

# Declarative and Procedural Approaches for Modelling Clinical Guidelines: Addressing Flexibility Issues

Nataliya Mulyar<sup>1</sup>, Maja Pesic<sup>1</sup>, Wil M.P. van der Aalst<sup>1</sup> and Mor Peleg<sup>2</sup>

<sup>1</sup> Eindhoven University of Technology  
GPO Box 513, NL5600 MB Eindhoven, The Netherlands  
{n.mulyar, m.pesic, w.m.p.v.d.aalst}@tue.nl

<sup>2</sup> Department of Management Information Systems, University of Haifa  
Mount Carmel, 31905, Israel  
{morpeleg}@mis.hevra.haifa.ac.il

**Abstract.** Recent analysis of clinical Computer-Interpretable Guideline (CIG) modelling languages from the perspective of the control-flow patterns has revealed limited capabilities of these languages to provide flexibility for encoding and executing clinical guidelines. The concept of flexibility is of major importance in the medical-care domain since no guarantee can be given on predicting the state of patients at the point of care. In this paper, we illustrate how the flexibility of CIG modelling languages can be improved by describing clinical guidelines using a *declarative* approach. We propose a CIGDec language for modelling and enacting clinical guidelines.

**Keywords:** Clinical guidelines, Computer-interpretable guidelines, flexibility, modelling languages, declarative model specification, temporal logic.

## 1 Introduction

Clinical Computer-Interpretable Guidelines (CIG) impact clinician behavior (i.e., quality of patient care, costs, etc.) to a great extent when they are implemented as computerized guidelines that deliver patient-specific recommendations during patient encounters [1]. A number of guideline modelling languages have been developed to represent guidelines in a machine and human understandable format that enables guideline execution. Control-flow perspective of guidelines significantly influences the clinical behavior, because it determines the order of actions in medical treatments. Other perspectives (e.g., a model of patient data including temporal data, a model of medical actions and decisions, etc.) add more contextual details to the control-flow perspective, determining the exact favorable clinical behavior. Unfortunately, due to the absence of a single standard for developing CIG modelling languages, the functionality of decision-support systems employing such modelling languages from the perspective of the control-flow differs to a great extent.

We analyzed the suitability of four modelling languages Asbru, *PROforma*, GLIF and EON for expressing control-flow patterns [2] and revealed that these languages do not offer more control-flow flexibility than process modelling languages employed by the Workflow Management Systems (WFMS) [3]. This is remarkable since one would expect CIG modelling-languages to offer dedicated constructs allowing for more build-time and runtime flexibility. Accommodating *flexibility* into guidelines means that the

CIG would be sensitive to the characteristics of specific patients and specific health care organizations [4].

The modelling languages we analyzed explicitly model a care process by specifying the steps and the order in which these steps are to be executed. Although process languages allow for some flexibility by means of modelling alternative paths, any of which could be taken depending on some a-priori available data, they are incapable of handling exceptional or unpredicted situations. Exceptional situations have to be modelled explicitly. However, modelling of all possible scenarios results in complex models and is not feasible since exceptional situations and emergencies may arise at any point in time. This makes it difficult or even impossible to oversee what activity should be performed next. To overcome these problems, i.e. reduce the complexity of models, and to allow for more flexibility in selecting an execution path, *in this paper* we propose CIGDec as a declarative language for modelling clinical guidelines. Unlike imperative languages, declarative languages specify the “what” task should be performed without determining of the “how” to perform it. CIGDec specifies by means of constraints the rules that should be adhered to by a user during a process execution while leaving a lot of freedom to the user in selecting tasks and defining the order in which they can be executed. CIGDec can be considered as a variant of ConDec [5] and DecSerFlow [6].

The remainder of this paper is organized as follows. In Section 2 we introduce CIG modelling languages Asbru, GLIF, EON and PROforma using a patient-diagnosis scenario. In Section 3 we introduce CIGDec and illustrate a CIGDec model of the patient-diagnosis scenario. We discuss the drawbacks and advantages of the proposed language in Section 4. Related work is presented in Section 5. Section 6 concludes the paper.

## 2 Computer-Interpretable Guidelines

This section describes the main concepts of four well-known CIG modelling languages: Asbru, EON, GLIF, and PROforma. These have been evaluated from the control-flow perspective using the workflow patterns [7]. We introduce the main concepts of these languages by modelling the following patient diagnosis scenario in the tools AsbruView, Protege-2000 (for EON and GLIF) and Tallis respectively. A patient is registered at a hospital, after which he is consulted by a doctor. The doctor directs the patient to pass a blood test and urine test. When the results of both tests become available, the doctor sets the diagnosis and defines the treatment strategy.

