Conformance checking of mixed-paradigm process models

Citation for published version (APA):
van Dongen, B. F., De Smedt, J., Di Ciccio, C., & Mendling, J. (2021). Conformance checking of mixed-paradigm process models. Information Systems, 102, [101685]. https://doi.org/10.1016/j.is.2020.101685

Document license:
TAVERNE

DOI:
10.1016/j.is.2020.101685

Document status and date:
Published: 01/12/2021

Document Version:
Publisher’s PDF, also known as Version of Record (includes final page, issue and volume numbers)

Please check the document version of this publication:

• A submitted manuscript is the version of the article upon submission and before peer-review. There can be important differences between the submitted version and the official published version of record. People interested in the research are advised to contact the author for the final version of the publication, or visit the DOI to the publisher’s website.
• The final author version and the galley proof are versions of the publication after peer review.
• The final published version features the final layout of the paper including the volume, issue and page numbers.

Link to publication

General rights
Copyright and moral rights for the publications made accessible in the public portal are retained by the authors and/or other copyright owners and it is a condition of accessing publications that users recognise and abide by the legal requirements associated with these rights.

• Users may download and print one copy of any publication from the public portal for the purpose of private study or research.
• You may not further distribute the material or use it for any profit-making activity or commercial gain
• You may freely distribute the URL identifying the publication in the public portal.

If the publication is distributed under the terms of Article 25fa of the Dutch Copyright Act, indicated by the "Taverne" license above, please follow below link for the End User Agreement:
www.tue.nl/taverne

Take down policy
If you believe that this document breaches copyright please contact us at:
openaccess@tue.nl
providing details and we will investigate your claim.
Conformance checking of mixed-paradigm process models

Boudewijn F. van Dongen a, Johannes De Smedt b,c,∗, Claudio Di Ciccio d, Jan Mendling e

a Eindhoven University of Technology, The Netherlands
b KU Leuven, Belgium
c University of Edinburgh, United Kingdom
d Sapienza University of Rome, Italy
e Vienna University of Economics and Business, Austria

ARTICLE INFO

Article history:
Received 2 March 2020
Received in revised form 4 September 2020
Accepted 22 November 2020
Available online 26 November 2020
Recommended by G. Vossen

Keywords:
Process mining
Conformance checking
Declarative process models
Imperative process models
Mixed-paradigm process models
Hybrid process

ABSTRACT

Mixed-paradigm process models integrate strengths of procedural and declarative representations like Petri nets and DECLARE. They are specifically interesting for process mining because they allow capturing complex behavior in a compact way. A key research challenge for the proliferation of mixed-paradigm models for process mining is the lack of corresponding conformance checking techniques. In this paper, we address this problem by devising the first approach that works with intertwined state spaces of mixed-paradigm models. More specifically, our approach uses an alignment-based replay to explore the state space and compute trace fitness in a procedural way. In every state, the declarative constraints are separately updated, such that violations disable the corresponding activities. Our technique provides for an efficient replay towards an optimal alignment by respecting all orthogonal DECLARE constraints. We have implemented our technique in ProM and demonstrate its performance in an evaluation with real-world event logs.

© 2020 Elsevier Ltd. All rights reserved.

1. Introduction

Alternative representations of models and computer programs have been investigated since the 1980s. Much of this research on computer programs is driven by a distinction between declarative (which rules have to be considered) and procedural (which sequence of steps have to be taken) [1]. Empirical research on the mutual strengths and weaknesses of declarative and procedural programs has found that none of them is generally more effective, but that there are programming tasks that benefit more from one or the other [1]. This finding has been further developed into the cognitive fit theory [2]: in essence, it states that cognitive effectiveness depends on the fit between representational paradigm and task. Much of these findings has been replicated for declarative and procedural process models [3,4]. A separate conclusion from this research is the fact that some behavior can be compactly represented in a declarative way while it is complex as a procedure, and vice versa. Following the principle of minimum description length [5] inspired the idea to represent parts of the behavior in a declarative and parts in a procedural way.

The idea of mixing the declarative and procedural paradigms in single process models has been further developed into different formalizations of so-called mixed-paradigm process models [6–8], sometimes also referred to as hybrid models [9,10]. These models are proposed not only for modeling but also as a target language for process mining algorithms. Those algorithms have in common that the respective process models are composed of fragments constructed with procedural and declarative process mining techniques. The formalization of such models builds either on modular subprocesses with independent state spaces [9] or intertwined state spaces [8,11], in which the execution of activities changes the state in both model fragments simultaneously.

The formal definition of the internal mechanism of mixed-paradigm process models is a challenging task. The mentioned state–space representations have provided a foundation for mining and model checking of mixed-paradigm models [7,12,13]. The formalization of conformance checking remains an open yet critical issue for the further proliferation of mixed-paradigm approaches. Due to the specifics of their state space, previous procedural and declarative approaches like [14–16] cannot be readily applied for mixed-paradigm conformance checking (MPCC).

Still, Conformance Checking for Mixed-Paradigm processes models is required for the following reasons. First, MPCC is required to compare conformance of mixed-paradigm models and classical declarative or procedural models. Second, MPCC is required

https://doi.org/10.1016/j.is.2020.101685
0306-4379/© 2020 Elsevier Ltd. All rights reserved.
The benefits of an effective representation are highly important for process mining and automatic process discovery. Mining algorithms generally struggle to strike a balance between different quality criteria including fitness, recall, simplicity and generalization [29]. Certain behavior observed in an event log might be easier to represent though a procedural model; other parts would be best specified through a declarative model. Arguably, a discovered process model might show the best balance between the quality criteria if the more flexible behavior is captured in a declarative way and the less adaptable behavior in a procedural way. Following up on this idea in other areas of mixing paradigms [30–32], various proposals for mixed-paradigm process models have been made [6,8,33], which we discuss in turn.

Different degrees of mixing can be distinguished for mixed-paradigm process models. First, modular mixing approaches provide a mix by the help of sub-processes. Following that approach, atomic sub-processes can be modeled in either paradigm without influencing each others state space [34,35]. Arguably, the modular approach does not truly integrate both paradigms in a single model. Second, process mining using flexible mixing approaches initially discover both procedural [36,37] and declarative models [38–40] in order to then obtain a mixed-paradigm model. This approach is incorporated in Fusion Miner/FusionMINERful [7,13] and extracts mixed-paradigm models with intertwined state spaces, which we discuss next.

2.2. Discovery and representation of mixed-paradigm process models

The representation of mixed-paradigm process models combines Petri nets with DECLARE constraints. Fig. 1 illustrates a mixed-paradigm process model that is inspired by the process described in [41, Ch. 4]. The model represents a simplified version of an order-to-cash process. For the sake of readability, we depict the procedural fragment with solid black lines. The declarative constraints are drawn as dotted grey graphical elements. The procedural fragment, represented here as a Workflow Net [42], describes the sequential flow of activities. The declarative constraints specify the effects of the receipt of a request for the cancellation of the order, which can occur at any time after the start of the instance. The presence of activities that can take place at any stage of the process tend to clutter the procedural model. They typically require adding several re-routing elements to capture the diverse paths, also including reset and inhibitor arcs in Petri nets [43], or cancellation events in YAWL [44] and BPMN [45]. While those constructs are suitable to represent the run-time removal of otherwise possible actions, they are not as expressive as declarative constraints, which are naturally suitable to compactly describe such circumstances. In the following, we describe the mixed-paradigm model focusing first on the procedural fragment and then on the declarative constraints.

A run of the sole procedural part representing the sunny-day scenario begins with the receipt of a new order, followed by the assemblage of the product, the emission of the invoice, the concurrent shipment and payment of the product, and the registering of the completion. Alternative runs of the procedural part exclude the assemblage of the product, lead to the abortion of the order prior to the emission of the invoice, or register that the item was returned.

The declarative fragment of the process handles the cases in which a cancellation is requested and better clarify the conditions under which the aforementioned alternative runs may take place. The effect and the required compensation actions change according to the stage at which the cancellation occurs. If the cancellation request comes before the emission of the
invoice, it inhibits the execution of the latter activity, thus forcing
the run to proceed with the abortion of the order (the upper branch in the model): this is indicated by the NotSuccession
(Receive cancellation, Emit invoice) DECLARE constraint, graphically depicted as
\[
\text{Receive cancellation} \quad \overline{\text{NotSuccession}} \quad \text{Emit invoice}
\]

in Fig. 1. Likewise, if Receive cancellation occurs before Assembly product or Register completion, it disables those tasks (see the NotSuccession constraints in the figure). If the cancellation request comes after the product has been shipped and the payment received, the product has to be withdrawn and the payment returned. We represent this statement through a set of constraints: (1) Precedence (Receive cancellation, Withdraw product), graphically depicted as
\[
\text{Receive cancellation} \rightarrow \text{Withdraw product}
\]

imposes that the Withdraw product activity can occur only after Receive cancellation; (2) AlternativePrecedence (Ship product, Withdraw product), graphically depicted as
\[
\text{Ship product} \quad \overline{\text{AlternativePrecedence}} \quad \text{Withdraw product}
\]

states that the Withdraw product activity can occur only after Ship product, and that Withdraw product cannot recur before another product is shipped (hence the alternation); (3) We enforce that for every process instance only one item withdrawal is allowed by the AtMostOne (Withdraw product) constraint (denoted with a ... annotation beneath the activity box); by the same line of reasoning, we impose similar constraints on the execution of Return money, i.e.: (4) AlternativePrecedence (Receive cancellation, Return money); (5) Precedence (Receive payment, Return money); (6) AtMostOne (Return money); (7) finally, the pair of constraints Succession (Withdraw product, Register item return) and (8) Succession (Return money, Register item return), depicted as
\[
\text{Withdraw product} \rightarrow \text{Register item return} \rightarrow \text{Return money}
\]
together impose that if and only if the product was withdrawn and the money returned, then the item return is subsequently registered. Notice that NotSuccession (Receive cancellation, Register completion) disables the Register completion task, thus making it mandatory to execute either Abort order or Register item return, depending on the status of the process instance when the cancellation request is received.

2.3. Challenges of mixed-paradigm conformance checking

Techniques for the discovery of mixed-paradigm models have so far relied on theoretic or small examples to illustrate the capabilities of mixed-paradigm models to capture an event log in a different way compared with single-paradigm models. Therefore, it is currently not possible to quantify to what extent mixed-paradigm models are better capable of internalizing strict procedures mixed with loosely-occurring behavior in terms of model fitness. Introducing an adequate conformance checking approach, preferably founded on the same principles as other conformance techniques to ensure comparability, is hence a strong motivation for proposing an alignment-based solution for mixed-paradigm models. Furthermore, this also helps to fine-tune mixed-paradigm models. The two main mixed-paradigm discovery algorithms [7, 9] both employ parameters to determine the ratio between either paradigm – a requirement that often cannot be gauged straightforwardly. Hence, mixed-paradigm conformance checking can help in finding this ratio in terms of identifying fitting models (in combination with model checking [13]) as well as optimize the parameters. Finally, alignment-based conformance checking can aid in understanding the contribution of each paradigm to the model. By generating potential alignment issues, problems in model conformance can be pinpointed to either paradigm which can indicate that the ratio of each paradigm needs to be shifted. In general, this will be linked with either procedural parts or declarative constraints that are too restrictive to allow for replay.

