

Compliance Verification of Business Processes with (Constraint) Temporal Answer Set Programming

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based on

L. Giordano, A. Martelli, M. Spiotta, D. Theseider Dupré, **Business Process Verification with Constraint Temporal Answer Set Programming**, TPLP 13 (4-5), 2013. Presented at ICLP 2013.

L. Giordano, A. Martelli, M. Spiotta, D. Theseider Dupré, **Business Processes Verification with Temporal Answer Set Programming**. KiBP, 1st Int. Workshop on Knowledge-intensive Business Processes, Rome, 2012

both build on:

L. Giordano, A. Martelli, D. Theseider Dupré. **Reasoning about actions with Temporal Answer Sets**. TPLP 13 (2), 2013

Goal

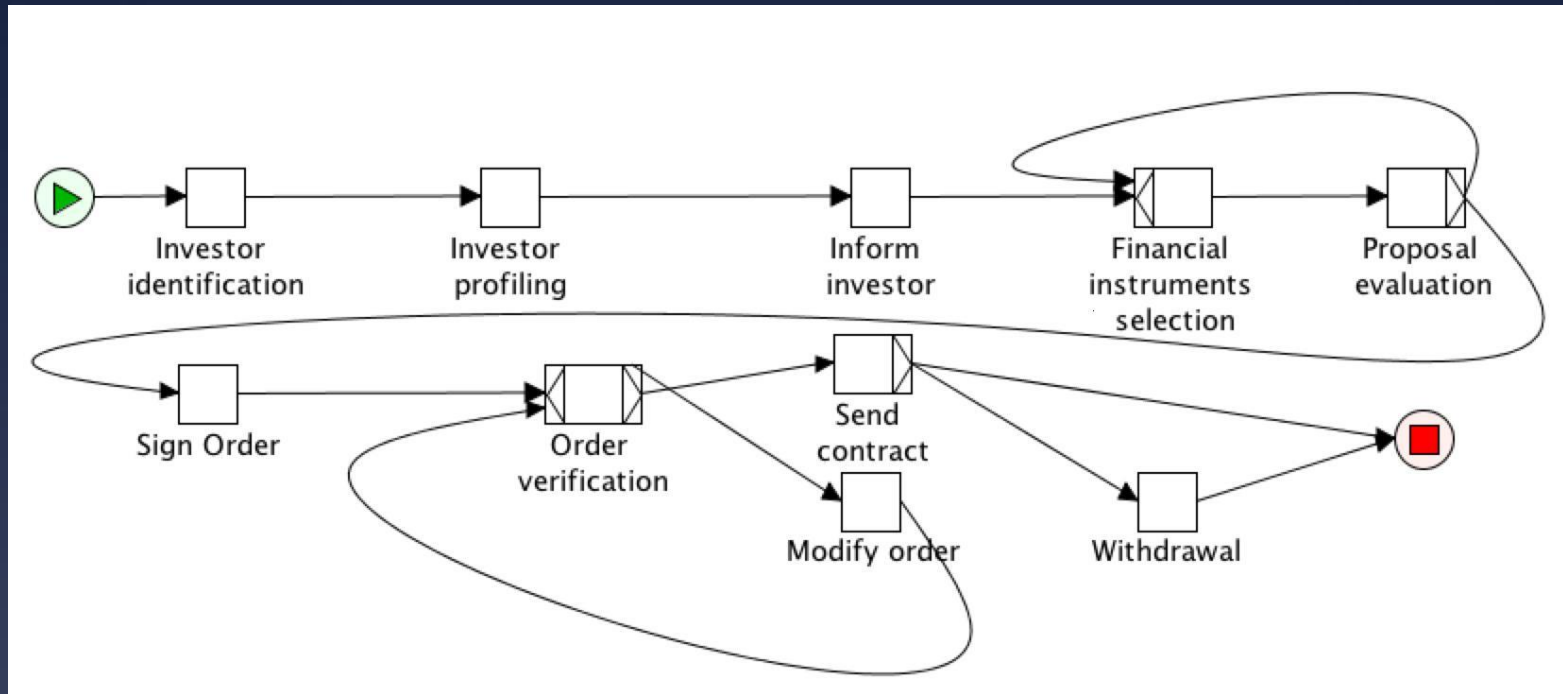
Cross-fertilize Business Process modeling with **Reasoning about actions and change in AI** and **(Constraint) Answer Set Programming**

(Constraint) Temporal Answer Set Programming combines (Constraint) ASP with Temporal Logic (DLTL)

useful for:

- Declarative or procedural process model
- Modeling background knowledge: direct effects of activities and side effects
- Constraint solving on numeric process data
- Compliance verification via Bounded Model Checking based on [Giordano, Martelli & T.D. TPLP 13]