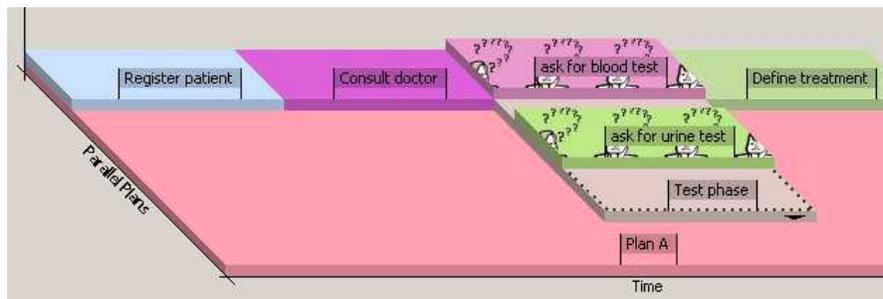
While specifying the behavior of the scenario, we immediately reflect on the possibilities to deviate from this scenario required for example in an emergency case. In particular, we indicate whether it is possible to skip a patient registration step and immediately start with the diagnosis; to perform multiple tests of the same kind or perform only one of them; and to perform the consultancy by the doctor after performing one of the tests again. Next to it, we indicate the degree of support of the control-flow patterns by the analyzed modelling languages. Table 1 summarizes the comparison of the CIG modelling languages from the perspective of the control-flow patterns [7]. The complete description of the patterns and how they are supported by the analyzed languages can be found in [8, 3] respectively.

Figure 1 presents the scenario modelled in AsbruView, which is a markup tool developed to support authoring of guidelines in Asbru [9]. A process model in Asbru

Basic Control-flow	1	2	3	4	New Patterns	1	2	3	4
1. Sequence	+	+	+	+	21. Structured Loop	+	+	+	+
2. Parallel Split	+	+	+	+	22. Recursion	+	-	-	-
3. Synchronization	+	+	+	+	23. Transient Trigger	-	-	-	+
4. Exclusive Choice	+	+	+	+	24. Persistent Trigger	-	-	+	+
5. Simple Merge	+	+	+	+	25. Cancel Region	-	-	-	-
Advanced Branching and Synchronization					26. Cancel Multiple Instance Activity	+	-	+	+
6. Multi-choice	+	+	+	+	27. Complete Multiple Instance Activity	+	-	-	+
7. Structured Synchronizing Merge	+/-	-	-	+	28. Blocking Discriminator	-	-	-	-
8. Multi-merge	-	-	-	-	29. Cancelling Discriminator	+	-	+	+
9. Structured Discriminator	+	+	+	+	30. Structured N-out-of-M Join	+	-	+	+
Structural Patterns					31. Blocking N-out-of-M Join	-	-	-	-
10. Arbitrary Cycles	-	+	+	-	32. Cancelling N-out-of-M Join	-	-	-	+
11. Implicit Termination	+	+	+	+	33. Generalized AND-Join	-	-	-	-
Multiple Instances Patterns					34. Static N-out-of-M Join for MIs	-	-	-	-
12. MI without Synchronization	-	-	-	-	35. Static N-out-of-M Join for MIs with Canc.	-	-	-	-
13. MI with a priori Design Time Knowledge	+/-	+/-	+/-	+/-	36. Dynamic N-out-of-M Join for MIs	-	-	-	-
14. MI with a priori Run-Time Knowledge	-	-	-	-	37. Acyclic Synchronizing Merge	-	-	-	+
15. MI without a priori Run-Time Knowledge	-	-	-	-	38. General Synchronizing Merge	-	-	-	-
State-Based Patterns					39. Critical Section	+	-	+	+
16. Deferred Choice	+	+	+	+	40. Interleaved Routing	+	-	+	+
17. Interleaved Parallel Routing	+	+	+	+	41. Thread Merge	-	-	-	-
18. Milestone	-	-	-	+	42. Thread Split	-	-	-	-
Cancellation Patterns					43. Explicit Termination	-	-	-	-
19. Cancel Activity	+	+	+	+					
20. Cancel Case	+	-	+/-	+					

**Table 1.** Support for the Control-flow Patterns in (1)Asbru, (2)EON, (3)GLIF, and (4)PROforma

[10] is represented by means of a time-oriented skeletal plan. Skeletal plans are plan schemata at various levels of detail, which capture the essence of the procedure, but leave room for execution-time flexibility. The root plan A is composed of a set of other plans that are represented as 3-dimensional objects, where the width represents the time axis, the depth represents plans on the same level of the decomposition (i.e. which are performed in parallel), and the height represents the decomposition of plans into sub-plans.