Conformance checking refers to techniques that determine to which degree an event log and a process model are consistent in terms of their behavior. This consistency can be measured by the help of some criteria. For procedural approaches, measures such as fitness, precision and generalization have been defined [46–48] while declarative approaches mainly rely on support and confidence [49].

The challenge of integrating conformance checking for both paradigms root deeper than the definition of quality measures and is related to the execution semantics of intertwined state spaces. The initial approach that was presented to obtain an intertwined state space was based on conjoining Petri nets and Declare constraints on the fly [6], which is tractable for models with a small amount of constraints. Alternatively, conversions to the same execution language can be obtained. However, the conversion turns out to be often intractable [50], or resorts to approaches that are too language-specific [51,52]. The latter has caused no previous work to apply existing techniques, either procedural or declarative, to mixed-paradigm models.

In this work, we update constraints in every step of the procedural model’s state space, which is traversed by using the alignment-based conformance checking technique of [18]. That technique applies heuristics in the search space exploration resulting in an execution time that is linear with the trace size in many scenarios.

Next, we present the formal notions behind mixed-paradigm models and their fundamental constituents, namely Finite State Automata, Workflow nets and Declare.
3. Preliminaries

Mixed-paradigm models, as defined in [6,8], are composed of both procedural and declarative process modeling fragments. These fragments can be defined over the same set of activities A to obtain intertwined state spaces. We instantiate mixed-paradigm models using Petri nets and DECLARE for the following reasons. First, Petri nets are widely used for modeling and formal verification of business processes as much as for process mining. In this work, we focus on a subset of Petri nets called Workflow nets [36], which exhibit structural properties that are helpful for formal analysis. Second, DECLARE is the most prominent declarative process modeling approach and extensively used for process mining with declarative constructs.

In the following, the respective process languages are formalized and illustrated. We focus in particular on finite state automata (FSAs) (Section 3.1), Workflow nets (WFNs) (Section 3.2), and declarative process models (Section 3.3). The reader who is knowledgeable about these concepts may choose to skip this section.

3.1. Finite state automata

In order to introduce the execution semantics of declarative constraints which underpin mixed-paradigm models, we first introduce the concept of finite state automata, which serve as state-based representations of DECLARE constraints. These machines allow to express the behavior of each constraint separately and, if needed, can be conjoined into a global automaton incorporating the full behavior of a constraint model.

Definition 1 (Finite State Automaton (FSA)). A (deterministic) finite state automaton (FSA) is a finite-state labeled transition system $\mathcal{A} = (T, S, \delta, s_0, S_f)$, where: $T$ is a finite set of symbols; we shall refer to every such symbol $t \in T$ as transition; $S$ is a finite non-empty set of states; $\delta : S \times T \rightarrow S$ is the transition function, i.e., a partial function that, given a starting state and a transition, returns the target state; $s_0$ is the initial state; $S_f \subseteq S$ is the set of final (accepting) states [53].

Without loss of generality, we assume that $\delta$ is left-total and surjective on $S \setminus \{s_0\}$, that is, the transition function is defined for every pair of states and transitions, and every state is on a path from the initial one — with the possible exception of the initial state itself.

Fig. 2 depicts four FSAs. States are represented as circles and transitions as arrows. Accepting states are decorated with a double line. The initial state is indicated with a single, unlabeled arc. For instance, Fig. 2(c) is such that $T \supseteq \{t_1, t_2\}$, $s_0 = \{s_0, s_1, s_2, s_3\}$, $s_f = \{s_0, s_2\}$, $\delta(s_0, t_1) = s_1$ and $\delta(s_1, t_2) = s_2$ among others.

Definition 2 (Run of an FSA). Let $\mathcal{A} = (T, S, \delta, s_0, S_f)$ be an FSA as per Definition 1. A computation $\pi$ of $\mathcal{A}$ is a finite walk on states of $\mathcal{A}$ starting from the initial state $(s_0)$ through the consequent transitions, i.e., a sequence $\pi = (\pi_1, \ldots, \pi_n)$ of length $n \in \mathbb{N}$ of tuples $\pi_i = (s_{i-1}, t_i, s_i) \in \delta$ with $1 \leq i \leq n$ [54]. We shall name the elements of $\pi$ as steps and denote them as $s_{i-1} \xrightarrow{t_i} s_i$. For the sake of conciseness, we shall denote a computation $\pi$ from $s_0$ to $s$ via finite string $t = t_1 \ldots t_n \in T^*$ as $s_0 \xrightarrow{t} s = s_0 \xrightarrow{t_1} \ldots \xrightarrow{t_n} s$.

A computation $s_0 \xrightarrow{t} s_f$ terminating in a final state $s_f \in S_f$ is a run.

In Fig. 2(c), $\pi_1 = s_0 \xrightarrow{t_1} s_1 \xrightarrow{t_2} s_2 \xrightarrow{t_1} s_1$, $\pi_2 = s_0 \xrightarrow{t_2} s_2$, and $\pi_3 = s_0 \xrightarrow{t_1} s_1 \xrightarrow{t_2} s_2 \xrightarrow{t_2} s_2$ are computations, but only $\pi_3$ is a run because $s_2 \in S_f$ whereas $s_1, s_1 \notin S_f$. Notice that, in Figs. 2(a) to 2(d), we additionally highlight with a gray background color those states that cannot be in a step of a run, that is, from which accepting states cannot be reached.

Definition 3 (Language of an FSA). Let $\mathcal{A} = (T, S, \delta, s_0, S_f)$ be an FSA as per Definition 1 where $s_f \in S_f$ and $t = t_1 \ldots t_n \in T^*$ be a finite string, $t$ is accepted by $\mathcal{A}$ if a run $\pi = s_0 \xrightarrow{t} s_f$ exists. The set of strings accepted by $\mathcal{A}$ is the language of $\mathcal{A}$, $\mathcal{L}(\mathcal{A}) \subseteq T^*$.

For the automaton in Fig. 2(c), the language contains the string $\sigma_1 = (t_1, t_2, t_2)$ as a run exists over this sequence of transitions $(\pi_2)$, whereas $\sigma_2 = (t_2, t_2)$ is not part of the language.

In the following sections, we will use the FSAs to express the behavior of declarative constraints, and use the same set of symbols (transitions) $T$ in the mixed-paradigm models to obtain a coherent alphabet between the declarative and procedural part of the mixed-paradigm models.

3.2. Workflow nets

The procedural part of the mixed-paradigm models will be constructed out of Workflow nets. In the following, we recall the formal definition of Petri nets and, thereupon, define Workflow nets.

Definition 4 (Petri Net). A place/transition (P/T) net [55] is a triple $\mathcal{N} = (P, T, F)$ such that (i) $P$ and $T$ are disjoint sets of places and transitions, respectively, and (ii) $F \subseteq (P \times T) \cup (T \times P)$ is the flow relation. Let $\Sigma$ be an alphabet, and $\tau \notin \Sigma$ be a special symbol denoting a silent action. A labeled P/T net, hereinafter referred to as Petri net for short, is a tuple $\mathcal{P} = (\mathcal{N}, \Sigma, \ell) = (P, T, F, A, \ell)$ where (i) $\mathcal{N}$ is a P/T net, (ii) $A = \Sigma \cup \{\tau\}$, and (iii) $\ell$ is a mapping $\ell : T \rightarrow A$, henceforth also referred to as labeling function.

Graphically, places are represented as circles and transitions as boxes. The flow relation depicts arcs connecting places to transitions and vice-versa. In the model of Fig. 3, e.g., $P = \{p_1, \ldots, p_9\}$, $T = \{t_1, \ldots, t_9\}$, and $F \supseteq \{(p_1, t_1), (p_2, t_2), (p_2, t_3), (t_4, p_4), (t_4, p_5), (t_5, p_6), (t_6, p_7), (t_6, p_8), (t_6, p_9)\}$. We shall name the elements of $P \cup T$ as nodes when we do not need to distinguish places from transitions. For every node $x \in P \cup T$, the preset is defined as $\mathcal{X}(x) = \{y \mid (y, x) \in F\}$ and the postset as $\mathcal{X}^+(x) = \{y \mid (x, y) \in F\}$.
(x, y) ∈ F}. In Fig. 3, e.g., p₁• = {t₁}, t₄• = {p₄, p₅}, and •₉ = {t₇, t₈, t₉}. Labels are typically assigned to transitions to denote tasks [56]. Transitions associated with ℓ to ℓ are named silent, because of their indistinguishable and unobservable nature. They are graphically depicted as narrow black rectangles. In Fig. 3, e.g., ℓ(t₇) = c, ℓ(t₈) = r, and ℓ(t₉) = ℓ(t₁) = r as t₃ and t₇ are silent transitions. Taking inspiration by the notation of Mealy machines, we shall henceforth use the notation t/a to denote the transition and its label at once whenever needed for the sake of readability (e.g., t₂/c, t₆/r, t₅/a, t₇/t). The state of Petri nets is defined by the distribution of tokens over places, graphically depicted as solid black discs such as the one drawn in Fig. 3 inside place p₁. The state is thus represented by a marking, namely a function μ : P → N that maps places to a number of tokens [57]. A transition t ∈ T is said to be (µ-)enabled iff ∀p ∈ t, μ(p) > 0. In the example of Fig. 3, t₁ is enabled, whereas t₂, ..., t₇ are not. A µ-enabled transition may yield a follower marking μ′ of μ such that for every p ∈ P:

μ′(p) = \begin{cases} μ(p)μ(p) = μ(p)μ(p) + 1 & \text{if } p ∈ \bullet \\
μ(p)μ(p) = μ(p) & \text{if } p ∈ \bullet ∪ t \bullet \\
\text{of } p ∈ \bullet \\
\end{cases}

We say transition t fires from μ to μ′ and write μ →[t] P N μ′. In the example of Fig. 3, the firing of t₁ leads to μ′ = \{(p₂, 1)\} ∪ \{(p₁, 0) | i ∈ \{1, 3, 4, ..., 9\}\}. For the sake of conciseness, we shall use the multi-set notation for markings, indicating the number of tokens assigned by μ to p as a multiplicity of p if μ(p) > 0: considering the previous example, we shall write μ′ = \{p₁\}², p₂\}. A (finite) sequence of transitions ω = (t₁, ..., t₈) where tᵢ ∈ T for 1 ≤ i ≤ n, n ∈ N is called a (finite) firing sequence enabled at marking μ if there are markings μ₁, ..., μₙ such that μ →[t₈] P N μ₉[55]. We shall also adopt μ →[t₉] P N μ₉ as a short-hand notation. In the example of Fig. 3, (t₂, t₄) is a firing sequence enabled at marking μ′ defined above. Indeed, μ′ = \{p₃\}², p₂\}. To describe the expected run of a process modeled with a Petri net, we introduce the notions of initial marking μ₀ and final marking μ₉, recalling the notion of initial and final states of FSAs seen in Section 3.1. This leads us to the notion of Workflow net, formally defined as follows.

