## RuM: declarative process mining, distilled



Anti Alman, U. Tartu Claudio Di Ciccio, Sapienza Fabrizio M. Maggi, U. Bolzano Marco Montali, U. Bolzano Han van der Aa, U. Mannheim







### The Team



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University of Tartu



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Marco Montali
Free University of
Bozen-Bolzano



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### Today

#### **Outline**

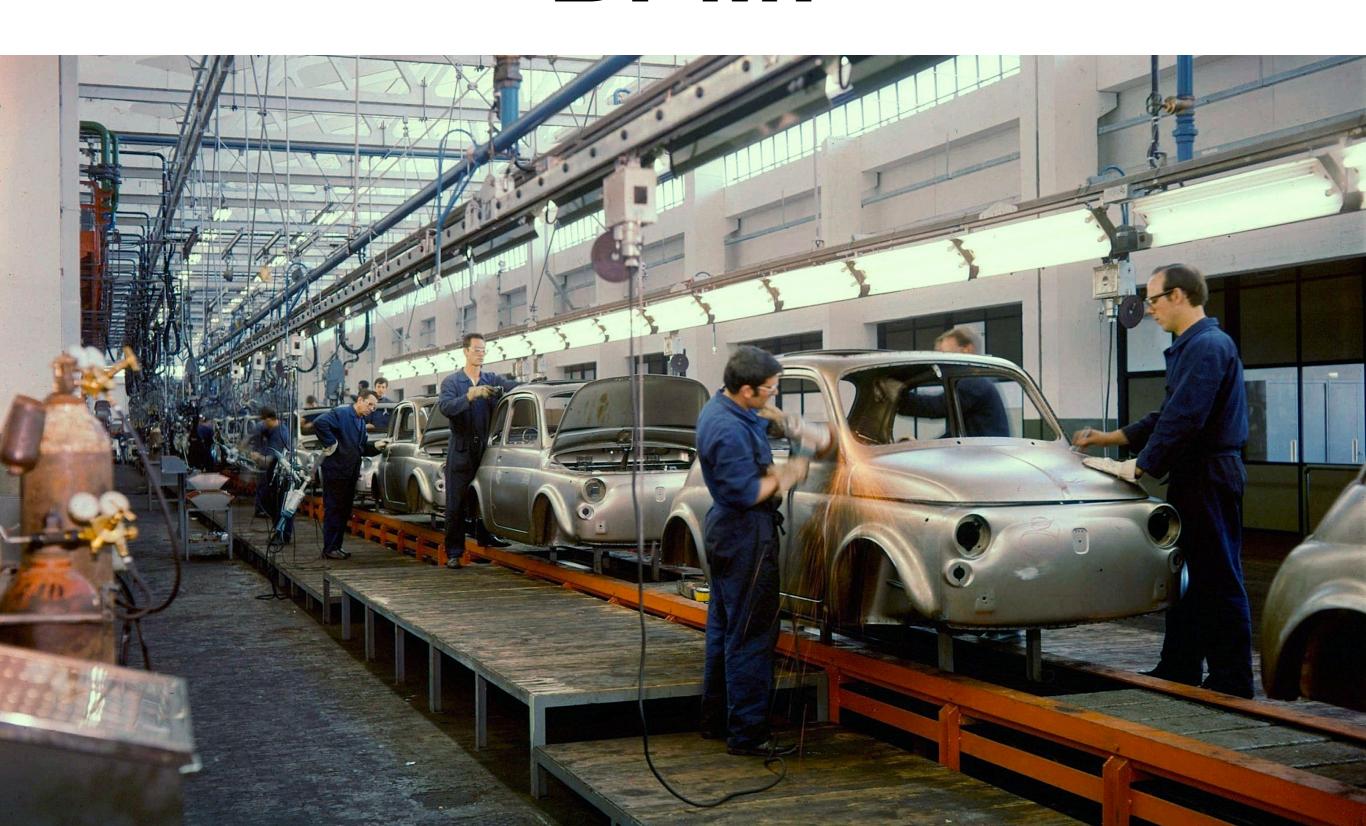
- Introduction to declarative process mining
- Declarative Process Mining with RuM
- Research opportunities

#### Setting

- Tutorial is hands-on using the RuM tool (download at: <u>www.rulemining.org</u>)
- Questions are welcome at any moment

### Introduction to declarative process mining

### BPM!



### BPM?

































### Types of Processes

### Not all processes are the same

- Even within one organization, processes can be very different in terms of their essential properties.
- Processes can be characterized through three dimensions:
  - complexity
  - predictability
  - repetitiveness

## Complexity

Degree of difficulty in collaboration, coordination, and decision making

- Low complexity: exchanging personal email messages and handling travel requests
- High complexity: handling medical treatments

### Predictability

How easy it is to **determine in advance** the way the process will be executed.