**Fig. 1.** The patient-diagnosis scenario modelled in Asbru

As the time axis shows, plans *Register patient*, *Consult with doctor*, *Test phase* and *Define the treatment* are executed sequentially. The *Test phase* is a parallel plan consisting of two activities *ask for urine test* and *ask for blood test*. The parallel plan requires all enclosed plans to be completed in order to pass the flow of control to the next plan. In this model, only two types of plans were used: sequential (root plan) and parallel plan (Test phase plan). In AbsruView plans of type: Any-order, Unordered, Cyclical, and If-then-else, and actions of type: Ask and Variable Assignment can be visualized.

Deviations from the modelled scenario are not possible in AsbruView, since all plans are structured and their order is strictly defined. It would be possible to adjust the model and implicitly incorporate all required execution paths. In particular, the Cyclical Plan should be used in order to iterate the execution of a certain task. In order to relax the parallel order of the blood- and urine-tests' tasks, an Any-order Plan could be used. However, the behavior of the model would be still deterministic and not allow for much flexibility. In Asbru there is a concept of plan activation mode. It allows conditions for aborting, suspending and resuming a plan. This can be relevant for the case of registering a patient and not having all the needed data initially: a plan is suspended and later resumed. As the pattern-based analysis showed [3], Asbru is able to support 20 out of 43 control-flow patterns. Asbru uniquely supports the recursive calls and interleaved parallel routing, which are the features not directly supported by other analyzed languages.

Main modelling entities in EON [11] are scenarios, action steps, branching, decisions, and synchronization [12, 13]. A scenario is used to characterize the state of a patient. There are two types of Decision steps in EON, i.e. a Case step (select precisely one branch) and a Choice step (select at least one branch). An Action step is used to specify a set of action specifications or a sub-guideline that are to be carried out. Branch and Synchronization steps are used to specify parallel execution. We omit EON model since it is very similar to model created in GLIF (see Fig. 2).

The following features offered by EON can be used in order to make the model of the patient-diagnosis scenario more flexible. A Scenario can be used to model different entry points to the model. This allows to "jump" into the middle of the model and to start execution from that point. This feature is useful for emergency cases where for example a registration step has to be skipped and immediate treatment procedure has to be started. Unfortunately, EON offers not much flexibility with respect to synchronization of multiple branches, i.e. it allows the *define treatment* task to be executed only if a single or all branches have been executed. However, it is incapable of predicting how many branches were selected and performing a partial synchronization after all selected branches were executed. From all analyzed modelling languages, EON supports the lowest number of the control-flow patterns, i.e. only 11 out of 43.

GLIF3.5 [14] is a specification method for structured representation of guidelines. To create a model in GLIF, an ontology schema and a graph widget have to be loaded into the *Protege-2000* environment. Figure 2(a) visualizes the GLIF model of the basic patient-diagnosis scenario. In GLIF3.5 five main modelling entities are used for process modelling, i.e. an Action Step, a Branch Step, a Decision Step, a Patient-State Step, and a Synchronization Step. An Action Step is a block for specifying a set of tasks to be performed, without constraints set on the execution order. It allows for including sub-guidelines into the model. Decision steps are used for conditional and unconditional routing of the flow to one out of multiple steps. Branch and Synchronization steps are used for modelling concurrent steps. A Patient-State Step is a guideline step used for describing a patient state and for specifying an entry point(s) to a guideline.

In order to allow the behavior of the basic patient-diagnosis scenario shown in Figure 2(a) to deviate, all possible paths have to be explicitly modelled. Figure 2(b) represents a scenario, in which *Register patient* step can be done in parallel to any other

step, but it has to be exactly once to complete the process (if more than once is desired, an iteration condition for *Register patient* step can be added which resembles a while loop: while not all patient data has been entered, repeat Register Patient. In this scenario, a decision can be taken to order tests or to proceed to treatment without tests. However, treatment or ordering of tests cannot be done before consulting with a doctor. One or two tests can be ordered before proceeding to treatment. Figure 2(b) shows how complex the model has become after we introduced several deviations from the basic scenario. Thus, this specification needs to model graphically all the possible paths of execution, and it is not very scalable.