**Definition 5 (Workflow Net).** A Workflow net WF/N [58] is a Petri net such that (i) there exists one and only one input place p₁ ∈ P such that •p₁ = ∅, (ii) there exists one and only one output place p₉ ∈ P such that p₉• = ∅, (iii) every node x ∈ P \ U T is on a walk from p₁ to p₉, (iv) the initial marking μ₀ is such that μ₀(p₁) = 1 and μ₀(p) = 0 for every p ∈ P \ {p₁}, (v) the final marking set is a singleton M₉ = {μ₉} where μ₀(p₉) = 1 and μ₀(p′) = 0 for every p′ ∈ P \ {p₉}.

The model depicted in Fig. 3 is a Workflow net having p₁ as the input place and p₉ as the output place. Graphically, we depict tokens on the net according to the initial marking and draw the final marking with an empty disc.

A firing sequence of a Workflow net is enabled by the initial marking. A firing sequence ω is full [59] when leading from the initial marking to the final marking of the Workflow net, i.e., μ₀ → ω P N μ₉ with μ₉ ∈ M₉. Full firing sequences of the Workflow net in Fig. 3 are ω₁ = (t₁, t₂, t₄, t₅, t₆, t₇, t₈), ω₂ = (t₁, t₂, t₄, t₅, t₆, t₇, t₉), and ω₃ = (t₁, t₉, t₁₀). For brevity, we shall name full firing sequences as runs, recalling the concept of run for FSAs. An observable run ρ is the sub-sequence of ω obtained by applying an order-preserving endomorphism that removes silent transitions. In Fig. 3, the observable runs stemming from ω₁, ω₂ and ω₃ are ρ₁ = (t₁, t₂, t₄, t₅, t₆, t₈), ρ₂ = (t₁, t₄, t₅, t₆, t₇, t₈), and ρ₃ = (t₁, t₉, t₁₀), respectively, because the label of t₁ and t₇ is r. To map observable sequences to non-silent symbols, we lift the notion of labeling to strings: \(λ^* : T^* → Σ^*\). The application of \(λ^*\) to observable runs generates strings commonly known as traces in process mining. In the example of Fig. 3, the traces derived from ρ₁, ρ₂ and ρ₃ are \{o, p, i, e, s, c\}, \{o, p, i, s, e, t\} and \{o, x\} respectively. A finite multi-set of traces is an event log \(L \subseteq \Sigma^*\). An example event log for Fig. 3 is \(L = \{(o, p, i, e, s, c)^5, (o, p, i, s, e, t)^5, (o, x)^3\}\).

3.3. Declarative process models

A declarative process model represents the behavior of a process by means of constraints, i.e., rules that must not be violated during the execution of process instances. Such rules are usually exerted over activities. Constraints express norms, best practices, and behavioral patterns that restrict the possible behavior. Declarative processes are thus characterized by the fact that they define process behavior from the outside in: rather than prescribing the workflow of the process upfront, they loosely capture the possible execution scenarios to allow for maximum flexibility at run-time [24].

Declarative process modeling languages come endowed with (rule) templates on which constraints are based. Templates have a graphical representation and their semantics can be formalized using formal logics, making them verifiable and executable. Each constraint inherits the graphical representation and semantics from its template. The major benefit of using templates is that analysts do not have to be aware of the underlying logic-based formalization to understand the models. They work with the graphical representation of templates, while the underlying formulae remain hidden. To date, DECLARE is one of the most well-established declarative process modeling languages. It provides a standard library of templates (a.k.a. repertoire [60,61]), i.e., behavioral constraints parameterized over activities. In our

---

1 With \(2N\) we denote the multi-set function \(2N : X → N\) mapping the elements of set \(X\) to their multiplicity.
investigation, we will consider DECLARE but our results can be readily extended to other declarative approaches, such as DCR Graphs [62] or DPIL [27].

Taking inspiration from the formalization proposed in [63], we formally define a declarative process model as follows.

**Definition 6 (Declarative Process Model).** Let $\Sigma$ be an alphabet, and $\tau \not\in \Sigma$ a special symbol denoting a silent action. A declarative process model is a tuple $\mathcal{DP} = (R, T, \ell, \kappa, K)$ where

$R$ is a repertoire of templates, i.e., predicates $r(x_1, \ldots, x_n) \in R$ on variables $x_1, \ldots, x_n$ (we say $n \in \mathbb{N}$ is the arity of $r$),

$T$ is a finite non-empty set of transitions,

$A = \Sigma \cup \{\tau\},$

$\ell$ is a labeling function $\ell : T \rightarrow A,$

$\kappa$ is a set of constraints, namely templates of arity $n$ whose variables are assigned by a mapping with labeled transitions $x_i \xrightarrow{\kappa} t_i$ with $x_i \in T, 1 \leq i \leq n$. We shall compactly denote a constraint $\kappa \in K$ as $\kappa(x_1, \ldots, x_n).$

**Fig. 4** shows a DECLARE model that encompasses the declarative constraints of the mixed-paradigm model in **Fig. 1**. For example, $t_1 \xrightarrow{r(n)} t_4$ uses the NotSuccession template over activities $\text{Receive cancellation} (t_1/n)$ and Emit invoice $(t_4/i)$. We recall that we use $t_1/n$ as a compact notation to indicate that $t_1$ is labeled with $n$. Despite the constraints are exerted over transitions, we shall also use transition labels to denote the assigned parameters of a constraint whenever it eases the readability of examples: for instance, we shall denote $\text{NotSuccession} (t_1, t_4)$ also as $\text{NotSuccession} (t_1/n, i)$ with the single-symbol abbreviations.

Different logic-based approaches have been used to define the semantics of the DECLARE templates. In principle, Pesic et al. [34, 64] adopted Linear Temporal Logic (LTL) [65]. Their interpretation on finite traces with Linear Temporal Logic on Finite Traces (LTLf) has been later clarified by De Giacomo et al. [66,67]. In [51], DECLARE constraints are translated into equivalent Petri nets with weighted, reset and inhibitor arcs. In [13,50], regular expressions (REs) are used to define the semantics of DECLARE templates. Since REs and Monadic Second Order Logic (MSO) over finite traces have equivalent expressiveness [67,68], REs have a higher expressive power than LTLf and, as such, are a suitable language to include the formulation of DECLARE. In the remainder of this paper, we will formalize DECLARE semantics as REs.

An overview of the semantics tied to the most commonly used constraints in the DECLARE language is given in **Table 1**. As DECLARE constraints are expressible as regular expressions, their semantics are fully captured by FSAs. This leads us to the following formal definition.

**Definition 7 (Constraint Automaton).** Let $\kappa = r(t_1, \ldots, t_n) \in K$ be a constraint of a declarative process model $\mathcal{DP} = (R, T, \ell, \kappa, K)$ as per **Definition 6**. The constraint automaton of $\kappa$ is a tuple $\mathcal{A}(\kappa) = (T, S, \delta, s_0, S_T, A, \ell)$ wherein (i) $T, S, \delta, s_0$ and $S_T$ are defined as for standard FSAs (**Definition 1**), (ii) $A$ and $\ell$ are the labels set and labeling function of $\mathcal{DP}$, respectively, such that $\mathcal{A}(\kappa)$ accepts all and only those traces $\sigma \in S^*$ that satisfy $\kappa$.

Examples of algorithms that produce the automaton of a given constraint can be found in [61,69]. **Fig. 5** illustrates some constraint automata for constraints that are part of the mixed-paradigm model in **Fig. 1**. For the sake of readability, we decorate states with the constraint they refer to as a suffix (e.g., $s_0^i$ in **Fig. 5(a)**).

Inspired by the approach in [61], we classify every state $s^e$ of a constraint automaton based on the truth value of the corresponding constraint in that state:

**Permanent violation:** No computation that traverses state $s^e$ can be a prefix of a run — thus, no final state is reachable; we mark those states with the subscript na, as can be noticed in Figs. 5(a) to 5(d);

**Temporary violation:** The computations that traverse state $s^e$ can be a prefix of a run but $s^e \not\in S_T$, i.e., $s^e$ is not a final state; this is the case, for instance, of $s_4^i$ in **Fig. 5(c)**;

**Temporary satisfaction:** The computations ending in $s^e$ are runs $(s^e \in S_T)$ but can also be prefixes of computations ending in $s^e_{na}$; this is the case, e.g., of $s_5^i$ in **Fig. 5(b)**;

**Permanent satisfaction:** All computations that traverse state $s^e$ are runs — thus, no other states than final ones are reachable from $s^e$; this is the case, e.g., of $s_3^i$ in **Fig. 5(a)**.

Our framework checks for the truth value in the state of every constraint (automaton) upon the replay of a trace, as it allows us to (1) single out the constraints being violated by a trace and (2) modularize the computation of alignments between the trace and the mixed-paradigm model without resorting on a cross-product of all constraint automata, which is known to be a barely tractable operation [63].
An overview of DECLARE constraint templates with their corresponding regular expressions, and verbose descriptions.