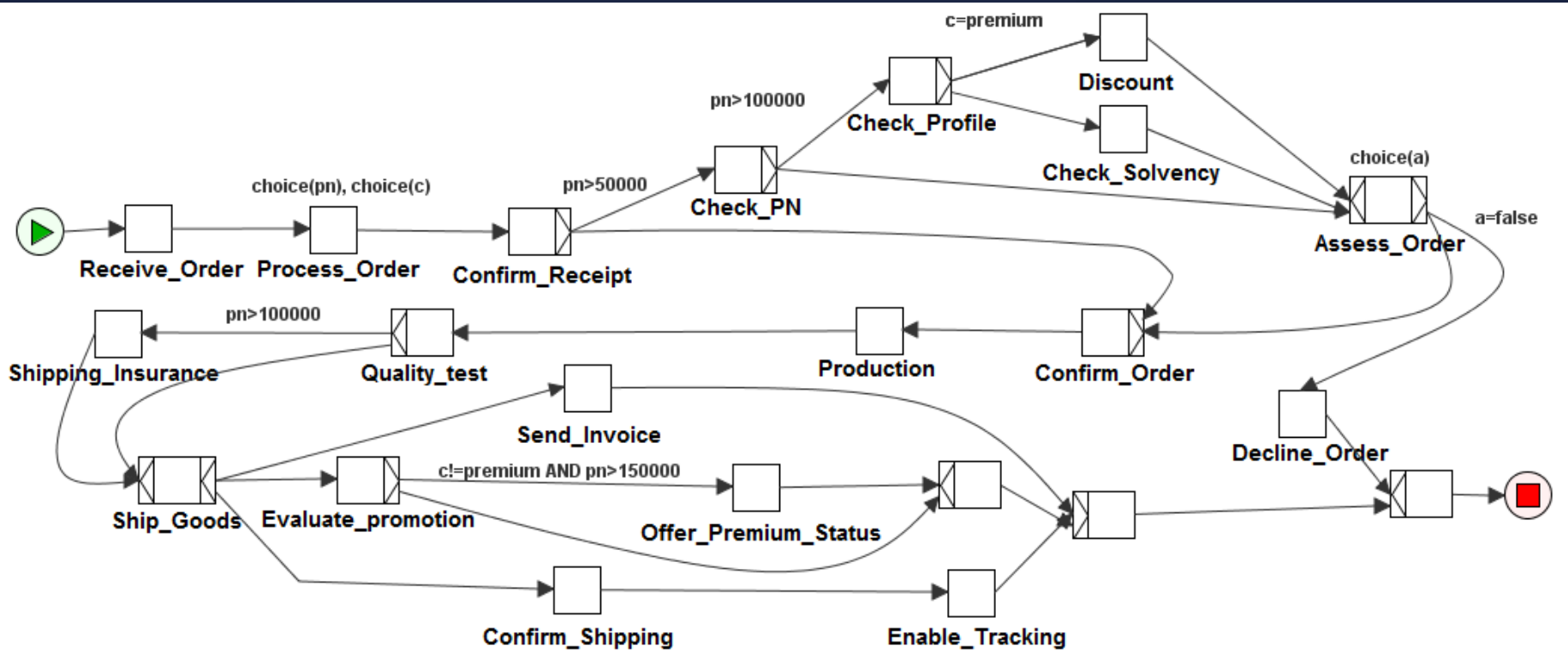
Example



Norm 1: “The firm shall provide the investor adequate information on its policies before any contract is signed”

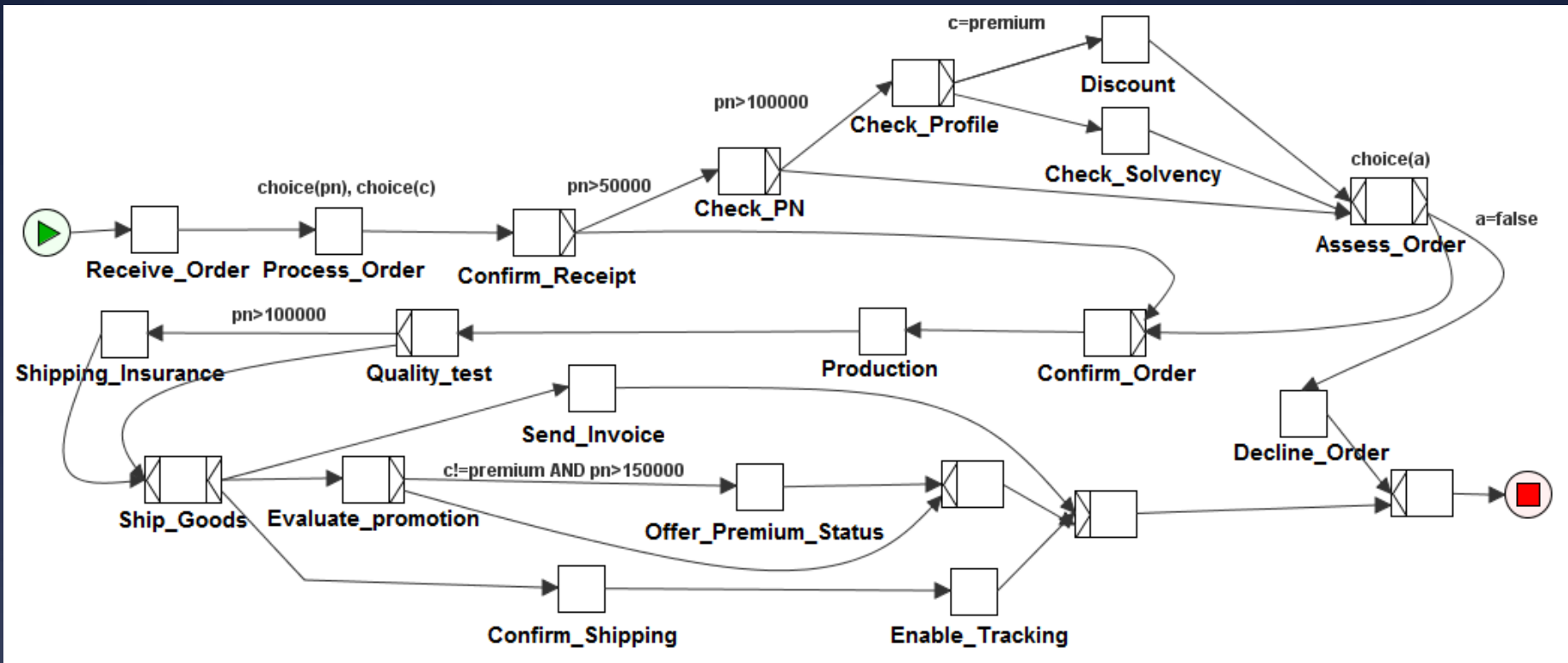
Norm 2: “If an investor signs a contract, the firm shall provide him a copy of the contract”

Example



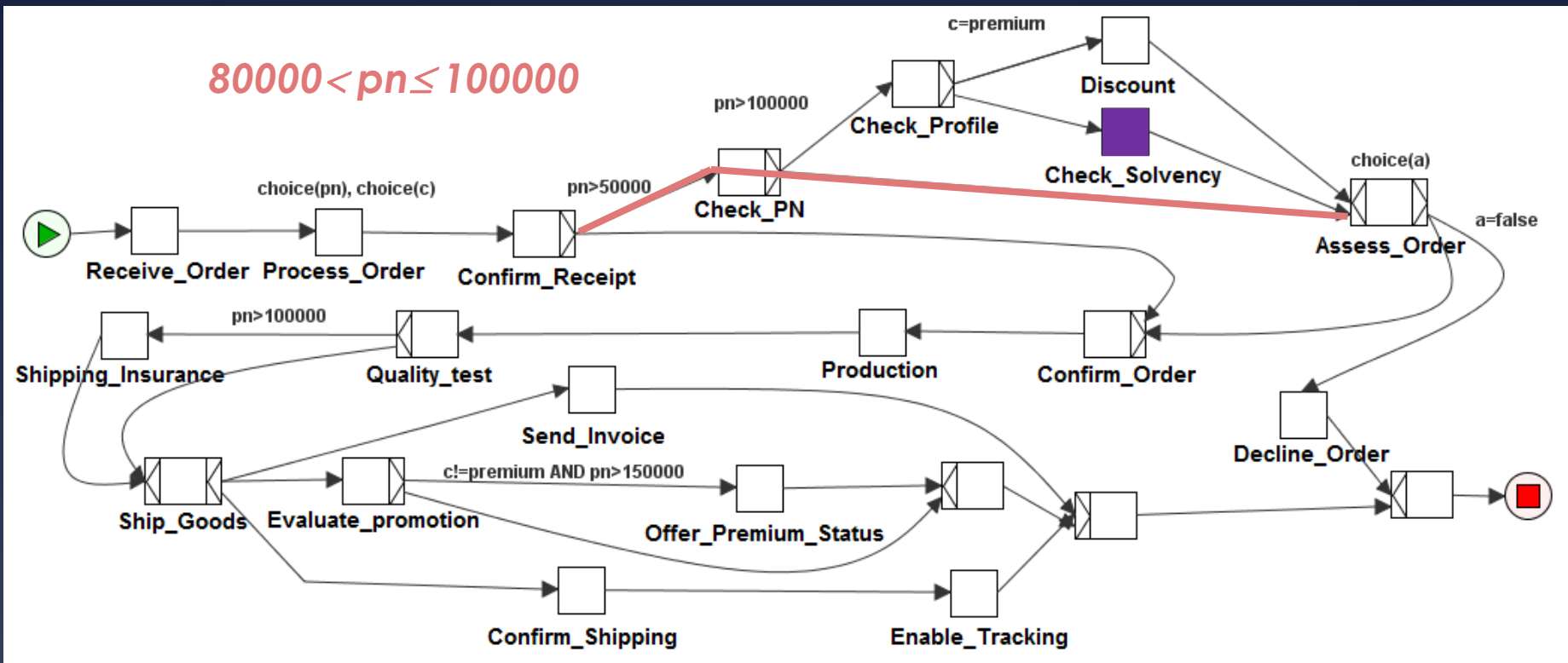
Order-delivery process adapted from [Knuplesch 2010],
branching depends on variables, esp. *piece number*