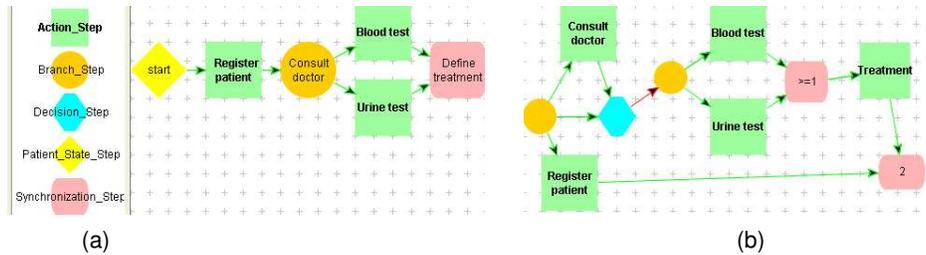


Fig. 2. The patient-diagnosis scenario modelled in GLIF3.5/Protege

Similar to EON, GLIF allows multiple entry points into the model to be specified by means of the Patient-State step. This allows the execution to start from any point where a patient enters a scenario model while skipping tasks-predecessors. GLIF offers more variants for synchronizing parallel branches, i.e. to synchronize after one, several or all tasks have been completed. However, GLIF is incapable of synchronizing branches when it is unknown which branches and how many of them will be chosen. This explains why the number of control-flow patterns supported by GLIF (17 out of 43) is bigger than in EON but still smaller than Asbru.

PROforma [15] is a formal knowledge representation language for authoring, publishing and executing clinical guidelines. It deliberately supports a minimal set of modelling constructs: actions, compound plans, decisions, and enquiries that can be used as tasks in a task network. In addition, a keystone may be used to denote a generic task in a task network. All tasks share attributes describing goals, control flow, preconditions, and postconditions. A model of the basic patient-diagnosis scenario created in Tallis is shown in Figure 3(a). Note that in PROforma control-flow behavior is captured by modelling constructs in combination with the scheduling constraints. Scheduling constraints are visualized as arrows connecting two tasks, meaning that the task at the tail of the arrow may become enabled only after the task at the head of the arrow has completed. To deviate from the basic scenario, some of the scheduling constraints should be removed as it is shown in Figure 3(b).

In contrast to all examined languages, PROforma allows for late modelling, i.e. if it is not clear in advance what steps exactly should be performed, these steps are modelled by means of keystones, which are substituted by a desired type of the task before the model is deployed. Furthermore, by means of triggers it is possible to specify that a task has to be performed even if the task's preconditions were not satisfied. PROforma also

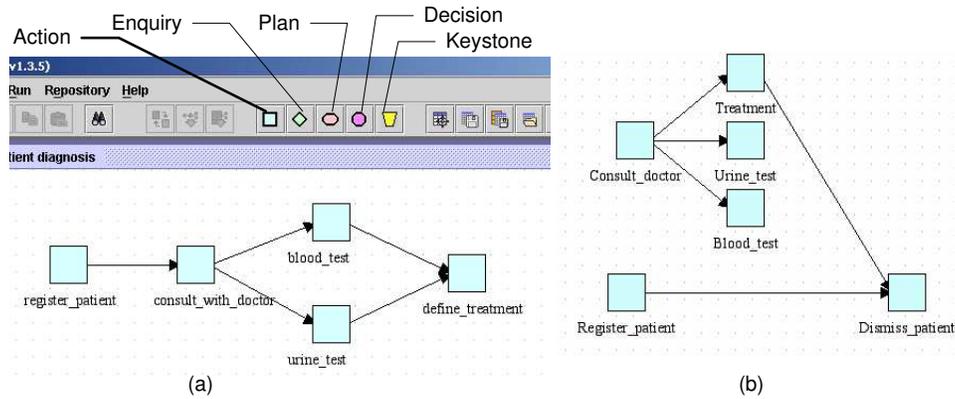


Fig. 3. The patient-diagnosis scenario modelled in PROforma/Tallis

allows for more flexibility during the synchronization of multiple paths, thus it is able to predict how many paths from the available ones were selected and to merge them when they have completed. Furthermore, scheduling constraints in PROforma are not obligatory. This means that stand-alone tasks may be activated upon the fulfillment of a pre-condition. PROforma has the highest degree of pattern-support from all analyzed languages, i.e. it supports 22 out of 43 patterns.

The medical community has always emphasized that it is impossible to use workflow formalisms because of specific requirements such as flexibility. However, when we examined guideline modelling languages we didn't find more flexibility than in today's workflow and BPM products. Given a large variety of process modelling languages nowadays it makes no sense to develop more complicated language which would support more control-flow patterns. Instead, we take a completely new approach and propose a CIGDec language for encoding clinical guidelines.

