pred-def-core) \ | \ (error) \\
(pred-def-core) \ ::= & \ (relation-id) \ ((\ (param)\ (param)* ))? \ :- \ (disjunction) . \\
(param) \ ::= & \ (var-id) \ | \ (relation-id) \ (relation-use) \\
(disjunction) \ ::= & \ (conjunction) \ (\; \ (conjunction))^* \\
(conjunction) \ ::= & \ (literal) \ (\; \ (literal))^* \\
(literal) \ ::= & \ (atom) \ | \ (atom) \\
(atom) \ ::= & \ (logic-value) \ | \ (4-valued \ logic \ constant) \\
(metric-expr) \ ::= & \ (real) \ | \ (real) \\
(real) \ ::= & \ (real) \ | \ +\ inf \\
(interval) \ ::= & \ ([\ (lower-bound) \ , \ (upper-bound) ] \\
(lower-bound) \ ::= & \ (real) \ | \ -\ inf \\
\end{align*}
\]
abstract class Spacecraft extends 
class GroundStationNetwork extends 
abstract class CommunicatingElement {
contains CommSubsystem [1, 2] commSubsystem
}

B. Example configuration file
To showcase the usability of our configuration language, we present a formalization of the case study from section 2 below.

% Xcore-style metamodel definition
class InterferometryMission {
contains GroundStationNetwork [1, 1]
groundStationNetwork
contains Spacecraft [2, +inf] spacecraft
int observationTime }
}
abstract class CommunicatingElement {
contains CommSubsystem [1, 2] commSubsystem
}

class GroundStationNetwork extends 
CommunicatingElement { }
abstract class Spacecraft extends 
CommunicatingElement { 
contains Payload [0, 1] payload
}
C. Regularity of partial models

Definition 11. A partial model $P = (\mathcal{O}_P, \mathcal{I}_P, \mathcal{V}_P)$ is structurally regular if it satisfies the following conditions:

S1. $\forall o \in \mathcal{O}_P: \mathcal{I}_P(\text{exists})(o) > \text{false}$
   $\triangleright$ non-existing objects are omitted

S2. $\forall o \in \mathcal{O}_P: \mathcal{I}_P(\text{equals})(o, o) > \text{false}$
   $\triangleright$ the equals relation is reflexive

S3. $\forall o_1, o_2 \in \mathcal{O}_P: \mathcal{I}_P(\text{equals})(o_1, o_2) = \mathcal{I}_P(\text{equals})(o_2, o_1)$
   $\triangleright$ equals is symmetric

S4. $\forall o_1, o_2 \in \mathcal{O}_P: (o_1 \neq o_2) \Rightarrow \mathcal{I}_P(\text{equals})(o_1, o_2) < \text{true}$
   $\triangleright$ if two objects are different, then they cannot be equal

S5. $\forall o \in \mathcal{O}_P: \mathcal{I}_P(\text{domain})(o) = \sim\mathcal{I}_P(\text{data})(o)$
   $\triangleright$ objects are partitioned into domain and data objects

Definition 12. A partial model $P = (\mathcal{O}_P, \mathcal{I}_P, \mathcal{V}_P)$ is naming regular if it satisfies the following conditions:

N1. $\forall o \in \mathcal{O}_P, 'N' \in \Sigma: \mathcal{I}_P('N')(o) \leq \mathcal{I}_P(\text{domain})(o)$
   $\triangleright$ named objects must be domain objects

N2. $\forall o \in \mathcal{O}_P, 'N' \in \Sigma: \mathcal{I}_P('N')(o) \neq \text{unknown}$
   $\triangleright$ names cannot be uncertain

N3. $\forall o_1, o_2 \in \mathcal{O}_P, 'N' \in \Sigma: [\mathcal{I}_P('N')(o_1) \equiv \text{true} \land \mathcal{I}_P('N')(o_2) \equiv \text{true}] \Rightarrow (o_1 \equiv o_2)$
   $\triangleright$ any name 'N' may only belong to a single object

N4. $\forall o \in \mathcal{O}_P, 'N_1', 'N_2' \in \Sigma: [\mathcal{I}_P('N_1')(o) \equiv \text{true} \land \mathcal{I}_P('N_2')(o) \equiv \text{true}] \Rightarrow ('N_1' = 'N_2')$
   $\triangleright$ an object cannot have more than one name

D. Parsing and serialization

A textual description of a model generation problem simultaneously encodes a signature ($\Sigma, a$), a partial model $P = (\mathcal{O}_P, \mathcal{I}_P, \mathcal{V}_P)$ and a theory $T$. The input language from section 3 can be parsed as follows:

1. First, syntactic sugars are removed from the input by translating them back into core concepts. This simplifies the input for the subsequent steps.

