

FBK PDI Activity Report

Report period: June 2016 - January 2017

PDI Work Package involvement

- ❑ WP1 - Foundations of organizational modeling
- ❑ WP2 - Management of event logs
- ❑ WP3 - Hybrid monitoring
- ❑ WP4 - Inference of accurate predictions
- ❑ WP5 - Case studies

WP2

Management of event logs (M1 - M24)

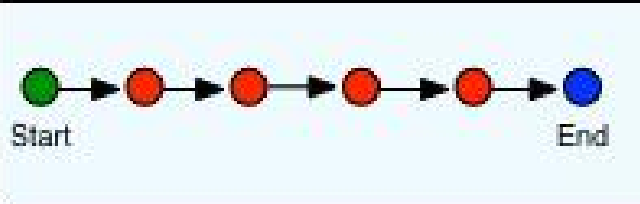
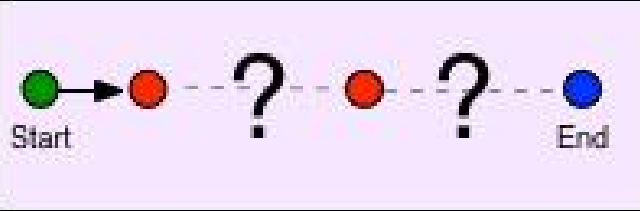

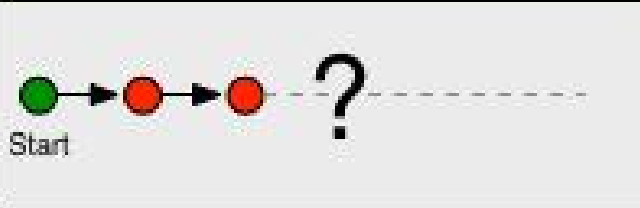
WP2 objective (from DoW)

Development of a comprehensive conceptual framework for the **extraction of** (possibly incomplete) **multi-dimensional event logs** from the legacy data sources of an organization.

Tasks: PDI current focus

- ❑ **T2.1** Definition of a language for the end-user modeling of log mappings. [D2.1, M6]
- ❑ **T2.2** Development of a combined framework for event log extraction by enriching OBDA with log mappings. [D2.1, M6; D2.2, M18]
- ❑ **T2.3** Management of incomplete logs. [D2.2, M18]
- ❑ **T2.4** Development of an architectural extension of the ProM OS backbone towards KOS. [D2.3, M24]

T2.3 incomplete logs: problems

Trace	Reasoning service
	Strong compliance
	Conditional compliance
	Model Consistency
	Runtime monitoring

Incomplete logs: approaches

- ❑ Abduction-based (ECAI '16, AI*IA '16);
- ❑ Planning-based.

Incomplete logs: abduction

An abductive logic program is a triple (KB, A, IC) where:

- KB is a static knowledge base;
- A is a special set of predicates, called *abducibles*;
- IC is a set of integrity constraints.

Given a goal G , the abductive reasoning looks for $\Delta \subseteq A$ such that:

$$KB \cup \Delta \models G \cup IC$$

The set Δ is referred as the *abductive explanation*.

Incomplete logs: abduction

$$KB \cup \Delta \models G \cup IC$$

- ❑ Process model \Rightarrow IC
- ❑ Partial trace \Rightarrow KB
- ❑ $G \Rightarrow \emptyset$
- ❑ $\Delta \Rightarrow$ trace completion

The (first-order logic based) framework is expressive enough to take into account **data** (currently: time).

Incomplete logs: planning

An action language defines a dynamic domain D by means of:

- *Fluents*;
- *actions*;
- *rules*;

A planning problem is a triple:

- dynamic domain D ;
- initial state;
- goal, i.e., the final state.

The solution is a sequence of actions from initial to final state.