| Template                | Regular expression | Description                                                                 |
|-------------------------|--------------------|-----------------------------------------------------------------------------|
| EXISTENCE(a, n)         | \( + (a, + ) \{n\} \) | Activity a happens at least n times.                                        |
| ABSENCE(a, n)           | \([-a] + (a? [-a] + ) \{n\} \) | Activity a happens at most n times.                                         |
| EXACTLY(a, n)           | \([-a] + (a[-a] + ) \{n\} \) | Activity a happens exactly n times.                                         |
| INIT(a)                 | \( a, + ? \)         | Each instance has to start with activity a.                                 |
| END(a)                  | \( . + a \)          | Each instance has to end with activity a.                                   |
| RESPONDEDEXISTENCE(a, b) | \([-a] \star ((a, + b, + ) \{b, + a, + \}) \) | If a happens at least once then b has to happen or happened before a.       |
| COEXISTENCE(a, b)       | \([-ab] \star ((a, + b, + ) \{b, + a, + \}) \) | If a happens then b has to happen or happened after a, and vice versa.     |
| RESPONSE(a, b)          | \([-a] \star (a, + b, + ) \{-a\} \) | Whenever activity a happens, activity b has to happen eventually afterward. |
| PRECEDENCE(a, b)        | \([-b] \star (a, + b, + ) \{-b\} \) | Whenever activity b happens, activity a has to have happened before it.    |
| ALTERNATERESPONSE(a, b) | \([-a] \star (a[-a] + b[-a] + ) \) | After each activity a, at least one activity b is executed. A following activity a can be executed again only after the first occurrence of activity b. |
| ALTERNATEPRECEDENCE(a, b) | \([-b] \star (a[-b] + b[-b] + ) \) | Before each activity b, at least one activity a is executed. A following activity b can be executed again only after the first next occurrence of activity a. |
| CHAINRESPONSE(a, b)     | \([-a] \star (ab[-a] + ) \star \) | Every time activity a happens, it must be directly followed by activity b (activity b can also follow other activities). |
| CHAINPRECEDENCE(a, b)   | \([-b] \star (ab[-b] + ) \star \) | Every time activity b happens, it must be directly preceded by activity a (activity a can also precede other activities). |
| NOTCOEXISTENCE(a, b)    | \([-ab] \star ((a[-b] + ) \{b, -a, + \}) \) | Either activity a or b can happen, but not both.                            |
| NOTSUCCESSION(a, b)     | \([-a] \star (a[-b] + ) \star \) | Activity a cannot be followed by activity b, and activity b cannot be preceded by activity a. |
| NOTCHAINSUCCESSION(a, b) | \([-a] \star (a[-ab] \{a[-a] + ) \star \star \) | Activities a and b can never directly follow each other.                    |
| CHOICE(a, b)            | \( . + [ab], . \)    | Activity a or activity b has to happen at least once, possibly both.       |
| EXCLUSIVECHOICE(a, b)   | \( [b] \star [a[-b] + ] \star \star \) \( [ab], . + [a[-a+ b[-a] + \star \star \) | Activity a or activity b has to happen at least once, but not both.        |

Fig. 5. Constraint automata derived from some constraints of the model in Fig. 1.

4. Approach

In this section, we describe the approach to align mixed-paradigm models with event logs, together with examples that illustrate the findings. We begin with the formalization and illustration of mixed-paradigm models and their alignments with the traces of an event log. Thereupon, based on existing techniques for aligning traces with procedural models, two new techniques are proposed to deal with mixed-paradigm models. First, we show how we can compute alignments while guaranteeing that all constraints are satisfied at the end of the trace. Second, we show how to allow for the violation of constraints if that leads to a better explanation of the trace.

4.1. Mixed-paradigm models

Based on the above definitions, we can now introduce the notion of mixed-paradigm models. Fig. 6 depicts a mixed-paradigm process model (notice that its structure is the same as that of Fig. 1 with one-letter abbreviations for the activities). A mixed-paradigm process model consists of (i) a Workflow net and (ii) a finite set of DECLARE constraints. Other mixed-paradigm solutions
Definition 8 (Mixed-Paradigm Model). A mixed-paradigm model is a tuple $\mathcal{MP} = (\mathcal{WFN}, \mathcal{DP})$, where

$\mathcal{WFN} = (P, T, F, A, \ell, \mu_0, M_\mathcal{F})$ is a Workflow net (as per Definition 5).

$\mathcal{DP} = (R, T, A, \ell, K)$ is a declarative process model (as per Definition 6), such that $\mathcal{WFN}$ and $\mathcal{DP}$ share the same transitions set $T$, labels ($A = \Sigma \cup r$) and labeling function ($\ell$).

We assume that Workflow nets and Declare models are defined over the same set of activities so that the models’ state spaces are synchronized during the execution. The separation of the mixed-paradigm model in its procedural and declarative parts, as well as the fact that the two fragments of the model share the same transitions and respective labels, can be observed by comparing Fig. 6 with Figs. 3 and 4. Without loss of generality, we take Declare as the repertoire of $\mathcal{DP}$.

A workflow run (or simply run, for short) of a mixed-paradigm model is a sequence of transitions that is a run to the Workflow net (i.e., it leads from its initial marking to the final one). If the run ends in a configuration that satisfies every constraint, it is a full run of the mixed-paradigm model. We formalize these notions as follows.

Definition 9 (Run of a Mixed-paradigm Model). Let $\mathcal{MP} = (\mathcal{WFN}, \mathcal{DP})$ be a mixed-paradigm model composed of a Workflow net $\mathcal{WFN} = (P, T, F, A, \ell, \mu_0, M_\mathcal{F})$ and a declarative process model $\mathcal{DP} = (R, T, A, \ell, K)$ as per Definition 8. Let $m = |K|$ be the number of constraints in $\mathcal{DP}$ and $\alpha(k) = (T, S, \delta, s_0, \delta_0, A, \ell)$ be the constraint automaton of $k \in K$ as per Definition 7. A computation $\Pi$ of $\mathcal{MP}$ is a finite sequence of steps $(\Pi_1, \ldots, \Pi_n)$ of length $n \in \mathbb{N}$ of tuples $\Pi_i = \left(\mu_{i-1}, \frac{\ell}{\alpha(k)_{i-1}} \mu_{i}, s_{i-1}^{m_{i}} \frac{\ell}{\alpha(k)_{i}} s_{i}^{m_{i}} \right)$ for $1 \leq i \leq n$, starting at the initial marking $\mu_0$ of $\mathcal{WFN}$ and the initial states $s_0^{\mathcal{F}}, \ldots, s_0^{\mathcal{M}}$ of all constraint automata $\alpha(k_1), \ldots, \alpha(k_m)$ of the constraints in $\mathcal{DP}$. A computation $\Pi$ of length $n$ having as its last step $\Pi_n = \left(\mu_{n-1}, \frac{\ell}{\alpha(k_n^{\mathcal{F}})}, s_{n-1}^{m_{n-1}} \frac{\ell}{\alpha(k_n^{\mathcal{M}})} s_{n}^{m_{n}} \right)$, such that $\mu_n \in M_\mathcal{F}$ is the final marking of $\mathcal{WFN}$, is a workflow run (or simply run, for short) of $\mathcal{MP}$. Let $s_i^{k}$ be the tuple $(s_i^{1}, \ldots, s_i^{m})$ of states of the constraint automata $\alpha(k_1), \ldots, \alpha(k_m)$ at the $i$th step, $S_i^n$ be the union of the sets of accepting states of $\alpha(k_1), \ldots, \alpha(k_m)$, and $S_0^n$ the set consisting of their initial states. If $\Pi$ is a run and $s_0^{k_1}, \ldots, s_0^{k_m}$ are final states of $\alpha(k_1), \ldots, \alpha(k_m)$, i.e., $s_0^{k} \in S_0^n$, then $\Pi$ is a full run of $\mathcal{MP}$.

In essence, a step of a mixed-paradigm model is an ensemble of firings in the procedural fragment and of a step in each of the constraint automata of its declarative fragment. Considering the example model of Fig. 6, by replaying $(t_1, t_2, t_3, t_4)$ and $(t_1, t_2, t_4, t_5, t_6, t_7, t_8, t_9, t_{10})$ we attain two full runs. By replaying $(t_1, t_2, t_4, t_5, t_6, t_7, t_8, t_{12}, t_{13})$ we have a computation: notice that the constraint automata of Succession $(t_{12}, t_6)$ and Succession $(t_{13}, t_6)$ are not in a final state (both are in a state of temporary violation, waiting for $t_9$ to occur), and marking $[p_1]$ is not final. Finally, $(t_1, t_2, t_4, t_5, t_6, t_7, t_8, t_{12}, t_7, t_9)$ yields a workflow run but not a full run because, although the final marking is reached, the constraint automata of Succession $(t_{12}, t_6)$ and Succession $(t_{13}, t_6)$ are not in a final state (temporary violation) and the constraint automaton of NotSuccession $(t_{11}, t_8)$ is in a state of permanent violation.

Notice that the state space exploration of these models is not trivial, as in a next state the workflow net might still be capable of reaching a final marking, but might not be able to reach an accepting state for all the constraints' automata any longer. Although we can use constraints to reduce the state space of the workflow net, in every state we need to check which transitions can still be fired in order to guarantee that the constraints can be brought to an accepting state in any future state. Consider for example the firing sequence $\varphi_1 = (t_1, t_2, t_4, t_{11}, t_2, t_3, t_6, t_7)$ over Fig. 6. It leads to a marking with a single token in $p_9$. However, $t_8/5$ is not enabled to fire due to the NotSuccession $(n, c)$ constraint and the only possible next firing is that of $t_9/\ell$.

An initial approach to calculate the future satisfiability of constraints on-the-fly was presented in [6]. However, this approach might prove intractable in case of larger sets of constraints and is not needed for conformance purposes. By using alignments we avoid a full state space exploration by iteratively polling queues of future states as will be discussed in Sections 4.4 and 4.5, where we also explore the impact of leaving constraints to be violated (at a cost).

We recall that, as this paper focuses on mixed-paradigm models consisting of (Declare) constraint automata and Petri nets, other mixed-paradigm approaches can readily apply these insights. Indeed, many procedural models can be converted into Petri nets (e.g., EPCs, YAWL, and BPMN). Secondly, the use of FSAs allows the use of an array of other declarative constraints.
Aside from DECLARE, e.g., DCR Graphs can be converted into automata [26]. Also, notice that the translation from DECLARE constraints to BPMN-D fragments [52] is based on the production of FSAs as an intermediate step.

4.2. Alignments

Conformance checking has been tackled in many ways [70], using concepts such as token replay [14], behavioral profiles [48], alignments [15], negative events [47], or topological entropy [71], among others. A general framework for introducing probabilistic weights to either recordings in the model and in the event log was proposed in [72]. While each of these approaches has its merits, the alignment-based approach [17,18] is elaborated and used in the rest of the paper for its overall efficiency. An alignment is a sequence of moves that synchronizes the computation of a model and a log’s trace. As we distinguish a (workflow) run from a full run of a mixed-paradigm model, according to whether the declarative constraints are satisfied or not at the end, so we make a distinction between a (workflow) alignment and a full alignment in the following.