Rules



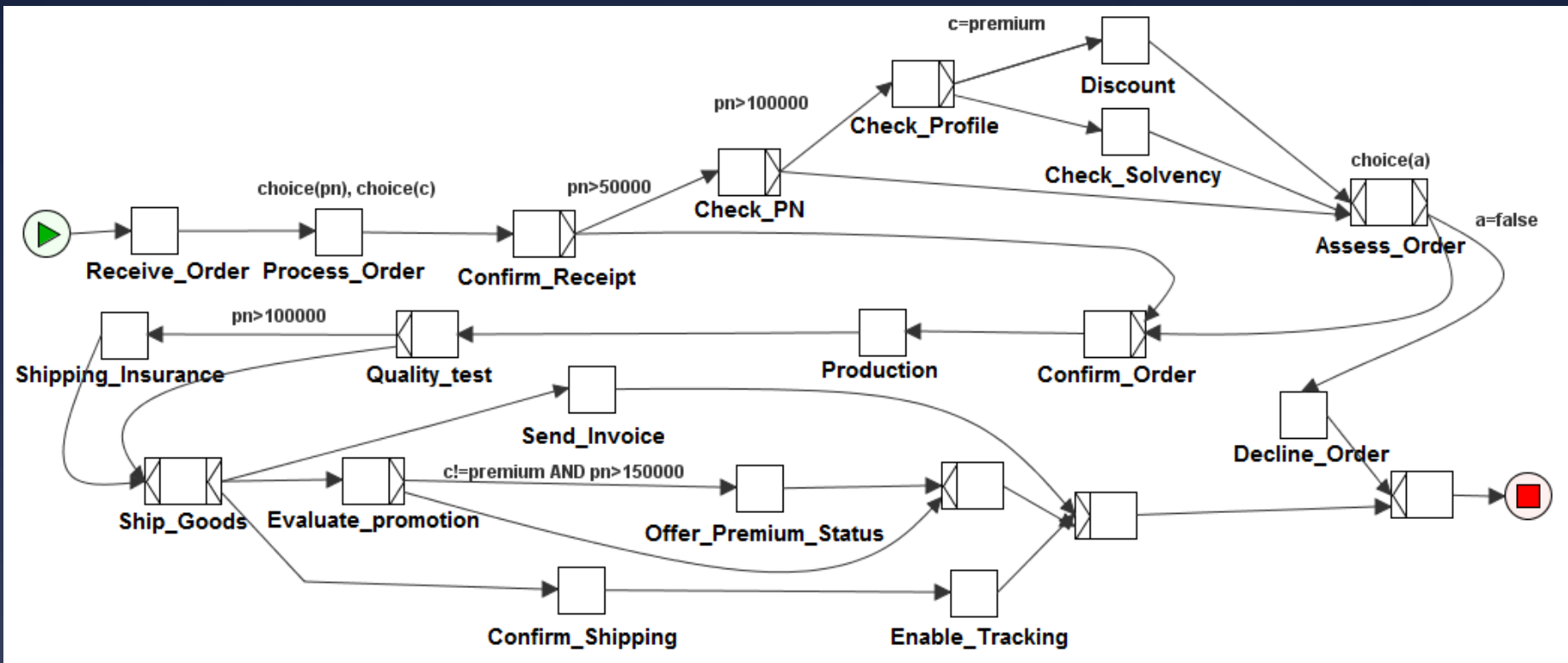
- After confirming an order, goods have to be shipped eventually
- An order shall either be confirmed or declined

Rules



- Orders with $pn > 50000$ shall be approved before they are confirmed
- For orders of a non-premium customer with $pn > 80000$ a solvency check is necessary before assessing the order

Rules



For orders of a non-premium customer with $pn > 80000$ a solvency check is necessary before assessing the order

$$\square(pn > 80000 \wedge c \neq \text{premium} \wedge \langle \text{Assess_Order} \rangle T \rightarrow \text{solvency_check_done})$$

Representation languages

Then, we use:

- An **action language**, used to describe a domain, where effects of atomic actions and their executability conditions may involve constraints
- A **temporal logic (with constraints)**, used to express (at least) formulae to be verified

Constraints (e.g. $pn > 80000$, or $x + y > k$) will be treated as atoms at the temporal logic level and the answer set level (as in [Gebser et al 09])

DLTL (with constraints)

DLTL [Henriksen & Thiagarajan 99] extends LTL:
temporal operators can be indexed with regular
expressions (programs) π

Temporal formulae include:

- $\langle \pi \rangle \alpha$ there is an execution of π after which α holds
- $[\pi] \alpha$ α holds after all possible executions of π
- $[a] \alpha$ α holds after a

and the usual temporal logic modalities:

$\diamond \alpha$ (eventually α), $\square \alpha$ (always α), $\circ \alpha$ (next α)

Their semantics is defined from the one of $\alpha \mathbf{U}^\pi \beta$
which means: there is an execution of π after which β
holds, and α holds in all previous states

DLTL (with constraints)

DLTL formulae with constraints are then:

$$\perp \mid \top \mid p \mid g \mid \neg\alpha \mid \alpha \vee \beta \mid \alpha \mathbf{U}^\pi \beta$$

where the p s are atomic propositions and the g s are constraints in a constraint language (which we assume to have a *finite domain*)