2. Symbols are gathered from the input. Base relations are turned into base relation symbols $R_1, \ldots, R_m$, relations with a corresponding predicate definition are turned into predicate symbols $P_1, \ldots, P_k$, names of named objects are turned into object name symbols $'N_1', \ldots, 'N_k'$ and metrics are turned into metric symbols $M_1, \ldots, M_n$. Together with the build-in symbols, these form the set of the signature. The arity $a(s)$ of each symbol $s \in \Sigma$ is set to the number of arguments in the input text.

3. To obtain $\mathcal{O}_P$, we add an object $o_x$ for every unnamed object identifier $x$ and an object $o'_{N'}$ for every object name $'N'$ in the input problem.

4. We initialize $\mathcal{I}_P$ by setting $\mathcal{I}_P('N')(o \cdot 'N') = \text{true}$ for each object name $'N'$. For all other $\mathcal{O}_P \ni o' \neq o \cdot 'N', \mathcal{I}_P('N')(o') = \text{false}.$
5. For all relation symbols \( r \in \Sigma_R \) and objects \( o_1, \ldots, o_{a(r)} \in \mathcal{O}_P \), we gather all assertions of the form \( r(o_1, \ldots, o_{a(r)}) \colon X \), and set \( \mathcal{I}_P(r(o_1, \ldots, o_{a(r)})) = \bigcup \{ X \mid r(o_1, \ldots, o_{a(r)}) = X \} \).

6. For all objects \( o \in \mathcal{O}_P \), we gather all value assertions of the form \( o : \textit{iv} \). and set \( \mathcal{V}_P(\textit{value})(o) = \bigcap \{ \textit{iv} \mid (o : \textit{iv}) \}. \) If there are no such assertions, we set \( \mathcal{V}_P(\textit{value})(o) = [-\infty, +\infty] \) instead.

7. For all metric symbols \( M \in \Sigma_M \) and objects \( o_1, \ldots, o_{a(M)} \in \mathcal{O}_P \), we gather all value assertions of the form \( M(o_1, \ldots, o_{a(M)}) : \textit{iv} \). and set \( \mathcal{V}_P(M)(o_1, \ldots, o_{a(M)}) = \bigcap \{ \textit{iv} \mid M(o_1, \ldots, o_{a(M)}) = \textit{iv} \}. \) If there are no such assertion, we set \( \mathcal{V}_P(M)(o_1, \ldots, o_{a(M)}) = [-\infty, +\infty] \) instead.

8. Lastly, we gather all predicate and metric definitions into the theory \( T \).

### D.1. Regularization

After parsing the textual description of a model generation problem, the resulting partial model \( P \) may not necessarily be regular. Hence, a \textit{regularization} procedure is applied to obtain an equivalent partial model \( P' \) as a refinement \( P \subseteq P' \) that is regular. To simplify presentation, in the steps below, \( P \) always refers to the output of the previous step. Thus, \( P \) serves as the input of the currently executed step, while \( P' \) is the output of the current step.

1. **Naming regularity** conditions N2–4 are satisfied by construction after parsing. To ensure N1, we only need to set \( \mathcal{I}_P(\text{domain})(o) = \mathcal{I}_P(\text{domain})(o) \lor \mathcal{I}_P(\text{'N'})(o) \) for all \( o \in \mathcal{O}_P \) and \( \text{'N'} \in \Sigma \).

