Combining Declarative and Procedural Views in the Specification and Analysis of Product Families

Maurice H. ter Beek
LIACS, Leiden University, NL
maurice.terbeek@istl.cnr.it

Alberto Lluch Lafuente
IMT
Lucca, IT
alberto.luch@imtlucca.it

Marinella Petrocchi
IIT–CNR
Pisa, IT
marinella.petrocchi@iit.cnr.it

ABSTRACT

We introduce the feature-oriented language FLAN as a proof of concept for specifying both declarative aspects of product families, namely constraints on their features, and procedural aspects, namely feature configuration and run-time behaviour. FLAN is inspired by the concurrent constraint programming paradigm. A store of constraints allows one to specify in a declarative way all common constraints on features, including inter-feature constraints. A standard yet rich set of process-algebraic operators allows one to specify in a procedural way the configuration and behaviour of products. There is a close interaction between both views: (i) the execution of a process is constrained by its store to forbid undesired configurations; (ii) a process can query a store to resolve design and behavioural choices; (iii) a process can update the store by adding new features. An implementation in the Maude framework allows for a variety of formal automated analyses of product families specified in FLAN, ranging from consistency checking to model checking.

Categories and Subject Descriptors

D.2.4 [Software Engineering]: Software/Program Verification—Formal methods, Model checking, Validation

General Terms

Design, Experimentation, Verification

Keywords

Product families, Variability, Process algebra, Concurrent constraint programming, Behavioural analyses, Maude

1. INTRODUCTION

Research on applying formal methods in SPLE traditionally focusses on modelling and analysing structural rather than behavioural constraints in product families. However, many software-intensive systems are embedded, distributed and safety critical, making it important to be able to model and analyse also their behaviour, as a form of quality assurance. Recent years have witnessed a growing interest in specifically considering also the behavioural variability of product families. This has resulted in variants of UML diagrams [33], extensions of Petri nets [27, 28] and a variety of frameworks with transition system semantics [14, 20, 17, 22, 11, 3]. As a result, behavioural analysis techniques such as model checking have become available for the verification of (temporal) logic properties of product families.

Specifying a product family directly in an operational model is often not feasible. Therefore it can be useful to resort to high-level formal languages with semantics over those operational models, as is common in the context of process algebra. Several extensions of CCS [26] have been proposed to model product families [15, 17, 18, 23], but none of these can combine behavioural constraints with all common structural constraints known from feature models.

We introduce here the feature-oriented language FLAN as a proof of concept for specifying product families by taking both structural and behavioural constraints into account. It is inspired by concurrent constraint programming [30] and its application in process algebra [8]. A store of constraints allows one to specify all common structural constraints known from feature models in a declarative way, including inter-feature constraints (also called cross-tree constraints). Moreover, a rich set of process-algebraic operators allows one to specify in a procedural way both the configuration and behaviour of products. FLAN’s semantics thus unifies static and dynamic feature selection.

The declarative and procedural views are closely related: (i) the execution of a process is constrained by its store, e.g. to avoid introducing inconsistencies; (ii) a process can query a store to resolve configuration and behavioural options; (iii) a process can update the store by adding new features.

Inspired by [15], we implemented FLAN in the executable modelling language Maude [12], whose rich toolkit enables the application of a variety of formal automated analysis techniques to product families specified in FLAN, from consistency checking to model checking.

The paper is organised as follows. Section 2 describes a running example of a family of coffee machines. In Section 3 we present the syntax and semantics of FLAN and a specification of the example. Section 4 illustrates the Maude-supported automated analyses of the example. We discuss related work in Section 5, report some concluding remarks in Section 6 and list promising future work in Section 7.
2. A FAMILY OF COFFEE MACHINES

We use a popular running example in the style of [2, 3, 4, 5, 11, 15, 27, 28]. It describes a (simplified) family of coffee machines in terms of the following list of requirements:

1. Initially, a coin must be inserted: either a euro, exclusively for products for the European market, or a dollar, exclusively for Canadian products;
2. Upon the insertion of a coin, a choice for sugar must be offered, followed by a choice of beverages;
3. The choice of beverage (coffee, tea, cappuccino) varies, but every product must offer at least one beverage, tea may be offered only by European products, and all products that offer cappuccino must also offer coffee;
4. Optionally, a ringtone may be rung after the delivery of a beverage. However, a ringtone must be rung after serving a cappuccino;
5. After the beverage is taken, the machine returns idle.