- Low predictability: exchanging personal email messages.
- High predictability: handling travel requests.

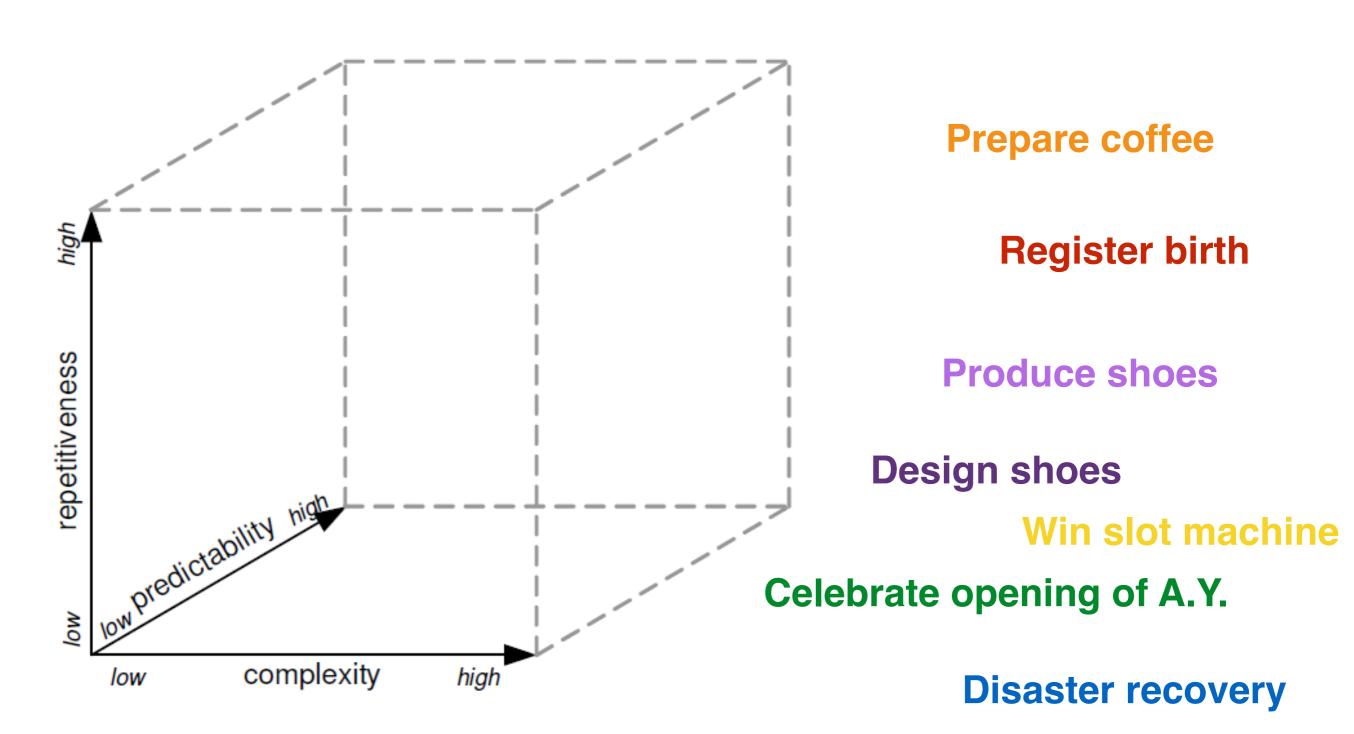
### Repetitiveness

Frequency of process execution.