### 3 Declarative description of clinical guidelines

In this section we present the CIGDec declarative language and show benefits of applying it for modeling clinical guidelines.

Modelers who use traditional CIG modelling languages have to represent all possible scenarios (normal and exceptional) that can occur during the execution. Such a model has to include all possible scenarios that can occur during the execution. This means that CIG modelers have to predict in detail all possible execution paths in advance for the guideline they are modelling. The model itself tends to be very complex and strictly predefines all relationships between all steps in the guideline. Such a model not only prescribes to users what to do, but it also contains a detailed specification about how to do it. Hence, traditional CIG modelling languages are of an imperative nature.

CIGDec is a declarative language, i.e., its models specify what to do and leave it up to the user to decide how to work depending on the case. CIGDec models do not require all possible scenarios to be predicted in advance. On the contrary, the model consists of a set of tasks and some dependencies (relationships) between these tasks.

Dependencies between tasks can be seen as some general rules that should always hold in the guideline. Any task in the model can be performed by a user if and only if none of the specified rules is violated. As an extreme example, a CIGDec model that consists only of a set of tasks without dependencies would represent a completely free guideline, where a user can execute any task in any desired order. As more rules in the model as less possibilities to deviate from a certain execution order is given to the user. Therefore, rules constrain the model. Hence, we refer to dependencies between tasks (rules) as to *constraints*.

Any CIG model consists of a set of tasks and some relationships between them specifying the exact order of tasks. Typically, traditional languages use a predefined set of constructs that can be used to define relations between tasks: 1) sequence, 2) choice, 3) parallelism, and 4) iteration. These constructs are used to define the exact control-flow (order of tasks) in the guideline. In CIGDec, this set of constructs is unlimited, i.e., constructs can be added, changed and removed, depending on the requirements of the application, domain, users, etc. We refer to constructs used for defining possible types of dependencies between tasks in CIGDec as to *constraint templates*. Each template has its semantics, which is formally represented by one Linear Temporal Logic (LTL) formula. This semantics is used for the computerized enactment of the guideline [5]. LTL is extensively used in the field of model checking, where the target model is verified against properties specified in LTL [16, 17]. For computerized enactment of CIGDec model we use algorithms for translating LTL expressions into automata developed in the model checking field [18, 6, 5]. Since LTL formulas can be very complex and hard to understand, each template also has unique graphical representation for users. In this way, we ensure that CIGDec users do not have to be LTL experts in order to work with models [5]. Although the set of templates is ‘open’, we propose a starting collection of eleven templates in [19].

When looking at a traditional CIG model, one usually tries to find the starting point and then follows the control-flow until the end point is reached. This cannot be applied to CIGDec models. Constructs (lines) do not necessarily describe the order of tasks, but rather various dependencies between them. In our starting set of constraint templates we distinguish between two types of templates: ‘existence’ (unary) templates, and binary templates that can represent a ‘relation’ or ‘negative relation’.

‘Existence’ templates are unary templates because they involve only one task. Generally, they define the cardinality (possible number of executions) of the task. Binary templates involve two tasks. For example, a special line between tasks might mean that these two tasks include each other (e.g., ‘co-existence’ template between tasks *A* and *B* specifies that if *A* happens then *B* happens and vice versa, without specifying in which order). The ‘responded existence’ constraint specifies that if one task is performed then the other task before or after the first one. There are also some binary templates that consider the order of activities. One example is the ‘response’ template, which specifies that the a given task has to be performed at least once after the other task has been completed. Note that in all these examples it was possible to have an arbitrary execution of other tasks between the two related tasks.

### 3.1 CIGDec model for the diagnosis scenario

Figure 4 depicts a CIGDec model of our patient-diagnosis scenario. It consists of five tasks. In an extreme case, it would be possible to make and use the model consisting only out of these tasks and without any constraints. This would be a unrestricted model allowing for maximum flexibility, where tasks could be executed an arbitrary number of times ('0..\*') and in any order. This model would have an infinite number of execution possibilities (different process instances). However, to develop a model that provides guidance, we add five constraints derived from three constraint templates.

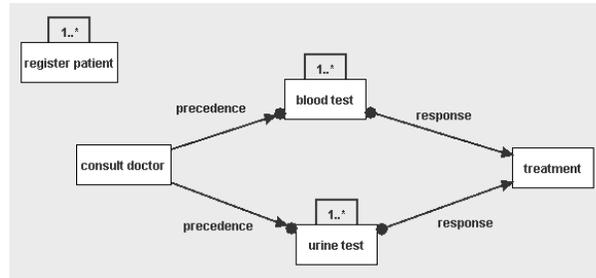


Fig. 4. CIGDec model for the diagnosis scenario.