These requirements define products by combining structural constraints defining valid feature configurations (e.g. “every product must offer at least one beverage”) with temporal constraints defining valid behaviour, i.e. action sequences (e.g. “a ringtone must be rung after serving a cappuccino”).

3. FLAN: SYNTAX AND SEMANTICS

The feature-oriented language FLAN we propose is loosely inspired by the CCS-like process algebra CL4SPL presented in [15], but it strongly differs in its treatment of inter-feature constraints and in the separation of declarative and procedural aspects inspired by the concurrent constraint programming paradigm [30] and its adoption in process calculi [8].

The core notions of FLAN are features, constraints, processes and fragments, which can all be identified in the syntax of FLAN given in Figure 1. More precisely, \( f \) and \( g \) range over features and syntactic categories \( S, P \) and \( F \) correspond to constraints, processes, and fragments, respectively.

**Features.** A feature is a term describing specific elements or properties of a product. The universe of features is denoted by \( F \). The features of our running example are the coins accepted (i.e. euro and dollar), the products offered (i.e. coffee, tea and cappuccino) and additional elements like sugar (the capability to regulate the quantity of sugar) and ringtone (the capability to emit a ringtone). Our approach is general enough to accommodate all common notions of features [10].

**Constraints.** The declarative part of FLAN is represented by a store of constraints which defines both constraints on features extracted from the product requirements and additional information (e.g. about the context wherein the product will operate).

Two important notions of constraint stores are (i) the consistency of a store \( S \), denoted by consistent\((S)\), which in our case amounts to logical satisfiability of all constraints constituting \( S \); and (ii) the entailment \( S \models c \) of constraint \( c \) in store \( S \), which in our case amounts to logical entailment.

A constraint store is any term generated by \( S \) in the grammar of FLAN. The most basic constraint stores are \( \top \) (no constraint at all), \( \bot \) (inconsistent) and ordinary boolean propositions (generated by \( K \)). Constraints can be combined by juxtaposition (its semantics amounts to logical conjunction).

We assume that the standard structural constraints on features (like options, obligations and alternatives) are expressed using boolean propositions (e.g. as explained in [31]). For this purpose, we assume that the universe \( P \) of propositions contains a boolean predicate \( has(f) \) that can be used to denote the presence of a feature \( f \) in a product. Boolean propositions can also be used to represent additional information such as contextual facts. Examples from our running example are \( in(Europe) \) and \( in(Canada) \), respectively used to state the fact that the coffee machine being configured is meant to be used in Europe or in Canada. Boolean propositions can state relations between contextual information and features, like \( in(Europe) \rightarrow has(euro) \) (i.e. a coffee machine for the European market needs a euro coin slot).

**Cross-tree constraints**, instead, are handled as first-class citizens. A constraint \( f \triangleright g \) expresses that feature \( f \) requires the presence of feature \( g \) while a constraint \( f \triangleright g \) expresses that features \( f \) and \( g \) mutually exclude each other’s presence (i.e. they are incompatible). Of course, also these constraints can be encoded as boolean propositions. For instance, \( f \triangleright g \) and \( f \triangleright g \) can equivalently be expressed as \( has(f) \leftrightarrow \neg has(g) \) and \( has(f) \rightarrow has(g) \), respectively. We indeed use such logical encoding to reduce consistency checking and entailment to logical satisfiability (and hence exploit Maude’s SAT solver). However, we prefer to keep here this first-class treatment in order to emphasise their use in the presentation of our work.