Incomplete logs: planning

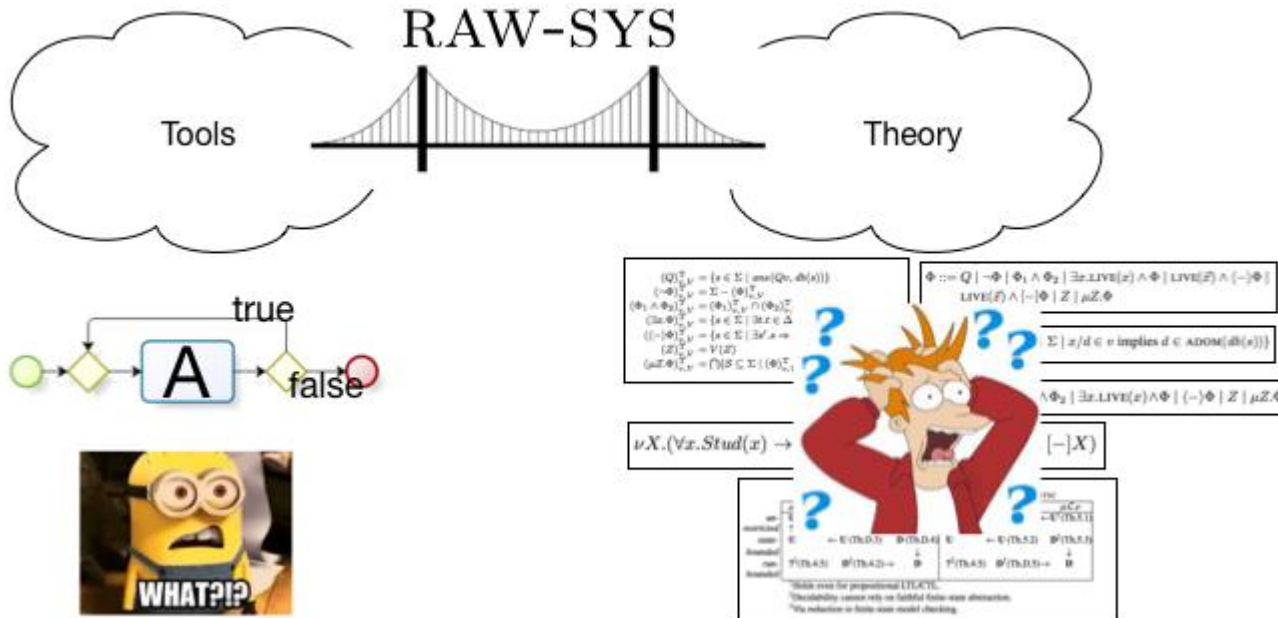
- ❑ Fluents \Rightarrow process states;
- ❑ actions \Rightarrow process activities;
- ❑ rules \Rightarrow workflow + **partial trace**;
- ❑ initial state \Rightarrow process initial state;
- ❑ final state \Rightarrow process final state.

Data (variables) can be encoded as process states.

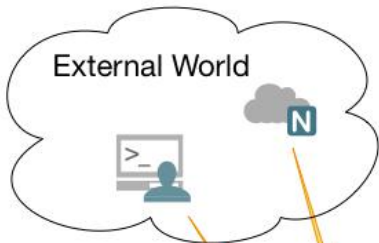
The role of data

Events in the trace can carry a complex payload.

We need an **expressive** and **intuitive** data-aware process model.