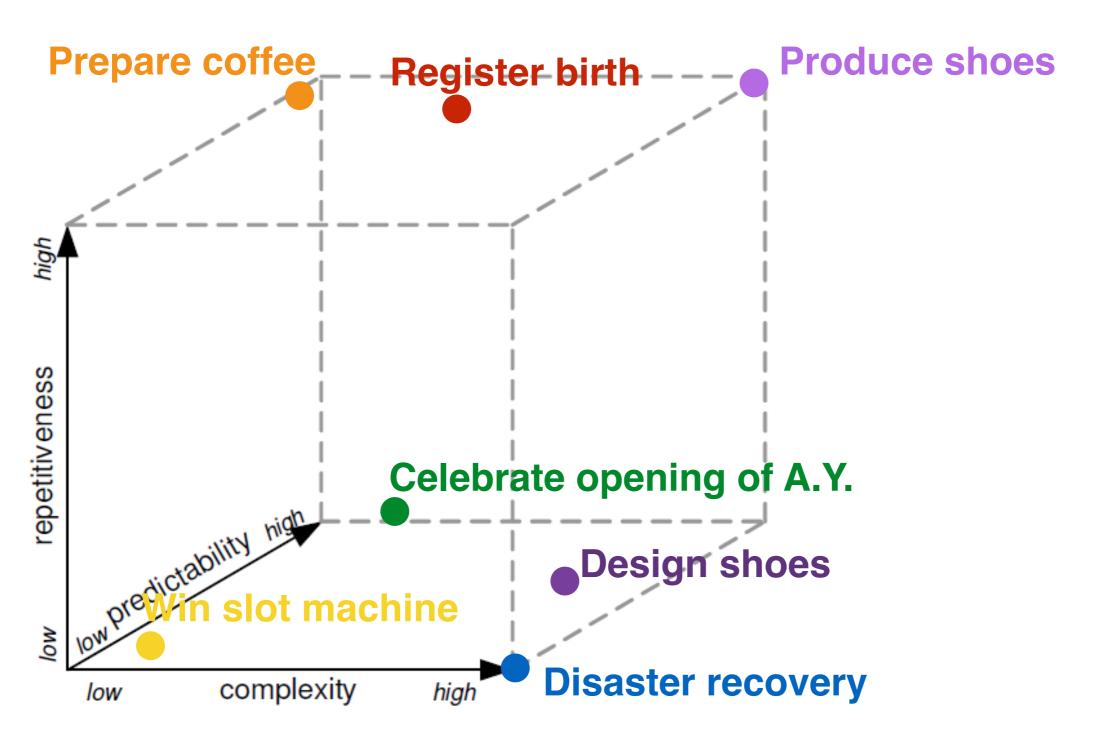
A business process that is executed once per year has a lower degree of repetitiveness than a process executed twice a day.

- Low repetitiveness: disaster handling.
- High repetitiveness: exchanging personal email messages.