Definition 10 (Alignment). Let \( \mathcal{MP} = (\mathcal{VfN}, \mathcal{DP}) \) be a mixed-paradigm model as per Definition 8. Let \( \sigma \in \Sigma^* \) be a sequence \( (\sigma_1, \ldots, \sigma_n) \) of length \( n \in \mathbb{N} \) and \( \gg \) be the symbol that denotes the skipping of an activity (i.e., no move is performed). Let \( T^\gg \) denote \( T \cup \{\gg\} \) and \( A^\gg \) denote \( A \cup \{\gg\} \). Then, a workflow alignment (or alignment for short) of \( \sigma \) and \( \mathcal{MP} \) is a sequence of moves \( \gamma = (\gamma_1, \ldots, \gamma_m) \in (T^\gg \times A^\gg)^* \) with \( \gamma_i = (t_i, a_i) \) for \( 1 \leq i \leq m \), with \( m \geq n \in \mathbb{N} \), between the trace \( \sigma \) and the mixed-paradigm model \( \mathcal{MP} \) if and only if:

1. For each \( 1 \leq i \leq m \) it holds that:
   - \( \gamma_i \) is a move on model: \( t_i \in T, a_i = \gg \) (with \( t \)-move being the special case where \( t(t) = t \)),
   - \( \gamma_i \) is a move on log: \( t_i = \gg \), \( a_i \in A \), or
   - \( \gamma_i \) is a synchronous move: \( t_i \in T, a_i \in A \);

2. By applying an order-preserving endomorphism that removes moves on model, we obtain \( \gamma’ = (\gamma’_1, \ldots, \gamma’_m) \) with \( m’ \leq m \) such that its sequence of events \( (a’_1, \ldots, a’_{m’}) \) is the trace \( \sigma’ \);

3. By applying an order-preserving endomorphism that removes moves on log, we obtain \( \gamma’’ = (\gamma’’_1, \ldots, \gamma’’_{m’’}) \) with \( m’’ \leq m \) such that its sequence of transitions \( (t’’_1, \ldots, t’’_{m’’}) \) is a run for \( \mathcal{VfN} \), i.e., \( \gamma \) induces a workflow run of \( \mathcal{MP} \) as per Definition 9.

If \( \gamma \) is a workflow alignment that induces a full run of \( \mathcal{MP} \) as per Definition 9, then \( \gamma \) is a full alignment.

Given a cost function for alignments, an optimal (workflow or full) alignment is defined as an alignment that minimizes this function, as we formalize in the following.

Definition 11 (Optimal Alignment). Let \( \mathcal{MP} = (\mathcal{VfN}, \mathcal{DP}) \) be a mixed-paradigm model as per Definition 8 and \( \sigma \in \Sigma^* \) be a sequence \( (\sigma_1, \ldots, \sigma_n) \) of length \( n \in \mathbb{N} \). Given a function cost : \( (T^\gg \times A^\gg)^* \rightarrow \mathbb{R}^+ \), an alignment \( \gamma^{op} \in (T^\gg \times A^\gg)^* \) of \( \sigma \) and \( \mathcal{MP} \) is an optimal alignment if and only if, for any other alignment \( \gamma \in (T^\gg \times A^\gg)^* \), it holds true that \( \text{cost}(\gamma^{op}) \leq \text{cost}(\gamma) \).

Notice that there may be more than one optimal alignment for any given sequence of events. The default cost function for alignments is based on a function \( c : T^\gg \times A^\gg \rightarrow \mathbb{R}^+ \) associating 0 to every synchronous move and \( t \)-move, and 1 to every move on log or model. The default total cost of the alignment, then, is the sum of the partial costs on the moves.

Consider for example the model in Fig. 6 and a trace \( (o, p, n, i, n) \), i.e., a trace showing that an order was received \((o)\) and a product assembled \((p)\), then a cancellation was received \((n)\), the invoice emitted \((i)\) and another cancellation received \((n)\). If we assume a standard cost function, an optimal alignment for reaching the final marking \( m_f = (p) \) from the initial marking \( m_0 = (p) \) is \( \gamma^{op} = ((t_1, o), (t_2, p), (t_{i1}, n), (\gg, i), (t_{i1}, n), (t_{i0}, \gg)) \). This alignment is full as it ensures that all constraints are satisfied at the end of the execution of the model. This alignment shows that the Emit invoice activity \((i)\) cannot happen in the model, even though it is in the event log. This is because of the Discount (Receive cancellation, Emit invoice) constraint. Finally, in order to reach the final marking, the model needs to execute theAbort order transition \((s)\) which is not part of the event log.

Now consider another trace, \((o, p, i, s, e, m, n)\). In this trace, an order is received \((o)\), a product assembled \((p)\), an invoice emitted \((i)\). The product is then shipped \((s)\) and paid \((e)\) and finally, money is returned \((m)\), a cancellation is received \((n)\) and the process stops. Again using the standard cost function, an optimal alignment would be \( \gamma^{op} = ((t_1, o), (t_2, p), (t_{i4}, i), (t_{i6}, s), (t_5, e), (t_7, \gg), (\gg, m), (\gg, n), (t_9, \gg)) \) with three deviations, i.e., the move on log for returning the money, the move on log for the cancellation, and a move on model on Register item return \((t_6)\). This alignment suggests that the model cannot explain that the cancellation took place and money was returned, and that, additionally, the case completed regularly although it was not observed in the log.

Sometimes, however, we could prefer an explanation stating that the cancellation did take place, money was returned and then the item return was registered, but the constraint Discount (Receive cancellation, Return money) is violated at the end of the case. The alignment in that case would be \( \gamma^{opt} = ((t_1, o), (t_2, p), (t_{i4}, i), (t_{i6}, s), (t_5, e), (t_{i2}, m), (t_{i1}, n), (t_{i3}, \gg), (t_7, \gg), (t_9, \gg)) \) again with three deviations, i.e., the synchronous move \((t_{i2}, m)\) which breaks the Discount constraint and the two model moves on Withdrew product after cancellation and on Register item return, indicating that these tasks have not happened yet, but are required to happen to complete the process.

In the next section, we show that existing techniques for finding alignments can be modified to solve both types of problems, i.e., finding alignments that guarantee that all constraints are satisfied and finding alignments in which constraints may be violated at the end of the trace.

4.3. Computing workflow alignments

Computing alignments for a trace and a procedural model is a non-trivial task. Many techniques exist in literature and for a full overview, we refer to [73]. All search-based techniques work in a similar fashion. The initial state used by the search algorithm is the initial marking of the model and the beginning of the trace to replay. From that state on, all successor states are computed by executing all possible moves, i.e., moves on model, moves on log or synchronous moves. This leads to a number of successor states which are then queued for further investigation.

Several techniques have been proposed to decide which states to investigate in which order. In [74] an overview of parameters is provided which influence this order. However, the search in general relies on a fundamental element: the estimation of the remaining cost. The underlying search algorithm for alignment computation is \( A^* \). This algorithm works best in the presence of a heuristic which underestimates the remaining costs of replaying the (remainder of the) given trace from the current marking. Such an estimator exists for Petri nets in the form of a marking equation that abstracts from the order in which transitions are fired to get an estimate.
Algorithm 1: Workflow alignment.

1 AlignTraceAndModel \( \langle \sigma = \{q_1, \ldots, q_m\} \in \Sigma^* \times MP, c \rangle \) input: A trace \( \sigma \); a mixed-paradigm model \( MP = (\mathcal{V} \mathcal{F}N, \mathcal{D} \mathcal{P}) \); a partial cost function \( c : (\mathcal{P} \times \mathcal{T} \times MP) \rightarrow \mathbb{R} \).

output: An alignment \( y \).

2 i ← 1;

3 \( \mu \leftarrow q_1 \); // \( \mu \) keeps track of the state in the workflow net \( \mathcal{V} \mathcal{F} \).

4 \( g \leftarrow 0 \); // \( g \) keeps track of the cost so far

5 \( h \leftarrow -\text{estimateRemainingCost}(\sigma, \mathcal{V} \mathcal{F}N, \mu, i, c) \); // \( h \) underestimates the remaining cost

6 \( p \leftarrow \text{NULL} \); // \( p \) stores the predecessor of the current state

7 \( q \leftarrow (\mu, i, g, h, p) \); // \( q \) is a queue of states to investigate;

8 while \( i < m \) \& \& \( \mu \neq \mu_D \) do

9 \( \text{Enabled} \leftarrow \text{getEnabledTransitionsPN}(\mathcal{V} \mathcal{F}N, \mu) \); // enabled moves on the workflow net.

10 for \( r \in \text{Enabled} \) do

11 \( \mu' \leftarrow \langle \rho, r, \lambda \rangle \); // \( \mu' \) is the new state

12 if \( h' \geq h \) then // \( h' \) is the estimated cost of \( r \).

13 \( \mu' \leftarrow \text{estimateRemainingCost}(\sigma, \mathcal{V} \mathcal{F}N, \mu, i, c) \);

14 \( Q \leftarrow \text{enqueue}((i + 1, \mu', g, h, q)) \); // Add new state to the queue

15 \( g \leftarrow g + c(\tau, r) \); // Compute next state for log move

16 \( h' \leftarrow \text{estimateRemainingCost}(\sigma, \mathcal{V} \mathcal{F}N, i + 1, \mu, c) \);

17 \( Q \leftarrow \text{enqueue}((i + 1, \mu, g', h', q)) \); // Add new state to the queue

18 \( (i, \mu, g, h, p) \leftarrow \text{pullNextBestState}(Q) \); // Pull next state to investigate

19 end for

20 end while

21 return \( y \).

Algorithm 2: Full alignment.

1 AlignTraceAndModel \( \langle \sigma = \{q_1, \ldots, q_m\} \in \Sigma^* \times MP, c \rangle \) input: A trace \( \sigma \); a mixed-paradigm model \( MP = (\mathcal{V} \mathcal{F}N, \mathcal{D} \mathcal{P}) \); a partial cost function \( c : (\mathcal{P} \times \mathcal{T} \times MP) \rightarrow \mathbb{R} \).

output: An alignment \( y \).

2 i ← 1;

3 \( \mu \leftarrow q_1 \); // \( \mu \) keeps track of the index in the trace

4 \( g \leftarrow 0 \); // \( g \) keeps track of the cost so far

5 \( h \leftarrow -\text{estimateRemainingCost}(\sigma, \mathcal{V} \mathcal{F}N, \mu, i, c) \); // \( h \) underestimates the remaining cost

6 \( p \leftarrow \text{NULL} \); // \( p \) stores the predecessor of the current state

7 \( q \leftarrow (\mu, i, g, h, p) \); // \( q \) is the current head of the queue

8 while \( i < m \) \& \& \( \mu \neq \mu_D \) do // Do not stop if there is an unsatisfied constraint in the queue.

9 \( \text{Enabled} \leftarrow \text{getEnabledTransitionsPN}(\mathcal{V} \mathcal{F}N, \mu) \); // enabled moves on the workflow net.

10 for \( r \in \text{Enabled} \) do

11 if \( c(\tau, r) \neq 0 \) then // Transition \( \tau \) cannot lead to a non-accepting state.

12 \( \mu' \leftarrow \text{nextState}(\sigma, \mu, i, r) \); // Compute next state in the set.

13 \( g' \leftarrow g + c(\tau, r) \); // Compute next state for log move

14 \( h' \leftarrow \text{estimateRemainingCost}(\sigma, \mathcal{V} \mathcal{F}N, i, \mu', c) \);

15 \( Q \leftarrow \text{enqueue}((i + 1, \mu', g', h', q)) \); // Add new state to the queue

16 \( (i, \mu, g, h, p) \leftarrow \text{pullNextBestState}(Q) \); // Pull next state to investigate

17 end for

18 if \( i > n \) then // Stop condition.