We also consider a class of action constraints, reminiscent of Featured Transition Systems [11]. However, while transitions in FTSs are subject to the mere presence of features, in our approach we associate arbitrary constraints to actions rather than transitions. For instance, in a coffee machine equipped with a slot for euro coins we will use \( euro \) for the action of inserting a euro coin and \( do(euro) \) as a proposition stating the execution of that action. The relations between the action \( euro \) and the presence of the corresponding feature \( euro \) can be formalised as \( do(euro) \rightarrow has(euro) \), i.e. the insertion of a euro coin requires the presence of an appropriate coin slot. In general, we assume that each action \( a \) may have a constraint \( do(a) \rightarrow p \). Such constraints act as a sort of guard to allow or forbid the execution of actions (as illustrated later on in the discussion of rule \( Act \)).

The constraint store \( S \) in Figure 1 formalises part of the requirements specified in Section 2 for our running example. It contains both contextual information (e.g. \( in(Europe) \)) and action constraints (e.g. \( do(euro) \rightarrow has(euro) \)). For instance, from requirement 1 we extract that \( euro \) and \( dollar \) are mutually exclusive features (formalised as \( dollar \triangleright euro \)),
while from requirement 3 we understand that cappuccino requires coffee (formalised as cappuccino ▷ coffee).

**Processes.** The procedural part of FLan is represented by processes. A process can be one of the following:

- 0, the empty process that can do nothing;
- X, where X is a process identifier. We assume that there is a set of process definitions of the form X ⊳ P. We also assume that recursively defined processes are finitely branching, which can be ensured in standard ways (e.g., prefixing every occurrence of a process identifier X with an action or constraining process definitions to be of the form X ⊳ A.P);
- A.P, a process willing to perform the action A and then to behave as P;
- P + Q, a process that can non-deterministically choose to behave as P or as Q;
- P ∥ Q, a process that must progress first as P and then as Q;
- P | Q, a process formed by the parallel composition of P and Q, which evolve independently.

It is worth remarking that we distinguish between ordinary actions (from a universe A) and the special actions install(f) (used to denote the dynamic installation of a feature f) and ask(K) (used to query the store). We will see that each action type is treated differently in rules of the operational semantics.

In our example, we will consider the following actions: euro and dollar (insertion of the respective coin); sugar (sugar selection); coffee, tea, and cappuccino (beverage selection); and ringtone (ringtone emission).

**Fragments.** Finally, a fragment F is a term [S || P], composed by a store of constraints S and a process P. Each of the components of a fragment may influence each other, along the lines of the concurrent constraint programming paradigm (30): a process may update its store which, in turn, may condition the execution of process actions.

The operational semantics of closed fragments (i.e., its reduction semantics) is formalised in terms of the state transition relation → ⊆ F × F defined in Figure 3 where F denotes the set of all terms generated by F in the grammar of the implementation. Technically, such reduction relation is defined in Structural Operational Semantics (SOS) style (i.e., by induction on the structure of the terms denoting a fragment) modulo the structural congruence relation ≡ ⊆ F × F defined in Figure 2. The reduction relation implicitly defines an unlabelled transition system.

![Figure 2: Structural congruence in FLan](image)

![Figure 3: Reduction semantics of FLan](image)

Considering terms up to a structural congruence allows us to identify different ways of denoting the same fragment. We consider the least congruence on fragments closed w.r.t. the commutativity and associativity of non-deterministic and parallel composition of processes; the associativity of sequential composition of processes; the identity of non-deterministic choice, sequential and parallel composition of processes; and the expansion of recursive process definitions. This choice of axioms (some of which may seem unusual) is not accidental. Indeed, all can be naturally and efficiently treated by Maude so that our semantics enjoys several nice properties: (1) it is (efficiently) executable; (2) each semantic rule of Figure 3 corresponds to exactly one conditional rewrite rule in the Maude implementation of FLan; (3) the number of reduction rules is small and the semantics and its implementation are thus compact and easy to read.

As usual, reduction rules are expressed in terms of a (possibly empty) set of premises (above the line) and a conclusion (below the line).

Rules Inst and Act are very similar, both allowing a process to execute an action if certain constraints are satisfied. In particular, rule Inst forbids inconsistencies due to the introduction of new features. It can be seen as a particular instance of the rule for the tell operation of concurrent constraint programming (30) instantiated as tell(has(f)).

Rule Act forbids inconsistencies with respect to action constraints. A typical case of action constraint is do(a) → has(f), i.e., action a is subject to the presence of feature f.