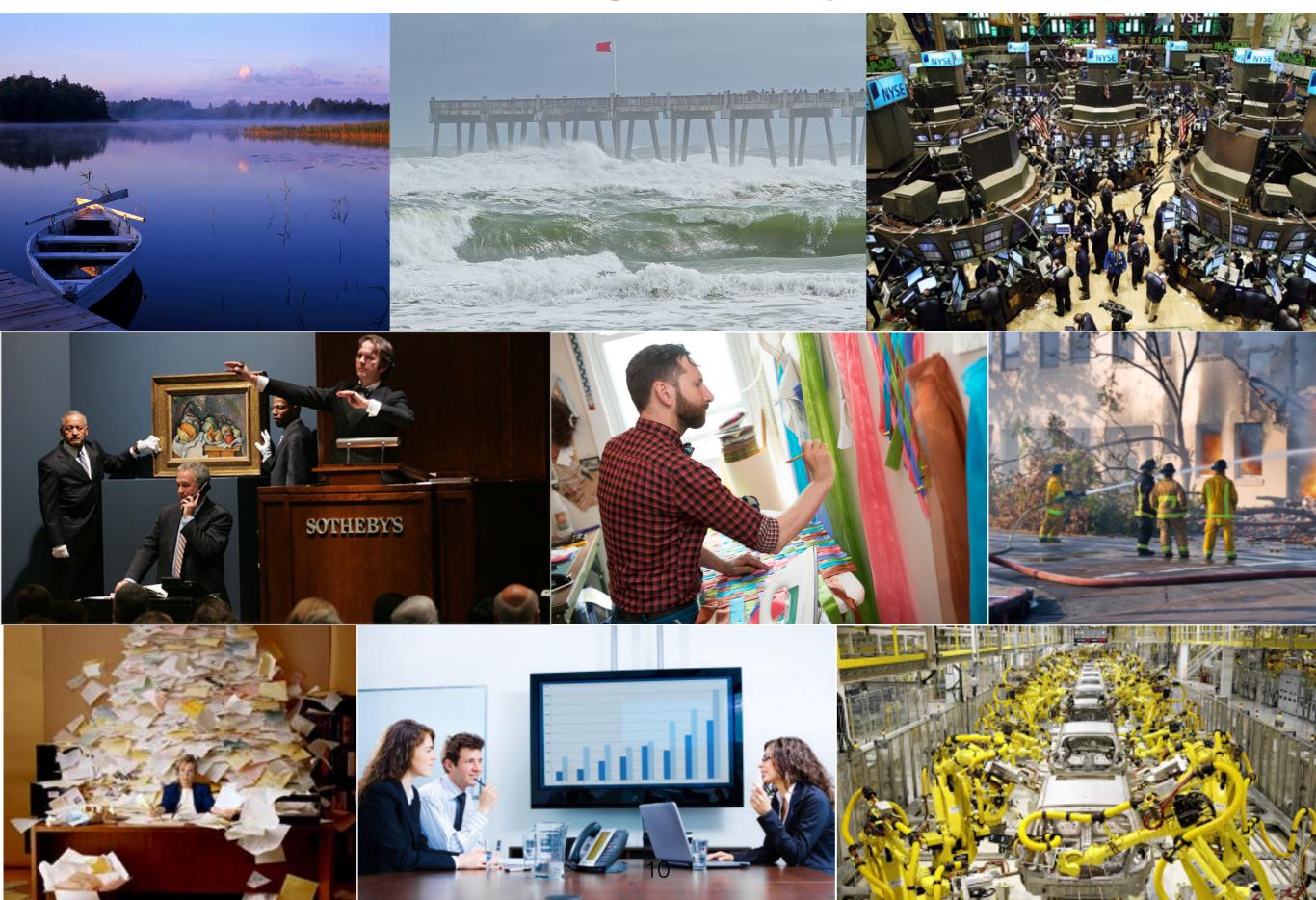
## Classifying processes



## Classifying processes



## Environment!

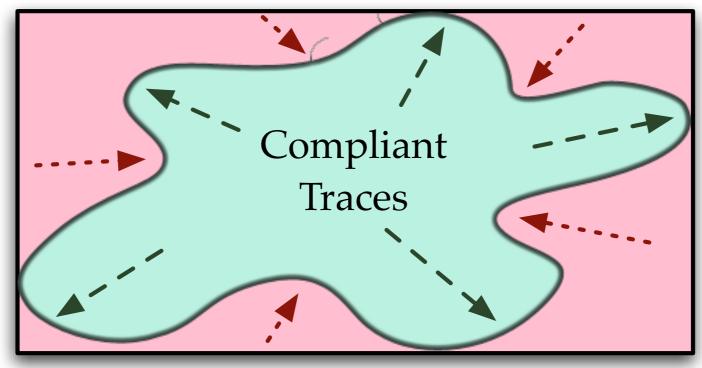


## Flexibility vs support

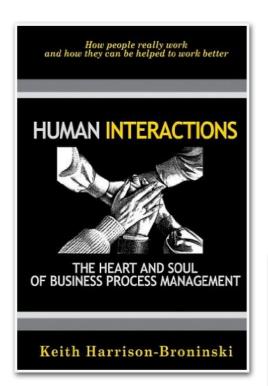
**Flexibility**: degree to which users can make local decisions about how to execute processes.

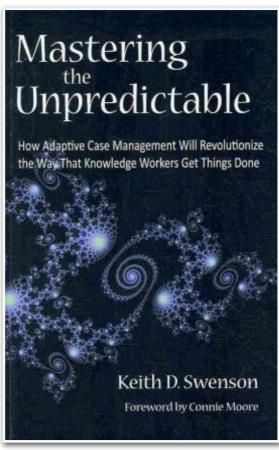
**Support**: degree to which a system makes centralized decisions about how to execute processes.