First, there is one unary (involving one task) constraint created from the template ‘existence’ - constraint presented as cardinality  $1..*$  above the task *register patient*. This constraint specifies that the task *register patient* has to be executed at least once within one process (guideline) enactment.

Second, there are two constraints created from the template ‘precedence’ as shown in Figure 4: one between tasks *consult doctor* and *blood test* and one between tasks *consult doctor* and *urine test*. Precedence is a binary template, i.e., it defines a dependency between two tasks. A ‘precedence’ between two tasks *A* and *B* means that task *B* can only be executed after task *A* was executed at least once [6]. It is possible that other tasks are executed between *A* and *B*. Hence, if we want to execute task *blood test* we can do so only after we have executed task *consult doctor*. Note that other tasks from the model can be executed between *consult doctor* and *blood test*. Task *test urine* also has a ‘precedence’ relation with task *consult doctor* and it can be executed only after task *consult doctor*. Similarly, there could be other tasks between them. Moreover, the doctor may be consulted multiple times before and after doing the tests.

Third, we use a binary template ‘response’ to create two constraints: one between tasks *blood test* and *treatment* and one between tasks *urine test* and *treatment*. Template ‘response’ between tasks *A* and *B* defines that after every execution of task *A* task *B* has to be executed at least once while it is possible that other tasks are executed between *A* and *B*. Thus, after every *blood test* at least one *treatment* should follow, and there could be other tasks from the model executed between them. The same holds for tasks *urine test* and *treatment*.

The possibilities given to a user during execution of the model depicted in Figure 4 are defined as a combination of all five constraints in the model. When looking at the models designed by means of the analyzed language Asbru, the execution always had

to start with the task *register patient*. This may cause problems in cases of emergency, when there is no time for the registration requiring the procedure with doctor (task *consult doctor*) to start immediately. While in EON and GLIF allow multiple entry-points to a scenario, these entrance steps have to be modelled explicitly. In PROforma a task can be modelled without use of scheduling constraints which allows this task to be executed at any moment. Note however, that the CIG languages assume that a task can be executed once during the model execution or *iteratively* a specified number of times. In CIGDec model a patient-registration step can be performed at any moment during the CIGDec process. Furthermore, CIGDec model allows to perform *register patient* multiple times in case the required data is not available on time.

If we look at the traditional models Figures 1, 2 and 3 (i.e. mode using Absru and EON, GLIF and PROforma), task *consult doctor* was executed exactly once. CIGDec model allows this task not to be executed at all, but it also allows it to be executed multiple times. For example, some patients use medication periodically. For them only the *treatment* task has to be performed either before or after the *register patient* has been executed. On the other hand, in some complex cases, task *consult doctor* can be performed more than once at various points during the CIGDec execution.

If necessary, a doctor can order a *blood test* many times or not at all during the CIGDec process. However, constraint ‘precedence’ between this task and *consult doctor* makes sure that *blood test* can not be done for a patient that has not seen the doctor before. Note that this holds only for the first *blood test*. Sometimes, the results can be unexpected and doctor can order a different type of *blood test* without having to see the patient again. After every *blood test*, task *treatment* is performed. It is possible that during *treatment* no medication is prescribed due to the good test results. However, it is also possible to wait and to perform several *blood tests* in order to make an informed decision before the task *treatment* is performed. Since task *urine test* has the same relationships as task *blood test* (‘precedence’ with *consult doctor* and ‘response’ with *treatment*), the same variants of execution paths hold like for the task *blood test*. However, note that none of the tasks ‘blood test’ and ‘urine test’ do not have to execute at all, or each of them can be executed one or more times, or only one of them can be executed one or more times.

CIGDec model from Figure 4 could be used to realize the following scenarios. First, in the ‘case A’ a periodical medication is prescribed to a chronic patient: only *register patient* and *treatment* tasks are executed. In the ‘case B’ an urgent visit starts directly with *consult doctor* and only afterwards the task *register patient* is executed. The *urine test* was not necessary. The results of the *blood test* were unclear so the *treatment* is executed only after the results of the second *blood test* became available and an additional *consult doctor* task. In the ‘case C’, the situation was not urgent, so task *register patient* was performed before the task *consult doctor*. Both *urine test* and *blood test* are performed. However, due to alarming results of the *urine test* an immediate *treatment* was executed to prescribe appropriate medication. The results of *blood test* arrived later, and an additional *treatment* task was executed to handle the *blood test* results as well.