19 \( g' \leftarrow g + c(\tau, r) \); // Compute next state for log move

20 \( h' \leftarrow \text{estimateRemainingCost}(\sigma, \mathcal{V} \mathcal{F}N, i, \mu, c) \);

21 \( Q \leftarrow \text{enqueue}((i + 1, \mu, g', h', q)) \); // Add new state to the queue

22 \( (i, \mu, g, h, p) \leftarrow \text{pullNextBestState}(Q) \); // Pull next state to investigate

23 end if

24 return \( y \).

Algorithm 1 describes in pseudo-code how our algorithm yields a workflow alignment — thus, not considering the satisfaction of constraints. States are stored in a priority queue and investigated one by one. In each iteration, new successor states are added to a priority queue and this queue is polled in further iterations. Note that we abstract from the closed set here, yet no state has to be visited more than once and no state has to be represented in the queue more than once (for algorithmic details, we refer to [75]). Notice also that the algorithm assumes that at least one alignment exists.

Let us consider the trace \( \langle o, p, i, s, e, s \rangle \) and the Workflow net of Fig. 3. Fig. 7 depicts the entire search space considered by the A* algorithm in Alg. 1. For clarity of presentation, the moves are identified using the transition labels, not the identifiers. The color scheme we use follows the typical scheme of alignment-based conformance checking tools. The moves on model are depicted top-down in purple and the moves on log are depicted in yellow from left to right. The green synchronous moves are also shown and, for each node, the shortest distance from the top-left node is written. \( \tau \)-moves are drawn in dark gray. Finally, the dashed brown line shows a shortest path in this graph, which corresponds to the following alignment: \( y_{\text{opt}} = (t_1, o, (t_2, p, (t_4, i, t_q, e, s, (t_7, \gg, t_9, \gg, \gg, t_3, \gg, \gg ) ,\gg, s)) \) with cost 2. The A* algorithm will not expand the full search space, but because of the quality of the estimation function, it will in this case only enqueue the nodes on this shortest path and their direct neighbors. Without this estimation function, however, all nodes with cost less or equal to 3 will be queued before this shortest path is returned.

The two techniques presented in this section adopt the existing search-based techniques in two ways. First, we provide a technique that guarantees that all constraints are satisfied at the end of alignment. Second, we provide a technique that allows for the violation of constraints using a cost for doing so, similar to the cost for moves on model and log.

4.4. Computing full alignments

Consider a mixed-paradigm process model which consists of a procedural model and additional declarative constraints on the transitions in this model. As we have seen, the model without any constraints is a straightforward procedural model. To calculate alignments for this model, we can simply use the existing A*-based search techniques.

The declarative constraints added to that model cause its language to shrink, i.e., certain traces are no longer part of the language as they violate one or more constraints (see Section 4.1). To yield a full run, namely the run that ends in a final marking for the workflow net and in the final states of constraint automata (Definition 9), we modify the algorithm by only changing Line 10 of Alg. 1 to check for the admitted transitions as well as Line 9 for the termination condition. With these modifications in place, we obtain a full alignment (Definition 10).

The algorithm is described in Alg. 2. Notice that it is only slightly different than the base algorithm. For each automaton, we need to keep track of the state in which the automaton is. In Line 9, we check if any of the automata could end up in a state of permanent violation \( s_{\text{FA}} \). If so, this transition cannot be executed and the algorithm continues with the next transition.

By ensuring that (a) no state of permanent violation is ever reached and (b) the algorithm only terminates if all constraints are satisfied, the resulting alignment algorithm is correct if and only if the estimate of the remaining cost is still an underestimate. This is trivially true since it is true for the larger language of the procedural model without constraints [76].

Let us consider the trace \( \langle o, p, i, s, e, \sigma, n, m, n \rangle \) again — we already discussed it at the end of Section 4.2. Recall that, in this trace, an order is received \( t_1\langle o \rangle \), a product assembled \( t_2/m \), an invoice emitted \( t_6/e \). The product is then shipped \( t_8/s \) and paid \( t_9/e \). Finally, money is returned \( t_{12}/m \) after which a cancellation is received \( t_{13}/n \) and the process stops.
state represents the case where cancellation was received after emitting the invoice. This part of the model requires handling the withdrawal of the product and the returning of the money as well as the registration of the item’s return. Fig. 9 shows the worst case expansion of the search space for Alg. 2 when aligning the model of the running example of Fig. 1 to the trace \( (o, n, n) \), i.e., the trace in which, after receiving an order, the order is canceled twice by the customer. The shortest path is highlighted and corresponds to the alignment \( y^{opt} = \langle t_1, o, (t_1, n), (t_3, \gg), (t_1, n), (t_{10}, \gg) \rangle \) with cost 1 for the move on model of Abort order.

The full search space for the \( A^* \) algorithm for trace \( (o, p, i, s, e, m, n) \) contains \( 27 \times 8 = 216 \) states and is structured in the same way as Fig. 7, i.e., with synchronous moves and moves on log between identical copies of the model’s state space.

Let us consider all the constraints in the model. There are 11 constraints in total: \( \kappa_1 \) to \( \kappa_{11} \). Their respective automata are depicted in Fig. 10. The initial state of the search algorithm puts all these automata in the initial state \( s_0 \). Then, in Line 12 of Alg. 2, we make sure that the current transition does not lead to any of the non-accepting states \( s_{na} \) in any of the automata. If it does not, then this transition can be executed; otherwise, it cannot be executed.

Assume we have reached a state in the search where we have explained the fitting prefix of the sequence, i.e., we executed transitions \( t_1/o, t_2/m, t_4/i, t_6/s \) and \( t_5/e \) in the model. The next event to explain is Return money. In the model, the marking is \( \{p_1^y, p_2^x\} \) and, as shown in Fig. 10, the automata are in their initial state, except for the automata of \( \kappa_3 \) and \( \kappa_5 \) which are in states \( s_3^x \) and \( s_3^y \) respectively. The corresponding transition \( t_{13} \) for task \( m \) is enabled according to the workflow model. However, if we execute this transition, automaton \( \kappa_4 \) would transition from state \( s_0^y \) to \( s_{na}^y \) which is a state of permanent violation. In other words, this transition is not allowed at this point in time and the only way to explain the event is by labeling it as a move on log, which does not cause the state of the Workflow net to change, neither does it change the state of any of the automata.

The next event to explain would be the event Receive cancellation \( (t_{11}/n) \). The execution of \( t_{11}/n \) is allowed as it would not violate any of the constraints. The algorithm will therefore explore this option. However, after executing \( t_{11}/n \), it is no longer possible to execute \( t_5/e \) without reaching a state of permanent violation because of \( \kappa_5 \). To terminate the run, this forces the execution of \( t_3/i \), which, in turn, requires \( t_{12}/w \) to be fired through \( \kappa_6 \) and \( \kappa_7 \). This would lead to an alignment
Fig. 9. The worst-case expansion of the search space of the basic A* algorithm for the trace $\langle o, n, n \rangle$ and the Workflow net of Fig. 1. The shortest path corresponding to $\gamma^{opt} = ((t_1, o), (t_{11}, n), (t_{11}, n), (t_{10}, \triangleright))$ is highlighted with a superimposed dashed line.

(a) \text{NotSuccession}(n, m)

(b) \text{NotSuccession}(n, i)

(c) \text{NotSuccession}(n, c)

(d) \text{Precedence}(n, w)

(e) \text{Precedence}(n, m)

(f) \text{AltPrecedence}(e, w)

(g) \text{AtMostOne}(w)

(h) \text{AtMostOne}(m)

(i) \text{AltPrecedence}(s, w)

(j) \text{Succession}(m, r)

(k) \text{Succession}(w, r)

Fig. 10. All the automata for the constraints in the model of Fig. 1. The light gray states are the states the automata are in when transitions $t_1/o, t_2/m, t_4/i, t_6/s$ and $t_5/e$ have been executed.
\[ \gamma = \langle t_1, o, (t_2, p), (t_4, i), (t_6, s), (t_5, \epsilon), (t_7, \gg), (m, m), (t_11, n), (t_12, \gg), (t_13, \gg), (t_9, \gg) \rangle \] with four deviations, namely a move on log on \[ t_{12}/m \] and three moves on model on \[ t_{12}/m \] and \[ t_9/s \]. We already know that there is a better alignment with only three deviations, namely \[ \gamma' = \langle t_1, o, (t_2, p), (t_4, i), (t_6, s), (t_5, \epsilon), (t_7, \gg), (m, m), (t_11, n), (t_12, \gg), (t_13, \gg) \rangle \] with two moves on model or log, this alignment would become the only optimal one. To overcome this issue, we modify the algorithm as follows.

4.5. Allowing for constraint violation at a cost

In this subsection, we show how we modify our algorithm to allow constraints to become permanently violated at a cost. To this end, we introduce a cost function that, for each constraint in the model, associates a cost to its permanent violation. When executing a transition in the model, we then no longer need to check if any of the automata becomes permanently violated, but we need to add costs for when they do. However, this alone is not sufficient as we also need to update the termination condition.

In Alg. 2, the main loop terminates if (1) the model reaches the final marking, (2) the trace is fully explained and (3) all constraints are satisfied. The latter condition needs to change to "all constraints are satisfied or permanently violated". We cannot leave an automaton in a temporarily violated state and simply add the costs of violating that constraint in the final step as there may be better alignments which we have not yet explored.

Algorithm 3 shows the final algorithm that allows for the violation of constraints while making alignments. An additional parameter \( C \) includes the cost for permanently violating a constraint at the end of the alignment. The part after Line 10 updates any automaton that is in a state of temporary violation to a permanent violation, updates the cost, and re-queues the state. Again, it is trivial to see that this algorithm produces optimal alignments. The underestimate function is correct for the model without constraints, and when we allow constraints to be violated, the language of the model does not change when adding constraints. Furthermore, the additional costs for violating constraints does not break the monotonicity of the heuristic required for the \( A^* \) algorithm to be correct [76].

It is important to notice that this algorithm introduces two new move types, namely moves on model and synchronous moves that violate a constraint — as opposed to the existing violations on model and synchronous moves that do not violate constraints. Because the constraints can now be permanently violated, the size of the search space for \( A^* \) also becomes larger. In the worst case (if all constraint automata are defined on disjoint sets of transitions), the size of the search space corresponds to the product of the sizes of the constraint automata and of the state space of the Workflow net. However, in practice, this rarely leads to problems, as we will see in Section 5.