As in [Gebser et al 09], a function γ maps constraint *atoms* (syntax) to constraints (relations on variables), then providing an interpretation for constraints (e.g. the usual interpretation for arithmetic ops and rels), $A \models_\gamma g$ means: $\gamma(g)$ is true for the assignment A of values to variables, e.g. $A \models_\gamma x+y>20$ if $A(x)=15$ and $A(y)=10$

DLTL with constraints

A model is $M = (\sigma, V, v)$ where σ is an infinite sequence of actions, V and v provide, for each prefix τ of σ (the state reached after τ) an interpretation of atomic propositions, and an assignment for constraint variables. Then:

$M, \tau \models p$ iff $p \in V(\tau)$

$M, \tau \models g$ iff $v(\tau) \models_{\gamma} g$

$M, \tau \models \alpha \mathbf{U}^{\pi} \beta$ iff in σ , after τ , there is an execution τ' of π such that $M, \tau\tau' \models \beta$ and for all intermediate states $\tau\tau''$,

$M, \tau\tau'' \models \alpha$

Temporal action language

implicit

$$\rightarrow \square (l_0 \leftarrow l_1, \dots, l_m, \textit{not} l_{m+1}, \dots, \textit{not} l_n)$$

l_0 is a fluent literal or temporal fluent literal ($[a]l$ or $\circ l$)

The l_i can be:

fluent literals,

constraint literals,

temporal (constraint or fluent) literals,

dynamic constraint literals, i.e. constraint literals also involving variables x° , i.e. 'x in the next state'

with some restriction ensuring that successor states only depend on current state

Action laws, causal laws, persistence can be expressed

Action laws, causal laws

Example of **action law**:

□ (*[inform]informed*)

Persistence:

□ (*[a] I ← I , not [a] ¬I*)

Static causal laws model dependencies within the same state and then also side effects, e.g.:

□ (*¬ confirmed ← deleted*)

where “deleted ” = “order deleted by customer” and
“ confirmed” = “order confirmed for the seller”

Causal laws

Dynamic causal laws:

$$\square (O I \leftarrow t_1, \dots, t_m, \text{not } t_{m+1}, \dots, \text{not } t_n)$$

The t_i 's can be of the forms I_i or $O I_i$

Then we can represent side effects of *changes* of fluents, e.g.:

$$\neg f, O f$$

i.e. f *becomes* true

Ramifications & BPs

[Weber et al. 2010] use clauses (in classical logic) to model dependencies, and the Possible Models Approach [Winslett 1988] to deal with ramifications

the intended states after an action are those:

- where direct effects hold
- where the background axioms are satisfied
- that differ minimally from the state before the action

But one of their examples is:

insurance claim accepted when accepted by
reviewer A and by reviewer B

Ramifications & BPs

If this is modeled as the material implication:

$$\textit{claimAccRevA} \wedge \textit{claimAccRevB} \supset \textit{claimAccepted}$$

and the PMA is used, if A already accepted and B accepts, this either makes *claimAccepted* true or *claimAccRevA* false

The static causal rule

$$\textit{claimAccepted} \leftarrow \textit{claimAccRevA}, \textit{claimAccRevB}$$

can be used to have only *claimAccepted* change as a side effect, while still intending that the implication holds

Ramifications & BPs

The implication may be false if e.g. we allow the acceptance to be overridden later by a supervisor

In this case dynamic laws are appropriate:

$$\begin{aligned} \bigcirc \text{claimAccepted} \leftarrow \bigcirc \text{claimAccRevA}, \\ \rightarrow \text{claimAccRevB}, \bigcirc \text{claimAccRevB} \end{aligned}$$

i.e., if the conjunction of acceptances *becomes* true, we have the side effect, which:

- remains true by default persistence
- may be made false while its original cause remains true

Constraint Temporal Answer Sets

Given a set P of rules, we define a **Constraint Temporal Answer Set** combining Temporal AS in [Giordano, Martelli & T.D.13] and Constraint AS in [Gebser et al 09]

It is a partial temporal interpretation (σ, S) where S is a set of temporal literals of the form $[a_1; \dots; a_k]$ where $a_1; \dots; a_k$ is a prefix of σ

It is defined relative to an assignment v to constraint variables at each prefix of a σ

Then, we define for the various types of literals their being satisfied by (σ, S) at the prefix $a_1; \dots; a_k$ given v

Constraint Temporal Answer Sets

Given an interpretation (σ, S) , for each prefix $a_1; \dots; a_k$ we compute a different constraint reduct, a set of rules

$$[a_1; \dots; a_k] H \leftarrow \text{Body}$$

obtained from rules in P :

- eliminating constraint literals true at $a_1; \dots; a_k$ given v (if all true)
- and extended literals not \perp true at $a_1; \dots; a_k$ (if all true)

reduct = union of reducts for all prefixes

(σ, S) is a constraint temporal answer set wrt v if S is minimal among the R such that (σ, R) satisfies the rules in the reduct

Extensions

Given a domain description (P,Q) where P is a set of rules, and Q is a set of (constraint) DTL formulas, its **extensions** (i.e. models) are constraint temporal answer sets of P whose corresponding temporal model satisfies formulae in Q

Validity of a formula α for a d.d. (P,Q) corresponds to verifying that there is no extension of $(P, \{Q \wedge \neg \alpha\})$

Modeling Business Processes

The control flow of a business process can be modeled in several ways

- a **program** (regular expression) in a DTL constraint:
 $\langle \pi \rangle T$ (only structured, sequential programs)
[Giordano et al. CLIMA 10]
- **declarative** temporal constraints (e.g. ConDec/Declare from van der Aalst et al)
- «classical» graphical **workflow notation** (BPMN, YAWL)

Modeling Business Processes

We used a **translation from basic workflow constructs** of YAWL to the temporal action language, based on the *enabling* of actions and arcs

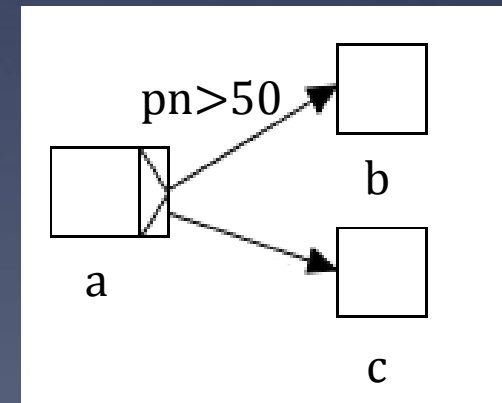
An action precondition is its being enabled

Causal laws define enabling of action based on enabling of incoming arcs (one/all for XOR/AND)