## 4 Discussion

We have shown that CIGDec can be used to define the degree of flexibility given to a user during the process execution. We have also indicated that a degree of the absolute flexibility can be reached by leaving out all constraints resulting in the freedom given to a user to select any task and execute tasks in any desired order. Since the degree of flexibility has to be controlled in the context of medical care in order to adhere to strict and desirable recommendations, the mandatory and optional constraints have to be specified for a modelled guideline. To control the adherence to the specified constraints, the execution engine CIGDec prohibits the violation of the mandatory constraints while allowing the optional constraints to be neglected. All user steps that might result in the violation of constraints are communicated to a user by means of warnings.

The advantages of the proposed CIGDec-based approach over the analyzed modelling languages that employ the imperative approach are as follows:

- CIGDec enables the *flexibility in selection*, meaning that a user executing a model specified in CIGDec gets a freedom in choosing an execution sequence, without requiring this sequence to be thought of in advance and explicitly modelled during the design-time.
- CIGDec enables *late binding*, meaning that it allows to choose an appropriate task at the point of care. This feature is particularly important in modelling of CIG since it is not always possible to predict what steps will need to be executed, thus the task selection is case-dependent.
- CIGDec ensures the *absence of change*, meaning that it prohibits choices of users that would violate mandatory constraints.
- CIGDec allows for extendability and allows new LTL formulas to be introduced, thus applicability of CIGDec could be tailored to a specific situation.

The disadvantages of using CIGDec are as follows:

- If a process to be modelled has to be very strict and should allow for flexibility, then the use of CIGDec may result in a complex model.
- CIGDec aims at the modelling of rather small processes, since the description of large processes (containing approximately several thousands of tasks) becomes difficult to understand.

Since both imperative and declarative languages have disadvantages, in order to improve the flexibility of the CIG modelling languages we recommend to augment the CIG languages with the features offered by CIGDec.

## 5 Related Work

The recent *Workflow Patterns initiative* [2] has taken an empirical approach to identifying the most common control constructs inherent to modelling languages adopted by workflow systems. In particular, a broad survey of modelling languages resulted in 20 workflow patterns being identified [7]. In this paper we have used the revised set of the control-flow patterns [8] to evaluate CIS's modelling languages.

There had been many attempts to enrich the flexibility of workflow (process) management systems. Case-handling systems are systems that offer more flexibility by focusing on the whole case (process instance), instead of individual tasks [20]. An example of such a system is FLOWer [21], where users can ‘move up and down’ the process by opening, skipping and re-doing tasks, rather than just executing tasks. Although users have a major influence on execution in FLOWer, their actions are seen as going backwards or forward in a traditional process model. Moreover, this might have some unwanted side-effects. For example, if the user wishes to execute again (re-do) an earlier task, s(he) will also have to execute again (re-do) all tasks that followed it. Unlike in FLOWer, deviations are not seen as an exception in CIGDec but as ‘normal’ behavior while the process instance unfolds further according to the choices of users.

Flexibility of process enactment tools is greatly increased by their adaptivity. ADEPT is an example of an adaptive system where users can change the process model during the enactment [22]. ADEPT is a powerful tool which enables users to insert, move and delete tasks from the process instance they are currently working on. However, the user has to be a process modelling expert in order to change the model. Moreover, in medical domain cases may have many differences and adaptations would be too frequent and time consuming. CIGDec does not see deviations as changes in the model and a good-designed CIGDec model can cover a wide variety of cases.

One of a promising ways to introduce flexibility is to replace imperative by declarative. Various declarative languages “describe the dependency relationships between tasks, rather than procedurally describing sequences of action” [23]. Generally, declarative languages propose modeling constraints that drive the model enactment [23–25]. Constraints describe dependencies between model elements. Constraints are specified using pre and post conditions for target task [25], dependencies between states of tasks (enabled, active, ready, etc.) [23] or various model-related concepts [24].

## 6 Conclusions

In this paper, we have proposed a declarative approach which could be applied to overcome problems experienced by the imperative languages used for modelling clinical guidelines. In particular, we have shown how by means of applying the CIGDec language more flexibility in selection can be achieved than the considered CIG modelling languages offer. Furthermore, we showed how the model declared in CIGDec can be enacted. In addition, we discussed differences between the proposed declarative and analyzed imperative languages, their advantages and disadvantages, and made a proposition to combine the features of imperative and declarative approaches in order to increase their applicability and usability.