2. **Structural regularity** is enforced as follows:

   - **S1.** Surely non-existent (\( \mathcal{I}_P(\text{exists})(o) = \text{false} \)) objects are removed.
   - **S2.** We set \( \mathcal{I}_P(\text{equals})(o_1, o_2) = \text{error} \) for all \( o_1, o_2 \in \mathcal{O}_P \) where \( \mathcal{I}_P(\text{equals})(o_1, o_2) = \text{false} \).
   - **S3.** For all \( o_1, o_2 \in \mathcal{O}_P \), we set \( \mathcal{I}_P(\text{equals})(o_1, o_2) = \mathcal{I}_P(\text{equals})(o_1, o_2) \lor \mathcal{I}_P(\text{equals})(o_2, o_1) \) in order to make \text{equals} symmetric.
   - **S4.** Surely equal objects \( \mathcal{I}_P(\text{equals})(o_1, o_2) = \text{true} \) objects are merged. While merging objects, relevant interpretations of relational symbols are combined with the \( \lor \) information merge operator \( \mathcal{I}_P(r)(o_1, o_2) = \mathcal{I}_P(r)(o_1) \lor \mathcal{I}_P(r)(o_2) \). Similarly, interpretations of metric symbols are combined with the \( \cap \) interval intersection operator.
   - **S5.** We set
     \[
     \mathcal{I}_P(\text{data})(o) = \mathcal{I}_P(\text{data})(o) \lor \neg^4 \mathcal{I}_P(\text{domain})(o) \\
     \lor \mathcal{I}_P(\text{real})(o) \\
     \lor \mathcal{I}_P(\text{int})(o), \\
     \mathcal{I}_P(\text{domain})(o) = \neg^4 \mathcal{I}_P(\text{data})(o)
     \]
     for all \( o \in \mathcal{O}_P \).

3. **Data regularity** is enforced as follows:

   - **D1.** For all objects \( o \in \mathcal{O}_P \) with \( \mathcal{I}_P(\text{data})(o) \equiv \text{false} \), we set \( \mathcal{V}_P(\text{value})(o) = \emptyset \).
   - **D2.** When we enforced S5, D2 was also enforced as a side effect.
   - **D3.** If \( \mathcal{I}_P(\text{real})(o) \land^4 \mathcal{I}_P(\text{int})(o) = \text{true} \) for an object \( o \in \mathcal{O}_P \), we set all of \text{real, int, data} and \text{domain} for \( o \) to \text{error}.
   - **D4.** If \( \mathcal{I}_P(\text{int})(o) \equiv \text{true} \) and \( \mathcal{V}_P(\text{value})(o) \cap Z = \emptyset \) for some object \( o \), we set \( \mathcal{V}_P(\text{value})(o) = \emptyset \).

In order to enforce D5 and D6, we merge \text{real} and \text{int} objects that are bound to the same numbers (similarly to how we enforced S4).

Each of these steps are partial model refinements, hence the ultimate output \( P' \) is a refinement of the initial input \( P \).

### D.2. Serializing model generation problems

Serializing a partial model allows us to inspect the internal state of model generators. Moreover, model generation can be paused and its state can be saved to derive a new model generation problem either equivalent to the original one, or extended with new objects, assertions and constraints.

Regular partial models \( P \) and theories \( T \) over a signature \( \Sigma_a \) can be serialized into a textual description of a model generation problem that, when parsed, gives rise to an identical signature, partial model and theory.

1. For each named object \( o \in \mathcal{O}_P \) with name \( \text{'N'} \), \( \mathcal{I}_P(\text{'N'})(o) \equiv \text{true} \), \( o \) will be uniquely represented by the name \( \text{'N'} \). For all other objects, we assign unique identifiers \( x_o \).
2. For each relation symbol \( r \in \Sigma_R \) and objects \( o_1, \ldots, o_{a(r)} \), an assertion \( r(o_1, \ldots, o_{a(r)}) : \mathcal{I}_P(r)(o_1, \ldots, o_{a(r)}) \) is printed, where each \( o_1 \) is represented either by its name or by its unique identifier.
3. For each object \( o \in \mathcal{O}_P \) with \( \mathcal{I}_P(\text{data})(o) \equiv \text{true} \), an assertion \( o : \mathcal{V}_P(\text{value})(o) \). is printed.
4. For each metric symbol \( M \in \Sigma_M \) and objects \( o_1, \ldots, o_{a(M)} \), an assertion \( M(o_1, \ldots, o_{a(M)}) : \mathcal{V}_P(M)(o_1, \ldots, o_{a(M)}) \). is printed.
5. Lastly, axioms of the theory \( T \) are printed as predicate definitions and metric definitions.