Actions enable and disable arcs

[a] en_arc_a_b \leftarrow pn > 50

[a] en_arc_a_c \leftarrow not pn > 50



Modeling Business Processes

The model provides information on which **actions** have a **variable as output**:

$$[a] x \in [0..1000000]$$

Across other actions, the value of x **persists**, we model this via a fluent change_x which is non persistent and false by default:

$$x^{\circ} = x \leftarrow \circ \neg \text{change}_x$$

$$[a] \text{change}_x$$

$$\neg \text{change}_x \leftarrow \text{not change}_x$$

Other **fluents persist**:

$$[a] f \leftarrow f, \text{not } [a] \neg f$$

Verification

In [Giordano, Martelli & T.D. TPLP13] we defined a **translation** of domain descriptions **to ASP** and an **encoding** in ASP of **Bounded Model Checking** (following [Heljanko & Niemelä 03])

- BMC, given a system and a formula, searches for a model
- Infinite paths are represented as finite paths of length k with a loop back from state k to a previous state
- The search proceeds iteratively, increasing k until a model is found (if one exists)

Translation

Our translation is defined so that **extensions** of domain descriptions correspond to (constraint) **answer sets** of the translation

- $\text{occurs}(\text{Action}, \text{State})$ (State is a number)
- $\text{holds}(\text{Literal}, \text{State})$

e.g. for $[a]f1 \leftarrow f2$:

$\text{holds}(f1, S') \leftarrow \text{state}(S), \text{next}(S, S'), \text{occurs}(a, S), \text{holds}(f2, S)$

- $\text{sat}(\text{Formula}, \text{State})$

defined inductively on the structure of the DLT Formula

Translation

Constraint literals are represented using CSP variables
 $\text{value}(x,s)$ for the value of process variable x at state s

$[a] \text{ en_arc_a_b} \leftarrow \text{pn} > 50$

becomes

$\text{holds}(\text{en_arc_a_b}, S') \leftarrow$
 $\text{state}(S), \text{next}(S, S'), \text{occurs}(a, S), \text{value}(\text{pn}, S) > 50$

$x \circ = x \leftarrow \circ \neg \text{change_x}$

becomes

$\text{value}(x, S') = \text{value}(x, S) \leftarrow$
 $\text{state}(S), \text{next}(S, S'), \text{not holds}(\text{change_x}, S')$

BP Verification

The approach in [Giordano, Martelli & T.D. TPLP 13] is suitable for verifying system with infinite computations (and finite state space)

In BPs only executions that reach the end are considered *sound*

Finite executions can be represented as infinite ones with a final dummy action

In practice, we restrict to finite traces

Completeness of BMC

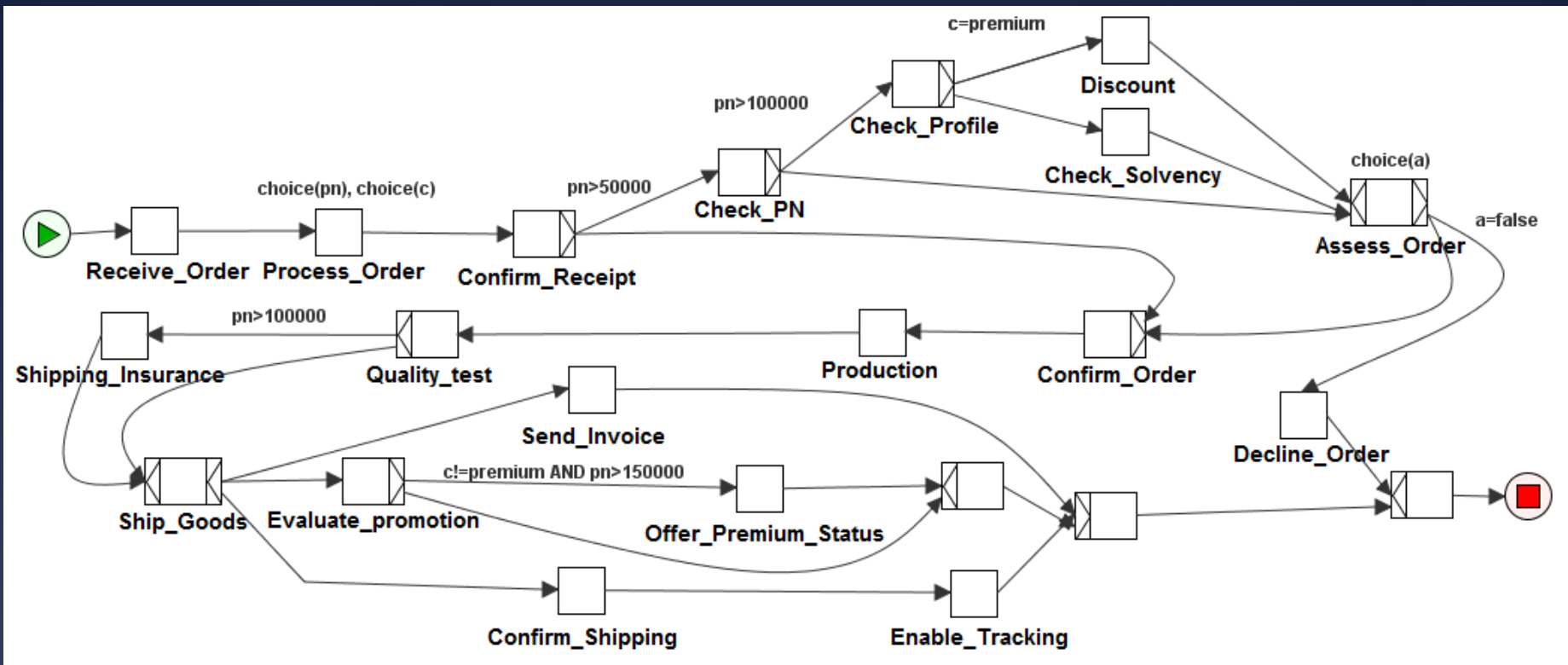
BMC is in general a partial decision procedure

Completeness can be obtained for special classes of formulae, or for general formulae, computing a **completeness threshold** t (using bounds up to t is enough to find a model if one exists) [Biere et al. 03,06, Clarke et al. 04, Giordano et al. KR12]