## References

1. Peleg, M.: Chapter 4-2: Guideline and Workflow Models. In R.A., G., ed.: *Medical Decision Support: Computer-Based Approaches to Improving Healthcare Quality and Safety*. Elsevier (2006)
2. Workflow Patterns Home Page. <http://www.workflowpatterns.com>

3. Mulyar, N., Aalst, W., Peleg, M.: A Pattern-based Analysis of Clinical Computer-Interpretable Guideline Modelling Languages. Technical report, Center Report BPM-06-29, BPMcenter.org (2006)
4. Wald, J., Pedraza, L., Murphy, M., Kuperman, G.: Requirements development for a patient computing system. In: Proc. AMIA Symp. (2001) 731–735
5. Pesic, M., Aalst, W.: A declarative approach for flexible business processes management. In: Business Process Management Workshops. (2006) 169–180
6. Aalst, W., Pesic, M.: Specifying, discovering, and monitoring service flows: Making web services process-aware. BPM Center Report BPM-06-09, BPM Center, BPMcenter.org (2006)
7. Aalst, W., Hofstede, A., Kiepuszewski, B., Barros, A.: Workflow Patterns. Distributed and Parallel Databases **14**(1) (2003) 5–51
8. Russell, N., Hofstede, A., Aalst, W., Mulyar, N.: Workflow Control-Flow Patterns: A Revised View. BPM Center Report BPM-06-22, BPMcenter.org (2006)
9. Shahar, Y., Miksch, S., Johnson, P.: The Asgaard Project: a task-specific Framework for the application and critiquing of time-oriented clinical guidelines. *Artif Intell Med* (14) (1998) 29–51
10. Seyfang, A., Kosara, R., Miksch, S.: Asbrus Reference Manual, V.7.3. Technical report, Vienna Univ. of Techn., Inst. of SW Techn., Vienna. Report No.: Asgaard-TR-2002-1 (2002)
11. Tu, S., Musen, M.: A flexible approach to guideline modeling. In: Proc AMIA Symp. (1999) 420–424
12. Tu, S., Musen, M.: From guideline Modeling to guideline execution: Defining guideline-based decision-support Services. In: Proc AMIA Annu Symp. (2000) 863–867
13. Tu, S.: The eon guideline model. <http://smi.stanford.edu/projects/eon> (2006)
14. Boxwala, A., M.Peleg, Tu, S., O.Oqunyemi, Zeng, Q., Wang, D., et al.: GLIF3: a representation format for sharable computer-interpretable clinical practice guidelines. *Biomedical Informatics* **37**(3) (2004) 147–161
15. Fox, J., Johns, N., Rahmzadeh, A.: Disseminating Medical Knowledge: The PROforma Approach. *Artificial Intelligence in Medicine* **14**(1) (1998) 157–182
16. Jr., E.C., Grumberg, O., Peled, D.: Model Checking. The MIT Press, Cambridge, Massachusetts and London, UK (1999)
17. Holzmann, G.: The SPIN Model Checker: Primer and Reference Manual. Addison-Wesley, Boston, Massachusetts, USA (2003)
18. Giannakopoulou, D., Havelund, K.: Automata-based verification of temporal properties on running programs. In: ASE '01: Proc. of the 16th IEEE int. conf. on Automated sw engineering, Washington, DC, USA, IEEE Computer Society (2001) 412
19. Mulyar, N., Pesic, M., Aalst, W., Peleg, M.: Towards the Flexibility in Clinical Guideline Modelling Languages. Technical report, Center Report BPM, BPMcenter.org (2007)
20. Reijers, H., Rigter, J., Aalst, W.: The Case Handling Case. *International Journal of Cooperative Information Systems* **12**(3) (2003) 365–391
21. Pallas Athena: Flower User Manual. Pallas Athena BV, Apeldoorn, The Netherlands. (2002)
22. Reichert, M., Dadam, P.: ADEPTflex: Supporting Dynamic Changes of Workflow without Losing Control. *Journal of Intelligent Information Systems* **10**(2) (1998) 93–129
23. Dourish, P., Holmes, J., MacLean, A., Marqvardsen, P., Zbyslaw, A.: Freeflow: mediating between representation and action in workflow systems. In: Proc. of CSCW '96. (1996) 190–198
24. Mangan, P., Sadiq, S.: On building workflow models for flexible processes. In: ADC '02: Proc. of the 13th Australasian database conf., Australian Computer Society, Inc. (2002) 103–109
25. Wainer, J., de Lima Bezerra, F. In: Groupware: Design, Implementation, and Use. Volume 2806. Springer Berlin / Heidelberg (2003) 151 – 158