Let us again consider the example of before \( \langle o, p, i, s, e, m, n \rangle \). Again, assume we have reached a state in the search where we have explained the fitting prefix of the sequence (i.e., we executed transitions \( t_1/o, t_2/m, t_4/l, t_6/s \) and \( t_5/e \) in the model). The next event to explain is Return money. Recall that the marking is \( \langle p_k, p_i, p_j \rangle \) and the automata are in their initial state, except for the automata of \( k_3 \) and \( k_9 \) in states \( s_9^3 \) and \( s_9^9 \), respectively. Transition \( t_1/m \) is enabled according to the workflow model. If we executed it, automaton \( k_4 \) would transition from state \( s_9^3 \) to \( s_9^4 \), which is a permanently violated state. For the sake of argument, let us assume all violation cause an additional cost of 1 to be added to the total, i.e., violating a constraint incurs a partial cost of 1, just like a move on model or a move on log. In Line 25, the cost of permanently violating constraint \( k_4 \) is added to the cost of that synchronous move. The new state is put in the queue for further investigation.

The next event to explain would be the event Receive cancellation. Its execution is allowed as it would not violate any of the constraints that are not already permanently violated. The algorithm will therefore explore this option. After executing \( t_{11}/n \), it is no longer possible to execute \( t_9/e \) without reaching a state of permanent violation, due to \( k_5 \). To terminate the run, this forces the execution of \( b_j/r \), which, in turn, through \( k_7 \), requires \( t_{12}/w \) to happen first. This would lead to an alignment \( \gamma'' = \langle (t_1, o), (t_2, p), (t_4, i), (t_6, s), (t_5, \epsilon), (t_7, \gg), (m, m), (t_11, n), (t_12, \gg), (t_13, \gg) \rangle \) with three deviations, namely a synchronous move violating a constraint on \( t_{13}/m \) and two moves on model on \( t_{12}/m \) and \( t_9/r \).

We already know that there is an alignment with three deviations, but this alignment is equally optimal. If the cost of violating a constraint is selected to be lower than the cost of a move on model or log, this alignment would become the only optimal alignment.

Both algorithms presented in this section rely on the original estimation function for alignments, which ignores the declarative constraints. While this is provably correct from an algorithmic
point of view, it implies that the part of the search space that needs to be visited by the algorithm increases. In the worst case, the model contains no places and only declarative constraints. Such a model would yield the largest possible search space and cause the performance of the \( A^* \) algorithm to degrade to the level of Dijkstra’s shortest path algorithm [77].

### 4.6. Calculating fitness

Fitness is a process quality measure that assesses the extent to which a process model explains the executions recorded in the event log [59,73]. In particular, it considers the ability of the model to retrace the traces. The measure is expressed as the fraction of the log behavior that is admitted by the model. As our approach is built upon alignments for conformance checking, we apply the usual calculation for fitness proposed in [46]. Notice, indeed, that the adaptations in the concepts of runs for mixed-paradigm models, as well as in the cost assignments to cater for possible violations of the declarative part of the mixed-paradigm, are compatible with the standard definition of alignment. We calculate fitness, in particular, as follows.

Given a trace \( \sigma = (\sigma_1, \ldots, \sigma_n) \) of length \( n \in \mathbb{N} \) and a mixed-paradigm model \( \mathcal{MP} \), let \( \gamma_{opt}^{\sigma,\mathcal{MP}} \) be the optimal alignment of \( \sigma \) and \( \mathcal{MP} \) as per Definition 11. Let \( \gamma_{worst}^{\sigma,\mathcal{MP}} \) be the worst-case alignment, calculated as the cost of the optimal alignment of the empty trace \( \epsilon \) and \( \mathcal{MP} \), plus the cost of a sequence of moves on log for every \( \sigma_i \in \sigma \) (\( 1 \leq i \leq n \)). The trace fitness \( \text{fitness}(\sigma, \mathcal{MP}) \) is calculated as follows:

\[
\text{fitness}(\sigma, \mathcal{MP}) = 1 - \frac{\text{cost}(\gamma_{opt}^{\sigma,\mathcal{MP}})}{\text{cost}(\gamma_{worst}^{\sigma,\mathcal{MP}})}.
\]

For example, consider again the example trace seen above \((t_0, p, i, s, e, m, n)\) and the process model of Fig. 6 and a unitary cost for moves on model, moves on log, and constraint violations. As shown in Sections 4.4 and 4.5, the optimal alignment is associated to a cost of 3. The cost of the worst-case alignment is the sum of 3 (the full alignment with an empty trace requires three moves on model, corresponding to the firing sequence \((t_1, t_2, t_0)\)) and 7 (corresponding to a move on log for every event in the trace). The trace fitness is thus \( 1 - \frac{3}{10} = 0.7 \).

In the next section, we show how the technique performs on a number of real-life event logs on which we discover mixed-paradigm models using an available mixed-paradigm discovery technique.

### 5. Evaluation

In this section, we evaluate our mixed-paradigm conformance checking approach on three real-life event logs. We implemented and integrated our algorithm in the ProM environment. The tool can be found in the ProM nightly build under the name MixedParadigm.

In the following, we introduce the event logs and discovery algorithm used. Next, we provide a qualitative analysis of one of the logs to show the capabilities of the mixed-paradigm conformance checking approach. Finally, we mine different mixed-paradigm models mined from each log and evaluate them by making use of our alignment-based algorithm.

#### 5.1. Event logs

The process mining literature has provided and discussed a wide range of exemplary, real-life event logs. In this paper, we present the results of our experiments on three such logs to test the conformance checking algorithm for its performance. Two of those are part of the Business Process Intelligence Challenge (BPI) collection. In addition, we consider an extra event log (Sepsis). We chose those logs as the parameters of the discovery algorithm used in subsequent experiments yield a variety of different models, thereby serving as a good illustration of the applicability of the replaying algorithm. The selection offers different blends of the main characteristics of event logs, i.e., size, number of distinct activities, and trace length. Table 2 provides an overview of the main characteristics of the logs used in this study:

| Event log | No. activities | No. traces | Avg. trace length |
|-----------|----------------|------------|-------------------|
| BPI 2012  | 24             | 13,087     | 20.04             |
| BPI 2013 - incidents | 13 | 7,554 | 8.68 |
| Sepsis    | 16             | 1,050      | 14.49             |

#### BPI 2012: The BPI 2012 log\(^3\) (henceforth, BPI12 for short) contains the application process for personal loans and credit lines at a Dutch financial institution. It consists of three subprocesses, each pertaining to a different part of the organization, i.e., the [A]pplications, the [O]ffers, and the [W]orkflow system. Customers submit an application for a loan through an online system. These applications are then fed into a workflow system, where call-agents can pick up work items to either assess the application or to contact the customer with offers. Eventually, every application should either be declined, canceled or approved and activated.

#### BPI 2013: The BPI 2013 log contains traces from an incident management system at Volvo IT Belgium. It contains a smaller range of activities, with more repetitive structures compared with the BPI 2012 log. Three different logs were made available. For the evaluation, we use the one that focuses on the separately raised incidents.\(^4\) We will indicate this log as BPI13 for short in the remainder of the paper.

#### Sepsis: The Sepsis log\(^5\) pertains to the treatment of sepsis in a hospital setting and contains various activities stemming from an ERP system supporting the medical as well as administrative tasks.

We performed similar experiments on other event logs as well. However, the variety of models generated was lower. The interested reader can find the results at [https://github.com/JohannesDeSmedt/mixed-paradigm-conformance-checking](https://github.com/JohannesDeSmedt/mixed-paradigm-conformance-checking).

---

\(^3\) [https://doi.org/10.4121/uuid:915d2bfb-7e84-49ad-a286-dc35f063a460](https://doi.org/10.4121/uuid:915d2bfb-7e84-49ad-a286-dc35f063a460).

\(^4\) [https://doi.org/10.4121/uuid:500573e6-accc-4b0c-9576-a55468b10cee](https://doi.org/10.4121/uuid:500573e6-accc-4b0c-9576-a55468b10cee).

\(^5\) [https://doi.org/10.4121/uuid:915d2bfb-7e84-49ad-a286-dc35f063a460](https://doi.org/10.4121/uuid:915d2bfb-7e84-49ad-a286-dc35f063a460).
Petri-net part of the model, we use the block-structured approach of Inductive Miner [37]. We use MINERful [40] to connect activities in the declarative model, and to link them with the procedural model. MINERful employs the concepts of support and confidence to mine constraints from an event log. Support is the proportion of traces that comply with the constraint. Confidence is the proportion of satisfying traces in which the activation of the constraint occurred. The higher the support, the lower the number of constraints, but the higher trace fitness is to be expected, as more traces will allow for the behavior of the constraints. Next to a varying level of entropy, a varying level of support and confidence will impact the amount and overall applicability of constraints in the mixed-paradigm model. Not every distinct combination of these parameters yields a different result. In particular, we report on the results of applying FusionMINERful setting support and confidence levels between 0.5 and 1.0, and entropy level between 0.0 and 1.0. For each event log, we display an overview of the different levels of entropy, support, and confidence which churned out a unique model with FusionMINERful, as well as how many synchronous moves of 1, and a cost of violating a constraint of 1. Replaying is performed by both allowing and disallowing constraint violations: we report the results on trace fitness separately, according to the adopted strategy.

5.2. Qualitative evaluation

As discussed, mixed-paradigm models excel at capturing different layers of flexibility in event logs. The procedural part of the model represents the more fixed workflows, while the activities that are subject to declarative constraints allow for more intricate behavior. An interesting aspect of adding constraints to procedural models, indeed, is the introduction of an immediate insight into this more intricate behavior.

The incorporation of the constraint checking in the replay provides a visual representation of which constraints are violated during the execution. This is illustrated in Fig. 11. Violated constraints are drawn in red. Additionally, transitions are decorated on the bottom side with lines that denote the occurrence of moves on model (purple). In order to get a reasonable visual representation in terms of model size (as measured in the number of model constructs), the offer-making subprocess of the BPI12 log is mined for a mixed-paradigm model with a conservative entropy level (40%) and high support and confidence (100%) to illustrate the constraint-checking capabilities of the algorithm. It can be seen that the NotChainSuccession constraints are violated during execution. This comes as no surprise, given that there is little possibility to circumvent these constraints during the replay without introducing many moves on model. Besides, all other constraints end in an accepting state. Nevertheless, fitting the O_SENT_BACK activity into the alignments requires many moves on log, which lower the overall trace fitness score to 80.21%.

Besides the visual representation of the model, our software shows the detailed information on the fitness scores on the right hand side, as it can be seen in Fig. 11. A detailed overview of the alignment steps is also available, highlighting the moves that introduce constraint violations, moves on model, and moves on log.
5.3. Quantitative evaluation

The main goals of trace-based conformance checking is to compute a fitness score for mixed-paradigm models over which event logs are replayed, and to gain further information regarding the events that could not be executed (or constraints that could not be satisfied) in accordance with the underlying model. In this section, we illustrate how we retrieve this information from numerous models mined and replayed on the event logs.