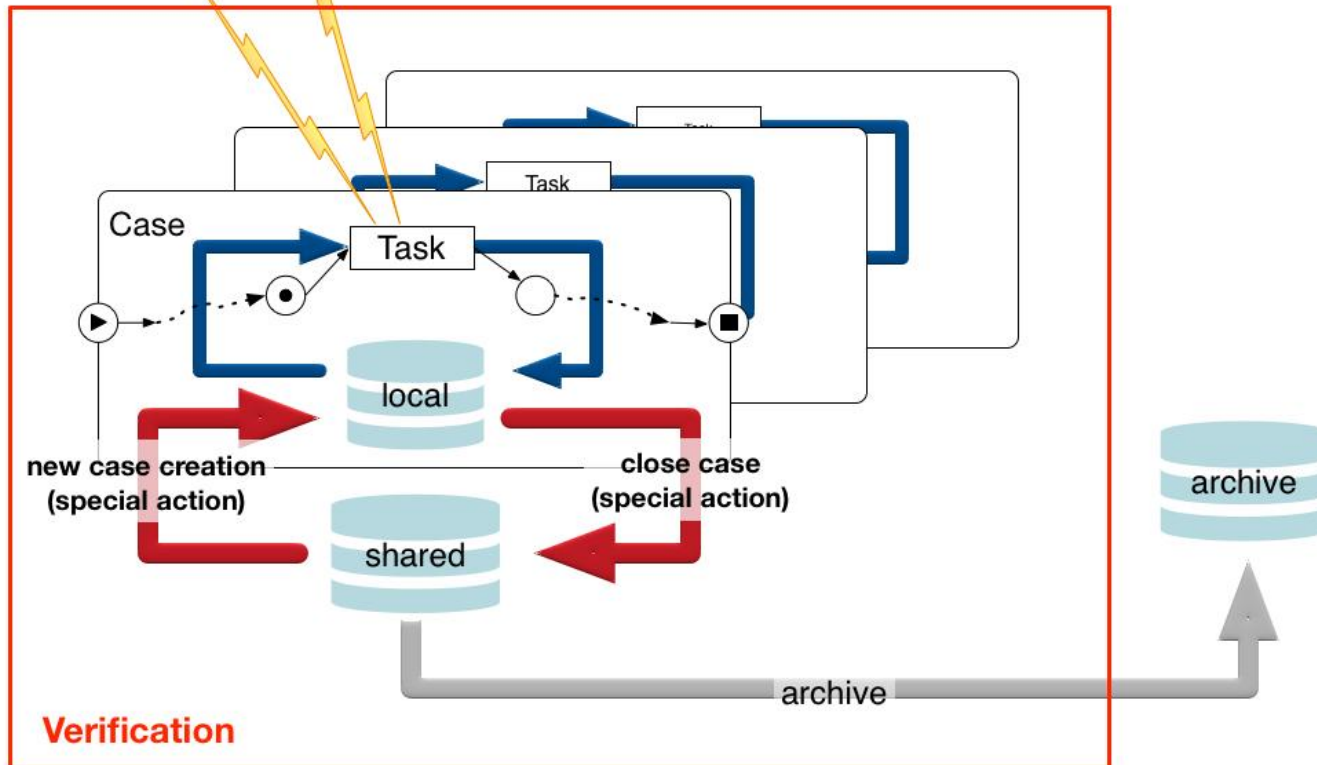


Relational-Aware SYStems



Features:

- ❑ Relational data model;
- ❑ tasks modifies data;
- ❑ conditions on data;
- ❑ "fresh" values can be introduced.



WP3

Hybrid Monitoring (M12 - M36)