Rule Ask formalises the semantics of the usual ask(·) operation as known from concurrent constraint programming (30). It allows a process to be blocked until a proposition can be derived from the store.

Rule Or is quite straightforward. It allows the process to evolve as any of the branches. It is worth remarking that non-determinism can be solved at the procedural level (by relying on ask(·) actions) or at the declarative level (by using a non-deterministic choice that may be solved by the constraint store), thus providing a lot of flexibility to fragment designers (as illustrated later on).

Rules Seq and Par are standard. The former formalises the usual sequential composition, while the latter formalises an interleaving parallel composition.
The store \( S \) specifies the run-time behaviour of the coffee machine. It has explicit and declarative forms of run-time procedures. The concurrent constraint approach of FLAN ensures that all executions will end up with a consistent configuration if the process begins with a consistent store. For instance, the semantics will forbid the installation of mutually exclusive features.

Process \( R \) describes the run-time operation of the coffee machine. Depending on the country it is meant for, the machine may either accept a euro or a dollar. After that, it may \((P_2)\) or it may not \((P_3)\) be subject to a sugar regulation. The next step is the beverage selection and delivery, which may be followed by a ringtone \((P_4)\) or not, after which it returns to its initial state.

It is worth noting that \( D \) and \( R \) are not pure configurations and run-time processes. Indeed, feature \( \text{ringtone} \) is not installed by \( D \) but by \( R \), i.e. the feature \( \text{ringtone} \) is installed dynamically and it can be thought of as, e.g., a software module. This is an interesting example of a partial configuration process, where some non-mandatory features are not installed and products are only partially configured, and a run-time configurable process that installs features when needed.

In the next section, we will see that this specification has some flaws that can be spotted with our implementation in Maude. This will eventually lead to the corrected specification following from the modified parts depicted in Figure 4.

### 4. MAUDE: AUTOMATED ANALYSES

In this section, we describe some automated analysis activities supported by the implementation of our approach in Maude’s formal environment.

We illustrate the use of some of the tools in what could be a typical specification and analysis life-cycle of a product family within our framework: (i) an initial constraint store (capturing the feature constraints described in the requirements) is specified and checked for consistency; (ii) a configuration process is specified and executed step-by-step; (iii) a consistency check is performed on all possible configurations allowed by the configuration process; and (iv) the product behaviour is specified and checked with respect to its requirements (that may include temporal requirements in addition to feature constraints). We emphasise that this is only an example. The tools and techniques we illustrate can be combined and applied in many other ways.

#### Checking the (in)consistency of the initial constraints.

The store consistency check is implemented by a function \( \text{consistent} \) that, given a constraint store, returns \text{true} if the store is consistent and \text{false} otherwise. Moreover, a function \( \text{inconsistency}(S) \) can be used to spot inconsistencies in an inconsistent store. It returns the empty set if applied to a consistent store and otherwise provides a subset of inconsistent constraints of the store. This function can be used to check, e.g., the consistency of the initial store \( S \).
Checking the consistency of all configurations. Even if the semantics of FLAN preserves consistency, we can use Maude’s LTL model checker to check consistency of all reachable configurations. To do so we specify the property \[
\square \text{consistent}, \]
i.e. consistency is an invariant. Here \( \square \) denotes the temporal modality \( 	ext{always} \), and \( \text{isConsistent} \) is a state predicate that given a state (i.e. a fragment) \([S \parallel P] \) returns the result of \( \text{consistent}(S) \). We can check the property as follows.

\[
\text{reduce in ANALYSIS-KRIPKE : modelCheck}( \ ( \ [S' | D'] \ ) , \ [\square \text{isConsistent}] ) .
\]

\[\ldots\text{result Bool: true} \]

Note that the model checker is implemented as a function so we have to use command \text{reduce} to invoke it. The result is \text{true} meaning that the formula is satisfied, i.e. no inconsistent configuration is reachable.