But computing the threshold may be unfeasible

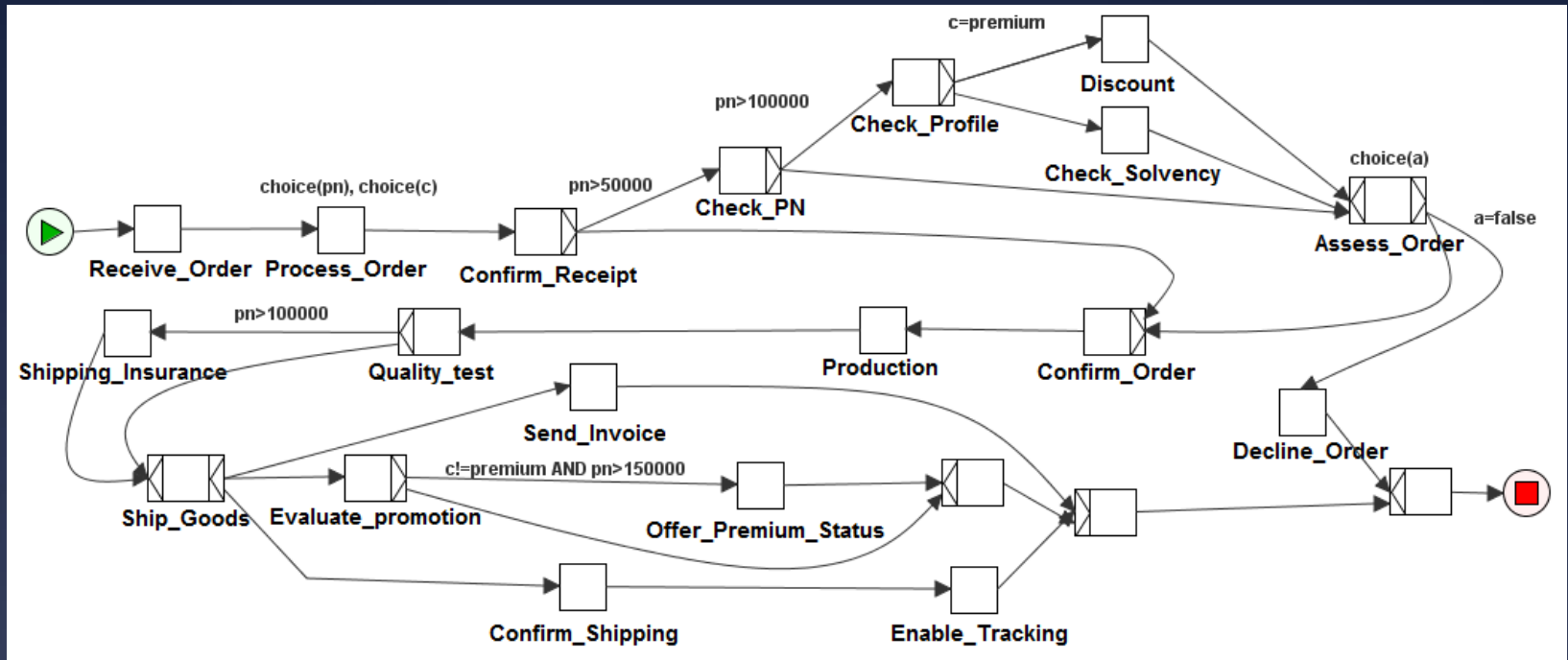
For loop-free workflows the length of the longest run can be used as threshold

BP Verification



- Orders with $pn > 50000$ shall be approved before they are confirmed
- For orders of a non-premium customer with $pn > 80000$ a solvency check is necessary before assessing the order

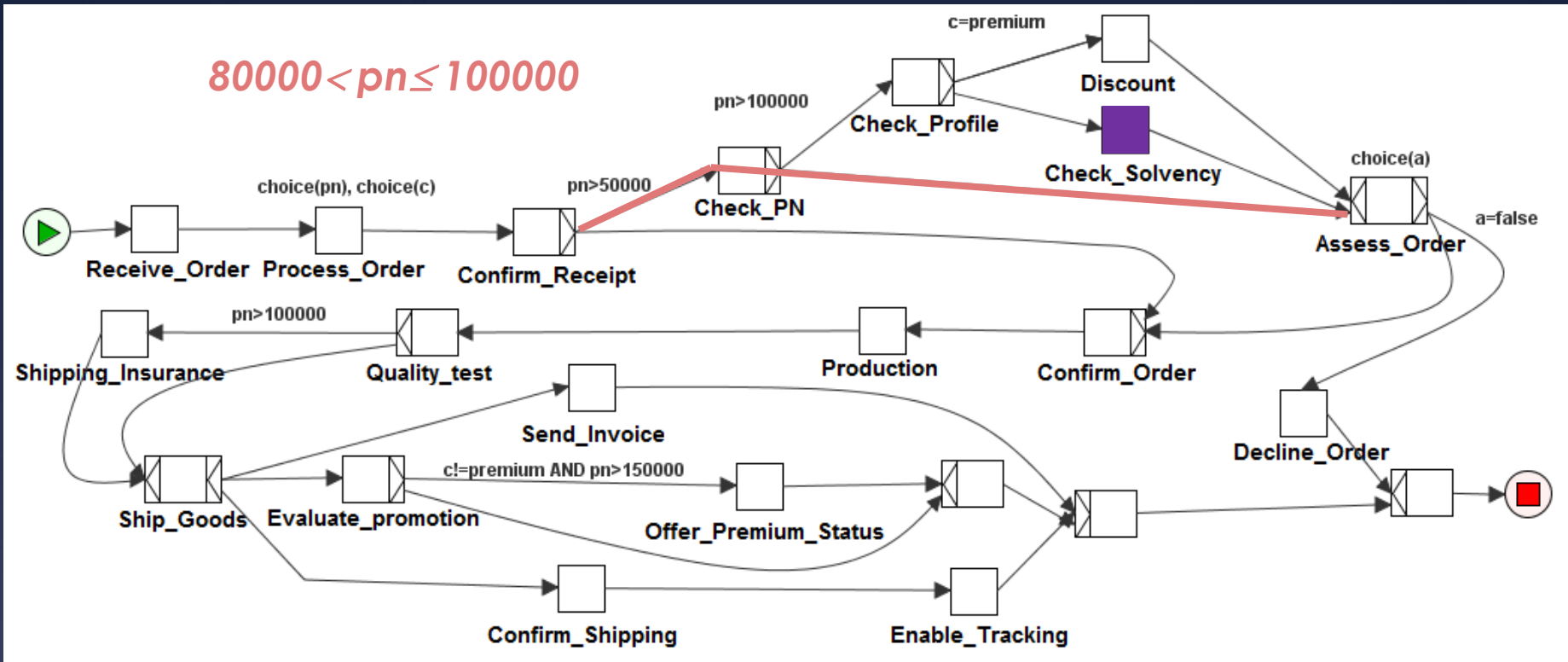
BP Verification



Running the translation in **clingcon** we get (in $\cong 0.1s$)