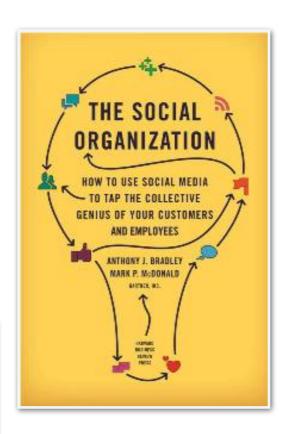
Universe of Traces

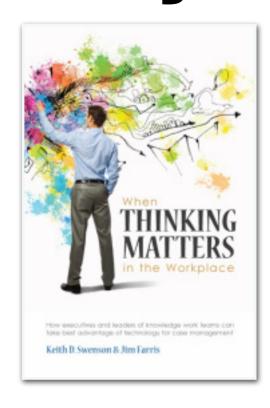


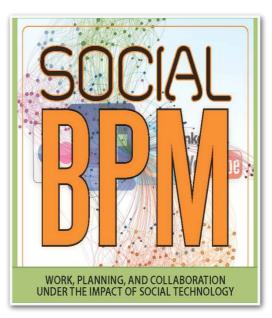
## The issue of flexibility



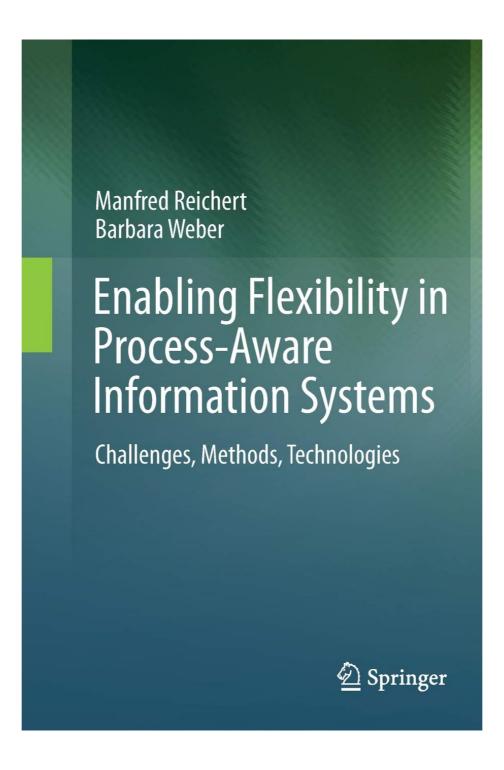








## The issue of flexibility



Our focus:

flexibility by design

## Two paradigms

### Imperative modeling

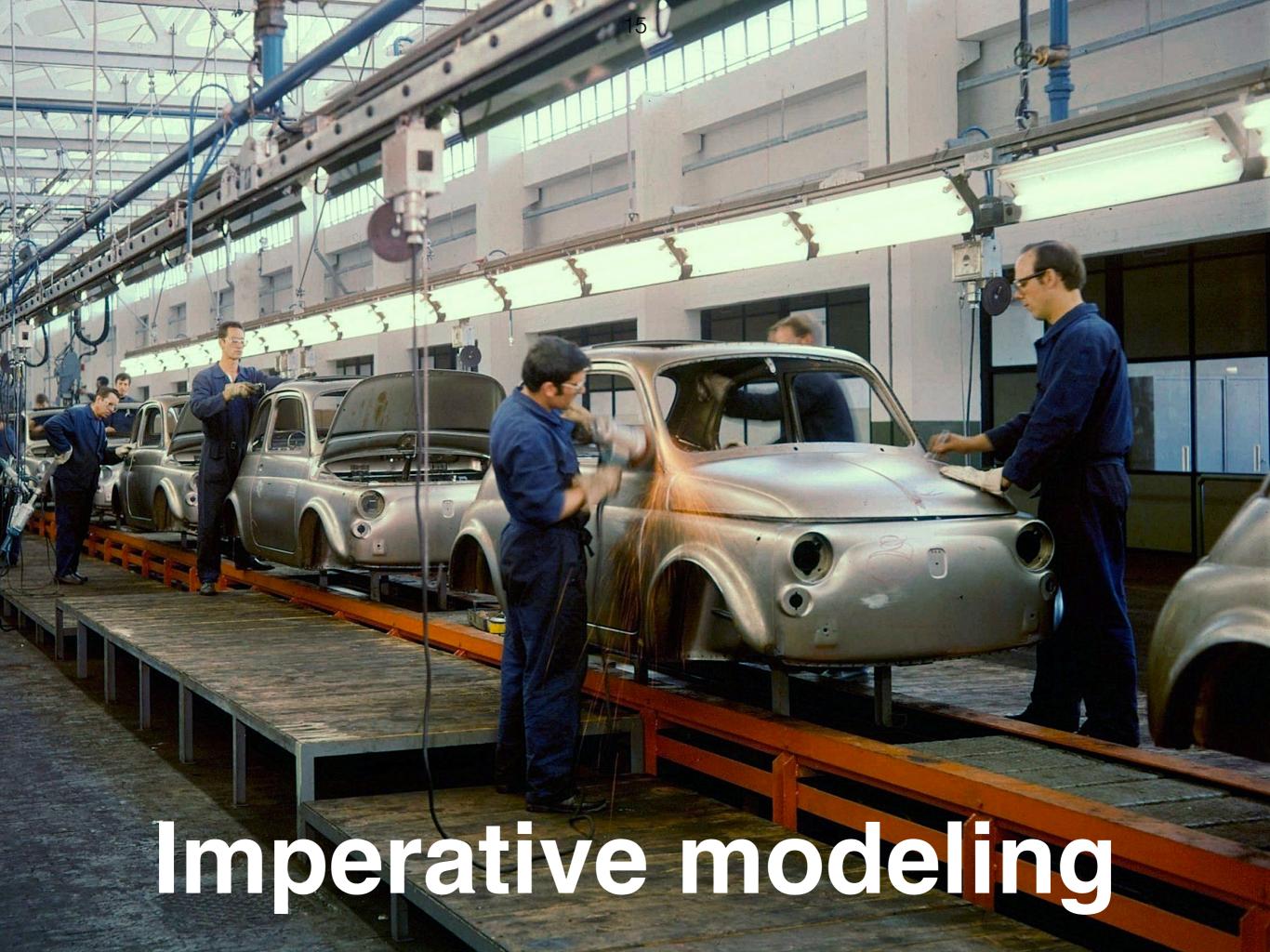
- Traditional repetitive BPs
- Metaphor: flow-chart