Checking behavioural properties. After fixing the specification of the design we can analyse the run-time behaviour of the product family. What we will do in the following is to check properties of the entire product family. In general we can perform checks of the form \([S \parallel P] \models \phi \) (i.e. does \([S \parallel P] \) satisfy the LTL property \( \phi \)). A positive result means that the whole family specified by \([S \parallel P] \) satisfies the property. A false result, instead, witnesses that at least one product has at least one behaviour that does not satisfy property \( \phi \). We can check, for instance, that the run-time behaviour does not introduce inconsistencies as follows.

\[
\text{reduce in ANALYSIS-KRIPKE : modelCheck}( \ ( \ [S' | D' | R'] \ ) , \ [\square \text{isConsistent}] ) .
\]

\[\ldots\text{result Bool: true} \]

The result confirms that consistency is preserved during the run-time operation of the coffee machine.

The LTL model checker can also be used to check additional requirements, like temporal requirement 4 of our case study (“a ringtone must be rung after serving a cappuccino”).

\[
\text{reduce in ANALYSIS-LTS : modelCheck}( \ ( \ ! \ ( (\{\text{machine}\})[S' | D'] \ ; R')) \ ) , \ [\square \{\text{cappuccino}\} \rightarrow \langle \{\text{ringtone}\}\} ) .
\]

\[\ldots\text{result Bool: true} \]

The result confirms that a ringtone eventually follows (the delivery of) a cappuccino.

We may however note that the conditional statement used to accept a dollar or a euro is actually redundant due to the introduced constraints. A possible, simpler run-time process is \( R' \) (cf. Figure 5). It is very much like \( R \), but the conditional statement has been replaced by a non-deterministic choice that will be consistently solved at run-time due to the presence of the action constraints \( \text{do(euro) \rightarrow has(euro)} \) and \( \text{do(dollar) \rightarrow has(dollar)} \) in the store, which will forbid the use of the actions \text{euro or dollar} if the corresponding feature has not been installed. This time, contrary to what we did earlier for the initial store and configuration process, we are replacing procedural information by declarative information. The resulting process preserves the above temporal property, which can be checked as follows.

\[
\text{reduce in ANALYSIS-LTS : modelCheck}( \ ( \ ! \ ( (\{\text{machine}\})[S' | D' | R']) \ ) , \ [\square \{\text{cappuccino}\}
\]

\[\]
The result confirms that consistency is still preserved during the run-time operation of the coffee machine.

The above example analyses illustrate how the implementation of FLAN in Maude allows us to exploit Maude’s rich analysis toolset. In this respect, it is worth noting that in the above analyses we have made use of only a limited number of Maude tools, namely its SAT solver, its reachability analyser and its LTL model checker. There are several other Maude tools whose use remains to be investigated.

5. RELATED WORK

There is an increasing body of research on how to successfully apply automated behavioural verification techniques, like model checking, in the particular context of (software) product families. The challenge, to the best of our knowledge first recognised in [24, 25], is to develop formal and modular modelling and verification approaches which specifically take cross-cutting feature constraints into account. In this section, we discuss a number of formal methods and analysis techniques that have been applied in SPL.

There are two well-known lines of research on modelling product families in terms of extensions of LTSs, which both define family behaviour as actions (features) and use advanced model-checking techniques for the verification of behavioural properties. One makes use of extensions of Modal Transition Systems (MTSs) [13, 20, 22, 3], the other of Featured Transition Systems (FTSs) [1].

Modal Transition Systems. MTSs [21] were recognised as a suitable behavioural model for describing product families in [14]. A fixed-point algorithm, implemented in a tool, is defined to check whether an LTS conforms to an MTS with respect to several different branching relations. In the context of SPL, it allows one to check the conformance of the behaviour of a product against that of its product family.

VCM [http://fmt.isti.cnr.it/vmc/][4, 5] is a tool for modelling and analysing behavioural variability in product families modelled as MTSs [3]. VCM thus accepts a product family specified as an MTS, possibly with additional variability constraints, after which it allows the user to interactively explore this MTS; efficiently model check properties (branching-time temporal logic formulae) over an MTS; visualise the (interactive) explanations of a verification result; automatically generate one, some, or all of the family’s valid products (represented as LTSs); browse and explore these; efficiently model check whether or not products (one, some, or all) satisfy certain properties; and, finally, help the user to understand why a certain valid product does or does not satisfy specific verified properties, by allowing such a product to be inspected individually.