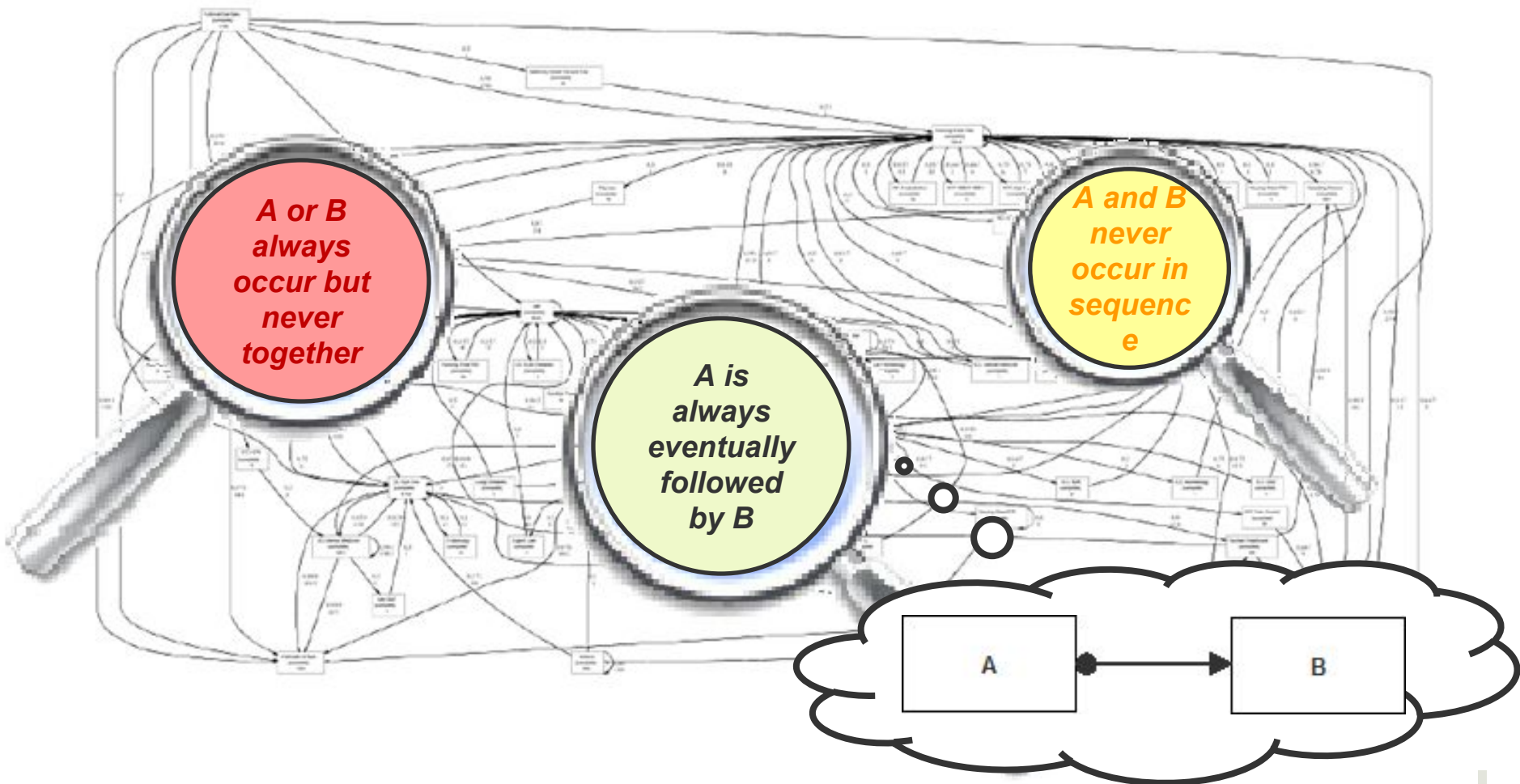
WP3 objective (from DoW)

Definition and development of mechanisms for the **hybrid monitoring** (combining *symbolic and statistical*) of processes.

Tasks: PDI current focus

- ❑ **T3.1** Definition of a framework to combine symbolic and statistical approaches for process monitoring. [D3.1, M24]
- ❑ **T3.2** Analysis and tuning of existing machine learning techniques for hybrid monitoring approaches. [D3.2, M30]
- ❑ **T3.3** Extracting symbolic representations of properties from data. [D3.3, M36]
- ❑ **T3.4** Concept drift detection. [D3.3, M36]

Declarative process discovery



What is already there?