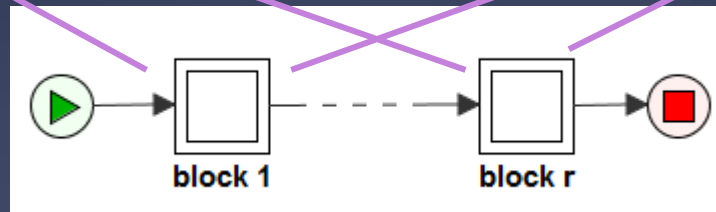
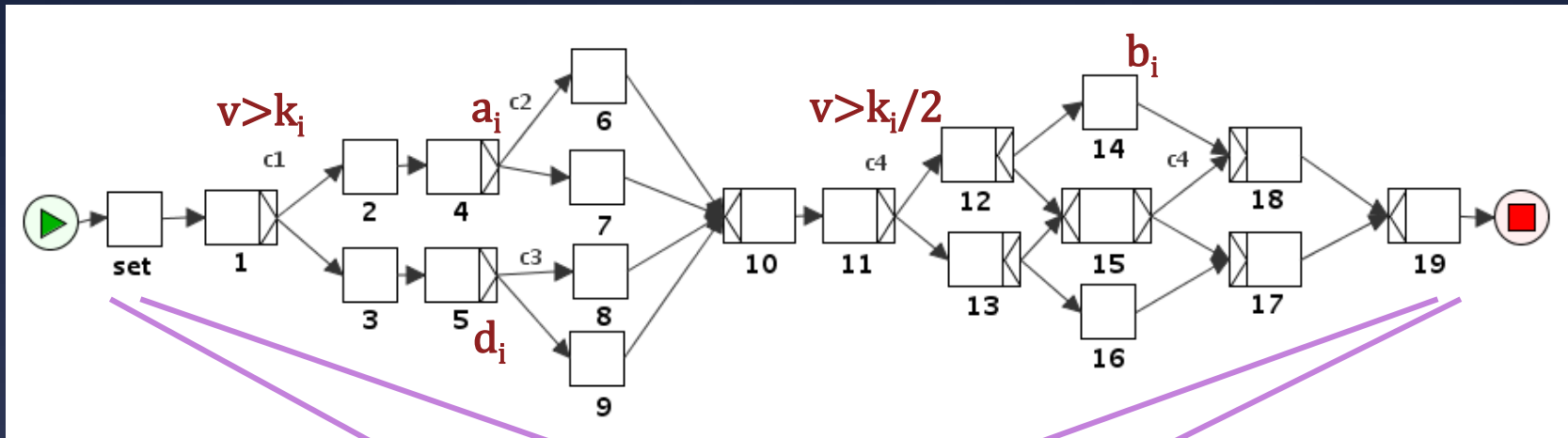
- $\Box(pn > 50000 \wedge \langle Confirm_Order \rangle T \rightarrow a = true)$ **valid**
- $\Box(pn > 80000 \wedge c \neq premium \wedge \langle Assess_Order \rangle T \rightarrow solvency_check_done)$ **non valid**

BP Verification



if *clingcon* is asked to provide *weak answer sets*,
for the *s* after *pn* is assigned,
 $value(pn, s)$ is given the domain $[80001..100000]$

Scalability



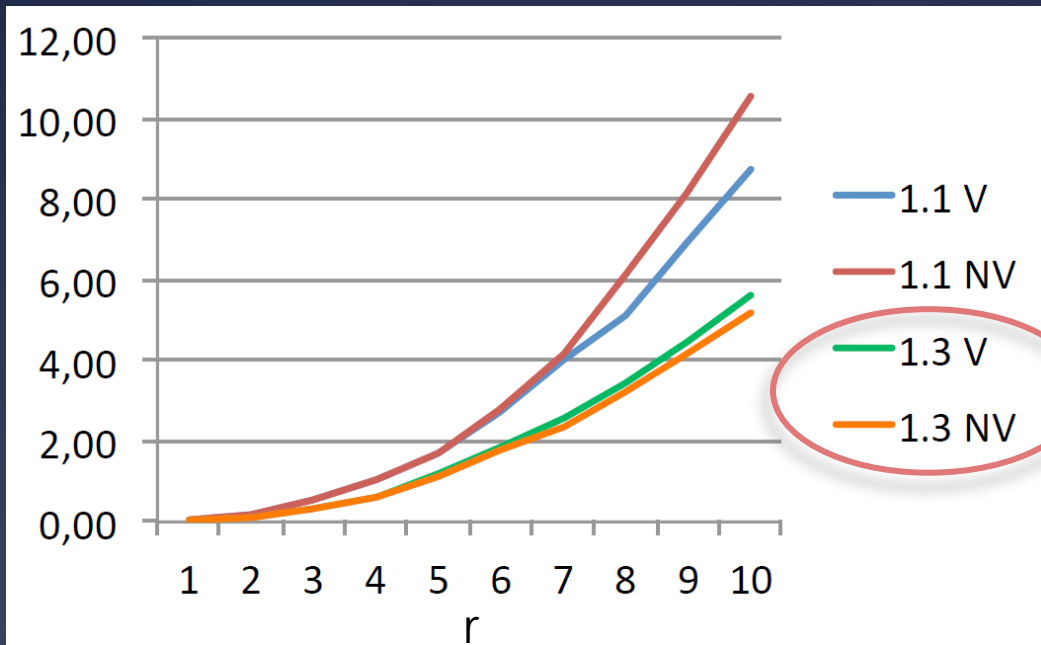
$c1_i$ implies $c4_i$ and c_{i+1} ($k_i > k_{i+1}$), then

$\square(a_1 \rightarrow \diamond b_r)$ **valid**

while

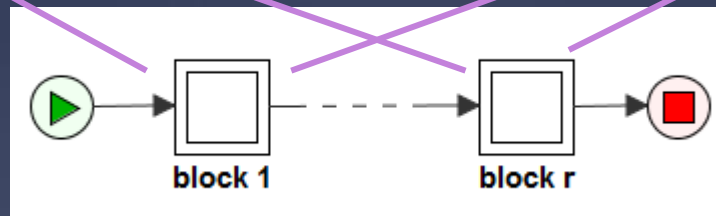
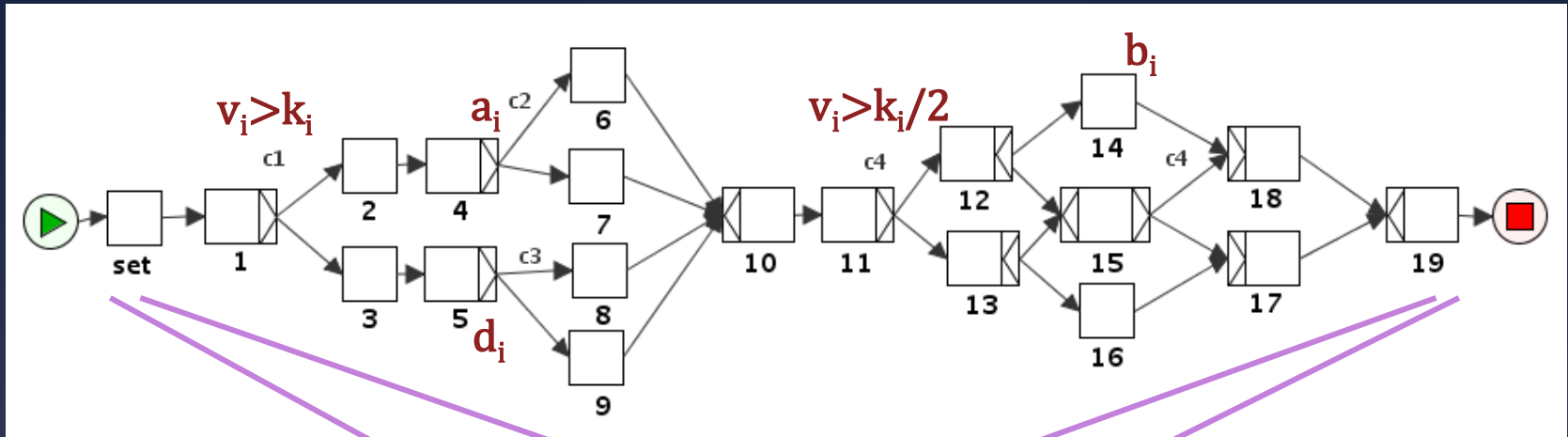
$\square(d_1 \rightarrow \diamond b_r)$ **non valid**