Featured Transition Systems. An FTS [11] is a doubly labelled transition system with an associated feature diagram. Its states are labelled with atomic propositions, while a specific distinction among its transitions is obtained by an edge-labelling defining which transitions refer to which features.

SNIP [3] is a model checker for product families modelled as FTSs specified in a language based on that of the SPIN model checker (http://spinroot.com/). Features are declared in the Text-based Variability Language (TVL) and are taken into account by the explicit-state model-checking algorithm of SPIN for verifying properties expressed in fLTL (feature LTL) interpreted over FTSs (e.g. to verify a property over only a subset of the set of all valid products). Exhaustive model-checking algorithms (which continue their search also after a violation was found) moreover allow the user to verify all products of a family at once and to output all of the products that violate a property. Unlike VMC, SNIP is a command-line tool without a GUI. SNIP, however, treats features as first-class citizens, with built-in support for feature diagrams, and it implements model-checking algorithms specifically tailored for product families.

In this paper, we proposed to specify product families in a high-level formal process-algebraic language, FLAN, which has transition systems as semantic domain. While, in principle, product family behaviour could be directly specified using transition systems from a practical point of view it is more convenient to resort to some more intuitive linguistic formalism. In fact, when used as a specification formalism, transition systems are too low level and, above all, suffer from the lack of compositionality—in the sense that they offer no means for constructing the transition system of a (sub)family in terms of that of its components. On the contrary, the process-algebraic linguistic terms offered by FLAN are more intuitive and concise notations. Using them, product families can be built in a compositional way.

Like the approach based on FTSs, we thus use a high-level language for modelling, treating features as first-class citizens, and a transition system semantics for analysis. While we currently use Maude for the automated verification of behavioural properties of product families specified in FLAN, in the future we hope to make their semantic models (LTSs, basically) amenable to model checking. FLAN is loosely inspired by the CCS-like process algebra CL4SPL presented in [15]. Unlike FLAN, however, CL4SPL has no language constructs for inter-feature constraints nor a store of constraints to separate the declarative aspects of a product family from its procedural aspects.

Adaptive Featured Transition Systems. A closely related variant of FTSs are Adaptive Featured Transition Systems (A-FTSs) [13] which were introduced for the purpose of model checking adaptive software (with a focus on software product lines). The main differentiating characteristic of A-FTSs is that the set of active features varies dynamically: features can also be deactivated, which is not possible in FLAN where we chose to guarantee monotonicity in feature activation. On the other hand, the action constraints of FLAN share some similarities with other adaptation mechanisms like those of context-oriented programming discussed and compared in [6]. To sum up, FLAN may also be seen as a language for specifying adaptive systems, which combines features (in a less flexible way than A-FTSs) and context variations.

Feature-aware verification. Tool suite SPL Verifier [1] uses standard off-the-shelf model-checking techniques to verify the absence of feature interactions by means of an approach called feature-aware verification. To this aim, the AutoFeature automata language for specifying features in separate and composable units was developed, while a variant of abstract syntax trees, called Feature Structure
Trees (FSTs), forms the basis for encoding the variability. SPLVerifier offers two methods: a brute-force one generates and verifies all valid products, while an alternative one avoids the generation of all individual products as it verifies all possible feature combinations on a single product that is purpose-built to contain all the family’s features. Like SNIP and FLAN, features are central to SPLVerifier, but only the (renowned) problem of detecting feature interactions is addressed. Unlike VMC, SNIP and FLAN, behavioural variability is not considered.

Process-algebraic approaches. A process-algebraic theory for the modelling and analysis of product families was developed also in [17–18, 23]. PL-CCS extends CCS by a variant operator that allows the user to model alternative behaviour in the form of alternative processes, with the meaning that only one of the alternative processes will exist at run-time. PL-CCS has an SOS semantics defined over multi-valued MTSs. To reason on the behaviour of product families specified in PL-CCS, a multi-valued version of the modal μ-calculus is defined, i.e. the interpretation of a logic formula over a product family no longer yields true or false, but rather a set of configurations characterising exactly those products of the family which satisfy the behavioural property under verification. Unlike FLAN, PL-CCS however does not cater for inter-feature constraints. Also, the analysis is limited to verification by model checking which is moreover not implemented.