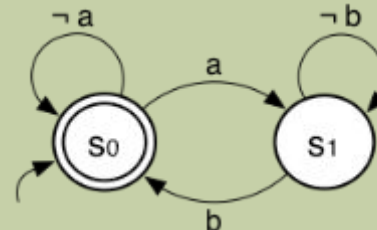
Declare Miner

**Apriori
Algorithm**

{a, b}

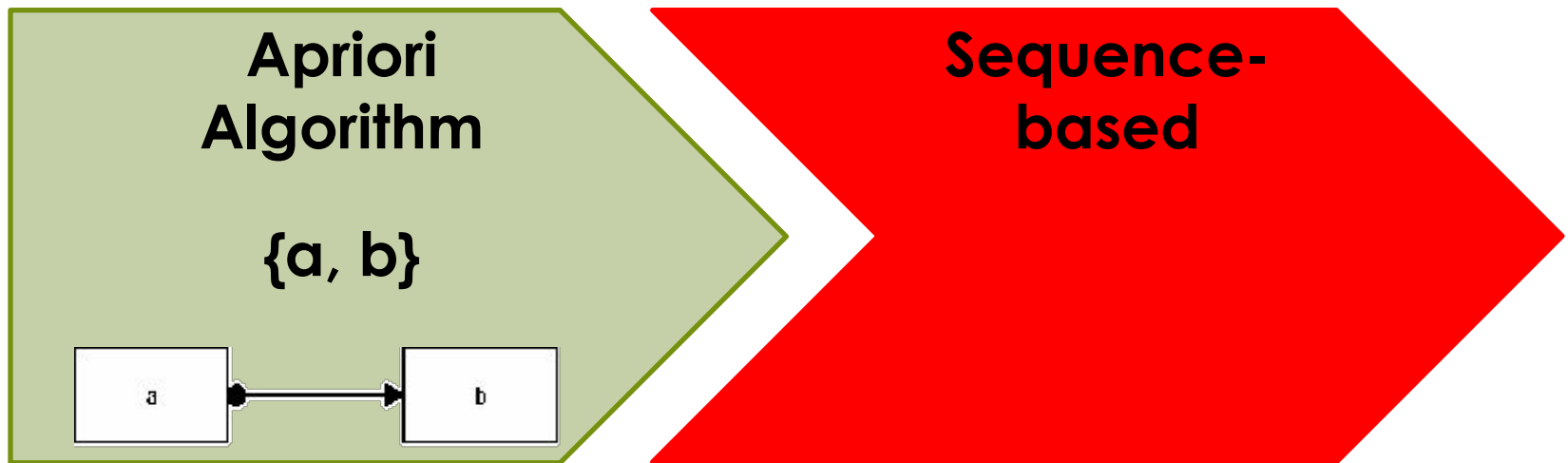


**Automata-based
constraint
check**



A new approach

Faster Miner (EDOC'16)



Apriori Algorithm

\mathcal{L} on Σ

<a b c j b b d a>
<a b b c d a>
<a b b i i a c f>
<a j j e e>
<a b b c j e f b>



Frequent activity sets

$A_1, A_2, \dots, A_n \subseteq \Sigma^*$

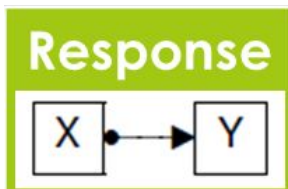
$A_k = \{a, b\}$

- ❑ **Support:** percentage of traces containing the set of activities
- ❑ Frequent Activity Set \rightarrow **high support** ($>$ `supp_threshold`)

Sequence Analysis

*Frequent activity set
instantiated on a
Declare template*

*Frequently satisfied
constraints*



(a,b)

(b,a)

(a,c)

(d,a)

(a,j)

(j,a)

...

- ❑ An algorithm for each Declare template
- ❑ Further optimizations: Pruning and Multithreading

Open Points

- ❑ Deliverable format and submission
- ❑ ProM plugin for log incompleteness

References

- [ECAI'16] Chesani, De Masellis, Di Francescomarino, Ghidini, Mello, Montali, Tessaris: Abducting Workflow Traces: A General Framework to Manage Incompleteness in Business Processes. ECAI 2016: 1734-1735
- [AI*IA'16] Chesani, De Masellis, Di Francescomarino, Ghidini, Mello, Montali, Tessaris: Abducting Compliance of Incomplete Event Logs. AI*IA 2016: 208-222
- [CILC'16] Chesani, De Masellis, Di Francescomarino, Ghidini, Mello, Montali, Tessaris: Abducting Workflow Traces: a General Framework to Manage Incompleteness in Business Processes, CILC2016

References

- [AAAI'17] De Masellis, Di Francescomarino, Ghidini, Montali, Tessaris. *Add Data into Business Process Verification: bridging the gap between theory and practice*. Conference on Artificial Intelligence, AAAI 2017, (to appear).
- [EDOC'16] Kala, Maggi, Di Ciccio, Di Francescomarino: *Apriori and Sequence Analysis for Discovering Declarative Process Models*. EDOC 2016: 1-9