Scalability



Variant with
(pure) ASP conditions
(run in *clingo*)

Scalability



but $c4_r$ is $\wedge(v_i > k_i'/2)$, with $k_i' < k_i$, then, $O(12^r)$ runs and

$$\square(\wedge a_i \rightarrow \diamond b_r)$$

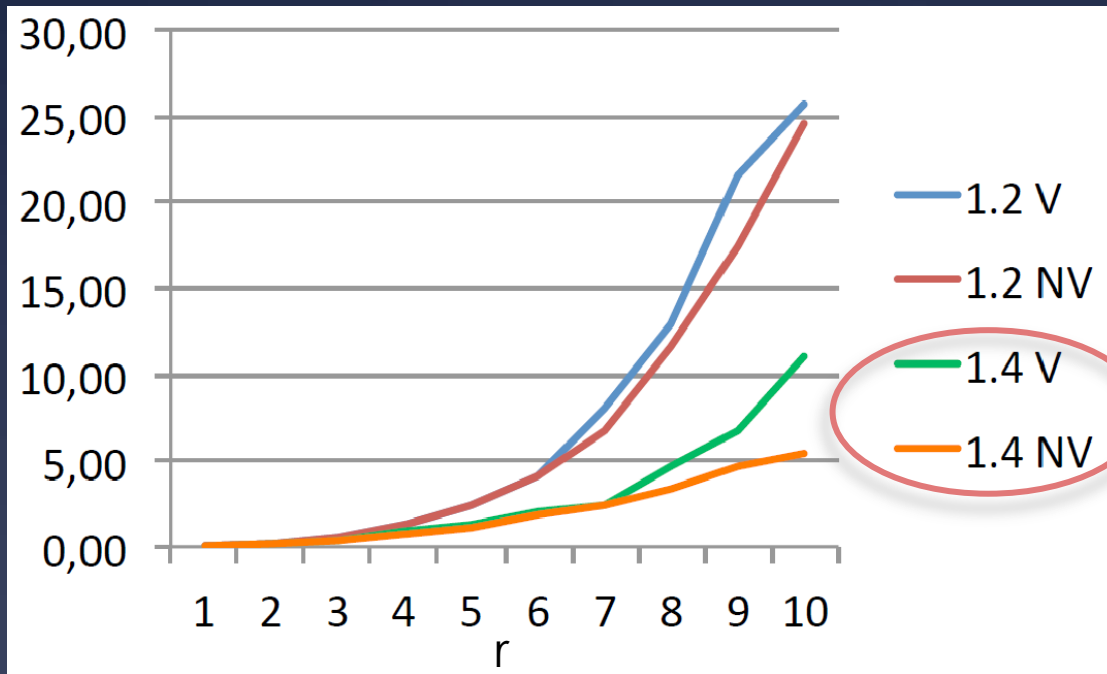
valid

while

$$\square(\wedge d_i \rightarrow \diamond b_r)$$

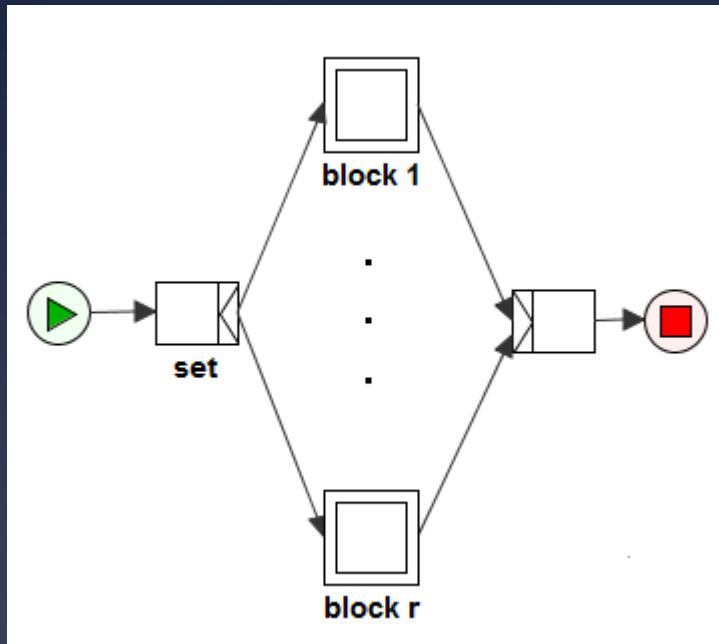
non valid

Scalability



Variant with
(pure) ASP conditions
(run in *clingo*)

Scalability



$c1_i$ is $v > k_i$ ($v_i > k_i$)

$c4_i$ is $v > k_i/2$ ($v_i > k_i/2$)

$\square \wedge (a_i \rightarrow \diamond b_i)$

valid

$\square \wedge (d_i \rightarrow \diamond b_i)$

non valid

$r=4$ $10^{25} \div 10^{28}$ different runs, 10 \div 100 s verification time

$r=5$ $10^{36} \div 10^{40}$ different runs, 100 \div 10000 s verif. time

Conclusions

Constraint Temporal Answer Set Programming

combines temporal logic with:

- Nonmonotonic knowledge representation of actions and change [Giordano, Martelli & T.D. TPLP 13], also suitable for flexible modeling of obligations [Giordano et al ICAIL 13]
- Constraint reasoning

We have shown that current (C)ASP technology already makes the framework useful for **verifying compliance of business processes** (also) **involving conditions on numerical data**