Petri net-based approaches. The same idea underlying FTSs, namely to explicitly label the transitions of an LTS with the set of features (i.e. products) for which the transition is available, was also applied to Petri nets in [27–28], resulting in feature (Petri) nets. Larger feature nets can be constructed from smaller ones to model the addition of new features to a product family, while correctness criteria can ensure that the resulting composition preserves the original behaviour. An extension can capture the dynamic reconfiguration of products by associating to each transition of a feature net also an update expression that describes how the feature selection evolves after firing (executing) the transition. The resulting feature reconfiguration model may remain disconnected from the ordinary behavioural model, thus offering orthogonality but at the same time allowing the reconfiguration to depend upon the underlying behaviour and vice versa. This has some similarities with the combination of declarative and procedural views that is at the heart of FLAN. Efficient formal analysis and verification techniques from Petri nets of course become available to feature nets, but their application in the specific context of product families has not yet been studied.

In [32], FTSs are translated into so-called adaptable featured Petri nets, after which projection and reachability techniques from Petri nets become available for product derivation and liveness analysis.

Other approaches. In [19], FTSs (including their associated feature diagrams) are translated into Maude specifications by graph transformation. Starting from a set of requirements, this means that first a feature diagram needs to be extracted (to model the variability) and only then the desired run-time behaviour can be specified (as an FTS). FLAN, on the contrary, allows the user to combine the specification of design and run-time processes directly from a given set of requirements, which may be very convenient, for instance to specify the behaviour of partially configured or run-time configurable products. Another difference is that the semantic foundation of our approach is based on techniques from concurrent constraint programming and process algebras rather than graph transformation.

In [16], a feature-oriented approach to modelling product families in Event-B by means of a chain of refinements is explored by applying existing Event-B (de)composition techniques to two case studies, using a prototypical feature composition tool. Behavioural variability is not considered, but it would be interesting to explore the feasibility of using this Feature Event-B as a high-level specification language on top of one of the aforementioned semantic models.

6. CONCLUDING REMARKS
We have introduced the feature-oriented language FLAN as a proof of concept for specifying and analysing both declarative and procedural aspects of product families. Its semantics neatly unifies static and dynamic feature selection.

We do not envisage FLAN to become the feature-oriented language, but we advocate that some of its features are very convenient and may be adopted by existing languages.

First, we think that the concurrent constraint programming paradigm provides a flexible mechanism for separating and (when necessary) combining declarative and procedural aspects. For instance, design decisions can be delayed until run-time, which is very convenient for software product families where features may be added while the system operates. Furthermore, the run-time specification can be discharged from design decisions such as feature constraints thus resulting in light-weight, understandable specifications.

Second, the implementation of FLAN in Maude allows one to exploit the rich analysis toolset of this framework. In this paper, we have essentially restricted ourselves to its SAT solver, its reachability analyser and its LTL model checker. However, there are other Maude tools whose use may be worth investigating. The statistical model checker PVeSta, for instance, could be used for evaluating the performance of product families in variants of FLAN with stochastic and quantitative aspects.

7. FUTURE WORK
We envisage several potentially interesting extensions of FLAN. For one, we can adopt further primitives and mechanisms from the concurrent constraint programming tradition. The concurrent constraint π-calculus [8], for instance, provides synchronisation mechanisms typical of mobile calculi (i.e. name passing), a check operation to prevent inconsistencies, a retract operation to remove (syntactically present) constraints from the store and a general framework for soft constraints (i.e. not only boolean). Such features have been shown successful for the specification of service level agreements and negotiation processes [2]. This may thus turn out to be useful when product families are to be designed by cooperating partners and are hence subject to negotiation mechanisms.

Another promising line of research is to provide an FTS and an MTS semantics of FLAN so that (i) FLAN becomes a high-level language for those semantic models and (ii) we can exploit the specialised analysis tools developed for them.
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