Fig. 12 depicts an overview of the results for computation time and fitness. A baseline result for a completely procedural model mined with Inductive Miner is provided in Table 3. Overall, the mixed-paradigm conformance checking algorithm is capable of obtaining very high levels of fitness with higher levels reported when constraint violations are allowed — especially for the BPI13 log. This might indicate that there are particular

| Event log | Trace fitness | Comp. time | No. transitions | No. places |
|-----------|---------------|------------|-----------------|------------|
| BPI12     | 0.948         | 5.916      | 40              | 19         |
| BPI13     | 0.819         | 1.629      | 25              | 14         |
| Sepsis    | 1.000         | 2.682      | 23              | 9          |

Fig. 13. Overview of the results for the BPI12 log.
constraints which are not compatible with the procedural model at lower levels of support. Relaxing the need for an execution where constraints can be violated can quickly increase fitness. It also has a positive impact on run-time for BPI13, with the BPI12 and Sepsis logs reporting similar timings with a larger spread for the latter. As shown in Figs. 13 and 15 to 18, the highest number of constraints are mined for the Sepsis log. Running our implemented algorithm on this log also results in the highest computation time despite the lower number of traces. This is in line with the intuition established in Section 4.5, as constraints extending the search space entail worse performance.

A more detailed overview of fitness, execution time, and the number of modeling constructs in relation to the entropy and support/confidence parameters can be found in Figs. 13 and 15 to 18. Note that the number of transitions and places are reported for the procedural part of the model, i.e., at an entropy of 100%, all activities are present in the declarative part of the model.

The high levels of fitness of BPI12 can be found throughout the full entropy range. Lower levels are reported for an entropy level of 20%, suggesting that a particular combination of constraints is present that seems to be incompatible with the Workflow net’s state space or rather with each other, as allowing for violations does not restore fitness. Besides, there is a steady increase in the number of constraints and a similar decrease in the number of transitions over the entropy range. This means that most activities are quickly eligible to be added to the declarative part of the model given their behavior. There is little influence of the support and confidence parameters and none of the constraints are violated (and are thus not reported separately). Compared to the fully procedural model, however, the mixed-paradigm model is capable of achieving higher fitness (100% compared to 95%) through adding parts of the model to the declarative part to avoid replay on any presumably overfitting procedural parts. Nevertheless, reducing the noise threshold of Inductive Miner could lead to perfect fitness, at the price of a less precise model. At an entropy level of 10%, only a few constraints are added to the procedural model and 100% fitness is achieved. Given the constraints cut into the state space of the procedural model, the mixed-paradigm model by definition gives a more precise result with a few more constructs. At higher entropy levels (e.g., 60% to 70%) we obtain the best trade-off between higher fitness and fewer model constructs of either paradigm. The number of constraints does not influence the computation time directly, while there is a single peak for the search at an entropy level of 60% in case constraints cannot be violated. This indicates that, indeed, some constraint combinations can result in a higher execution time if they result in similar costs in the procedural state space leading to longer queues with equal cost, or are defined over a particularly disjoint set of transitions as explained in Section 4.4. This makes convergence towards a unique optimal solution harder. In general, it is not straightforwardly possible

![Fig. 14. Output of the mixed-paradigm conformance checker with two parameter settings (violations allowed) on the BPI13 log.](image)
to pinpoint which constraints or combinations are causing difficulties in the state space search as this relates to both the type of constraint(s), the combination of constraint(s) (types), as well as to which transitions they constrain and their connections in the Workflow net. Besides, the impact of DECLARE constraints on the size of the state space is non-monotonic [63]. Given that the size of transitions remains relatively stable around this level of entropy, it appears that a particularly disruptive constraint (type) is introduced, which is often part of the negative constraints set (e.g., NotSuccession, ExclusiveChoice, and the like).

The picture is different for the BPI13 log. Fitness levels vary slightly over the entropy range but mostly differ according to the level of support and confidence. Higher levels of fitness are achieved by either allowing violations, or setting support at a threshold of 100% to extract fewer yet non-conflicting constraints during the discovery phase [63]. Both strategies result in models that improve over the baseline fitness of the purely procedural model. No constraints were mined for 100% confidence. The lower the confidence, the higher the number of constraints though the number of constraints violated on average per trace does not increase (Fig. 16(d)). The models with the higher confidence levels (80% to 90%), however, result in low fitness even when violating constraints is allowed (except for 100% support). Hence, it shows that only a small portion of constraints can cause problems for achieving high fitness, whereas allowing them to be violated restores fitness to some extent. Again, we conclude that there are both incompatible constraints, due to the support threshold set at less than 100%, and conflicts in the state space with the procedural model.

Fig. 14 shows two models with different fitness levels due to the confidence of constraints mined, leading to different parts of the procedural model being cut away and replaced by constraints. In the case of the higher confidence level, more of the procedural model is cut away and substituted by incompatible constraints, which are violated, and more invisible transitions. This shows how the mixed-paradigm conformance checker can give insights into what blends of constructs can work well for a particular event log.

The execution time in this case is higher for exploration when violations are allowed for confidence levels at 80 and 90%, although this is not related to the number of constraints. It coincides with the lower fitness levels, meaning a solution is hard to find and may not exist due to a high level of conflicting behavior between the declarative and procedural model, and/or the constraints themselves.

Finally, the Sepsis event log shows the most interesting spread of results over the entire entropy spectrum. Here, it can be seen that a lower level of confidence increases the number of
constraints generated, which has a negative impact on fitness. Indeed, the average number of violated constraints per trace generated with support below 100% rises steadily when more constraints are generated. At 100% support, naturally there are close to no violated constraints which is to be expected as these constraints hold for all traces, but given the <100% fitness there still seem to be constraints incompatible with the procedural model. The addition of constraints mined with high confidence and support do not affect the fitness, but overall the procedural baseline of 100% is only obtained in a fully declarative model or when violations are allowed. Hence, the mixed nature of the model cannot offer any particular benefit over a model of either paradigm in this case. The execution time in this case is again worse for when violations are allowed, meaning there are multiple parts of the procedural state space that result in a similar cost in combination with a large body of constraints. This makes finding an optimal solution with the least amount of violated constraints hard as is illustrated by the low fitness values. It is especially noticeable for lower levels of support, which return more constraints but not necessarily result in more constraints being violated per trace on average. The execution time for completely declarative models is very low, despite potentially resulting in the largest search space (Section 4.5) which makes the approach also competitive against the previous work on alignments for declarative models presented in [18].

Overall, the mixed-paradigm conformance checker is capable of offering some interesting results to analyze the trade-offs which exist when blending constructs of either model paradigm. Most notably, it helps to find the right level of mixture, besides offering a fully procedural or declarative result if needed. For every event log we have analyzed, a different interplay between the model constructs took place: either the declarative constraints replaced procedural fragments to increase fitness, or the declarative constructs were rather added (with low confidence/support) to achieve lower fitness due to contradicting behavior – let it be between the declarative constraints and the procedural model or among constraints themselves. Finally, in the case of BPI12 and BPI13, the mixed-paradigm model offers a lower number of constructs (i.e., constraints, transitions and places) at certain parameter settings compared to a fully procedural model (consisting of transitions and places), while offering more detailed behavior (higher precision) compared to the 100% fitness which is by definition achieved at 100% support by declarative models consisting of only a few constraints. The execution time is influenced by particular combinations of constraints which coincides with low levels of fitness, but is overall low and relatively stable over different combinations of procedural and declarative constructs in a mixed-paradigm model.
5.4. Implications and limitations

The empirical evaluation has shown that mixed-paradigm models can improve replay performance over the procedural and declarative paradigms, as it can deal with fewer constructs once the right mixture of either paradigm is found. It is thus possible to obtain alignments for mixed-paradigm models with reasonable computational expense. The addition of constraints does not drastically increase execution time and often poses no extra requirements on the state space analysis. Even fully declarative models can be analyzed in comparable time to procedural models with only a few constraints. Exceptions to this occur mostly when particular combinations of constraints, not necessarily confined to a fixed range of the entropy spectrum, are present although high levels of confidence and support tend to limit this issue. Therefore, it is best to first check the compatibility of the state spaces of either paradigm [13].

The conformance checking approach, however, is also limited in certain aspects. The A*-algorithm, while heavily optimized for conformance checking, is still not efficient in every case. While not apparent from the experimental evaluation, the computational effort can increase in case the activities over which constraints are defined have less overlap with the Workflow net when violations are allowed, or in case a high number of constraints reflect contradictory behavior leading to long state space explorations. This will especially be the case for process models with a high number of activities. However, for a mixed-paradigm model with intertwined state spaces to be useful, it is expected to have a reasonable amount of overlapping activities and state spaces that are compatible to a certain extent. Besides, the current efforts focus on computing fitness and tracking the number of constructs of either paradigm to capture overall readability/simplicity of the models. However, they are only two aspects of conformance checking. Most notably, it would be interesting to cover precision and generalization to quantify to what extent the mixed-paradigm models, when being more or less fitting, are still performing well in not being overly precise or too general in terms of the allowed behavior. This is closely related to the number of constraints and often interacts with fitness. This can be covered in future iterations by incorporating alignment-based approaches such as [78].

Finally, the evaluation has only focused on one particular combination of procedural and declarative languages. It would be interesting to further uncover how different languages mixed together can result in the best mixed-paradigm representation, e.g., whether DECLARE with Workflow nets are better suitable compared to DECLARE with BPMN, or to DCR Graphs with the
Fig. 18. Overview of the results for the Sepsis log (ii).

6. Conclusion

In this paper, we have presented the first approach enabling conformance checking for mixed-paradigm models. More specifically, our approach handles the intertwined state space of mixed-paradigm models by the help of alignment-based concepts for Petri nets extended with automata for each DECLARE constraint. We investigated both options of prohibiting state space exploration that induces constraint violation and of penalizing such violations with costs. In an experimental evaluation with three real-life event logs, we illustrate how alignments are efficiently calculated over a wide range of mixed-paradigm models with various levels of procedural and declarative constructs. Our contribution has strong implications for research and practice. Our approach fills the research gap of mixed-paradigm conformance checking and represents an important step for the further development of mixed-paradigm process mining. On the practical side, our approach and its implementation in ProM helps process analysts to find the most appropriate mix of either paradigm depending on the characteristics of the event log.

For future work, we intend to investigate other conformance checking metrics besides fitness such as precision and generalization. Aside from the trade-off that can now be made between the number of constructs generated and trace fitness, it will be interesting to see how much more the reduction in state space by the declarative constructs can improve the precision of procedural models and whether this impacts generalization drastically. Besides, we will apply the insights from the alignment results to improve mixed-paradigm process mining algorithms.

Declaration of competing interest

The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

Acknowledgment

The work of C. Di Ciccio was partly supported by the Italian Ministry of Education, University and Research (MIUR) under grant “Dipartimenti di eccellenza 2018-2022” of the Department of Computer Science at Sapienza University of Rome.

References

[1] D.J. Gilmore, T.R.G. Green, Comprehension and recall of miniature programs, Int. J. Man-Mach. Stud. 21 (1) (1984) 31–48, http://dx.doi.org/10.1016/0020-7373(84)80037-1.