## Decla prive modeling

- Highly-variable BPs
- Metaphor: rules/constraints







### Idea

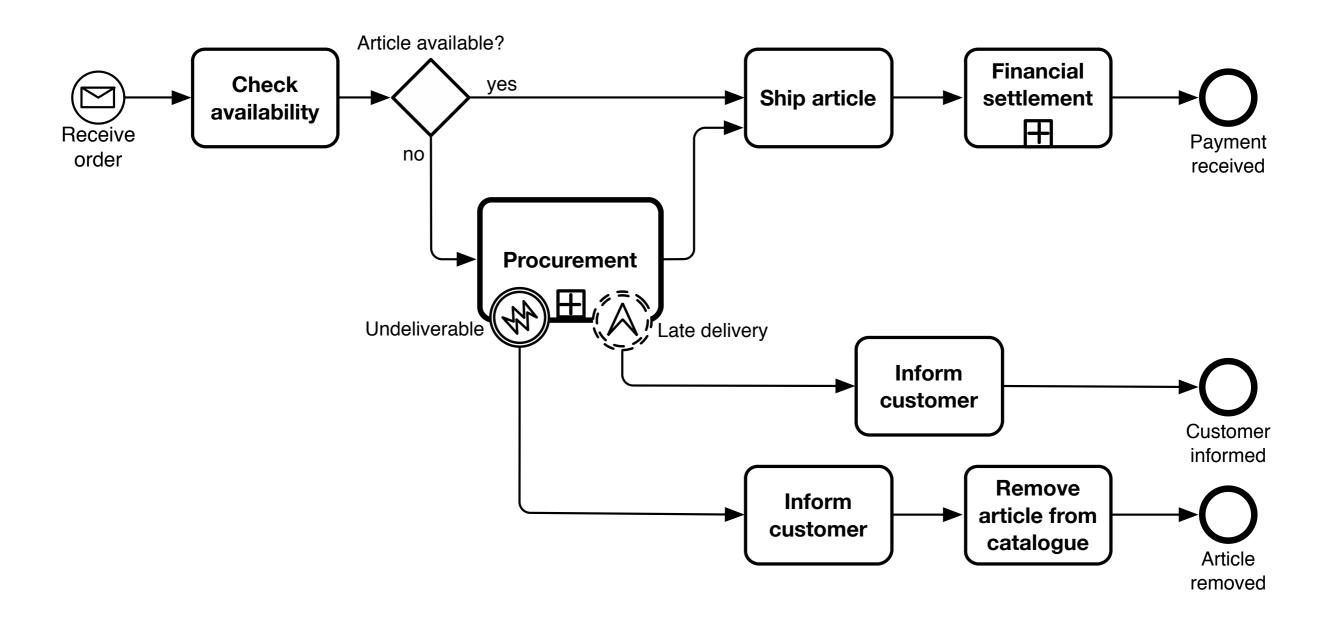
Focus: **how** things must be done

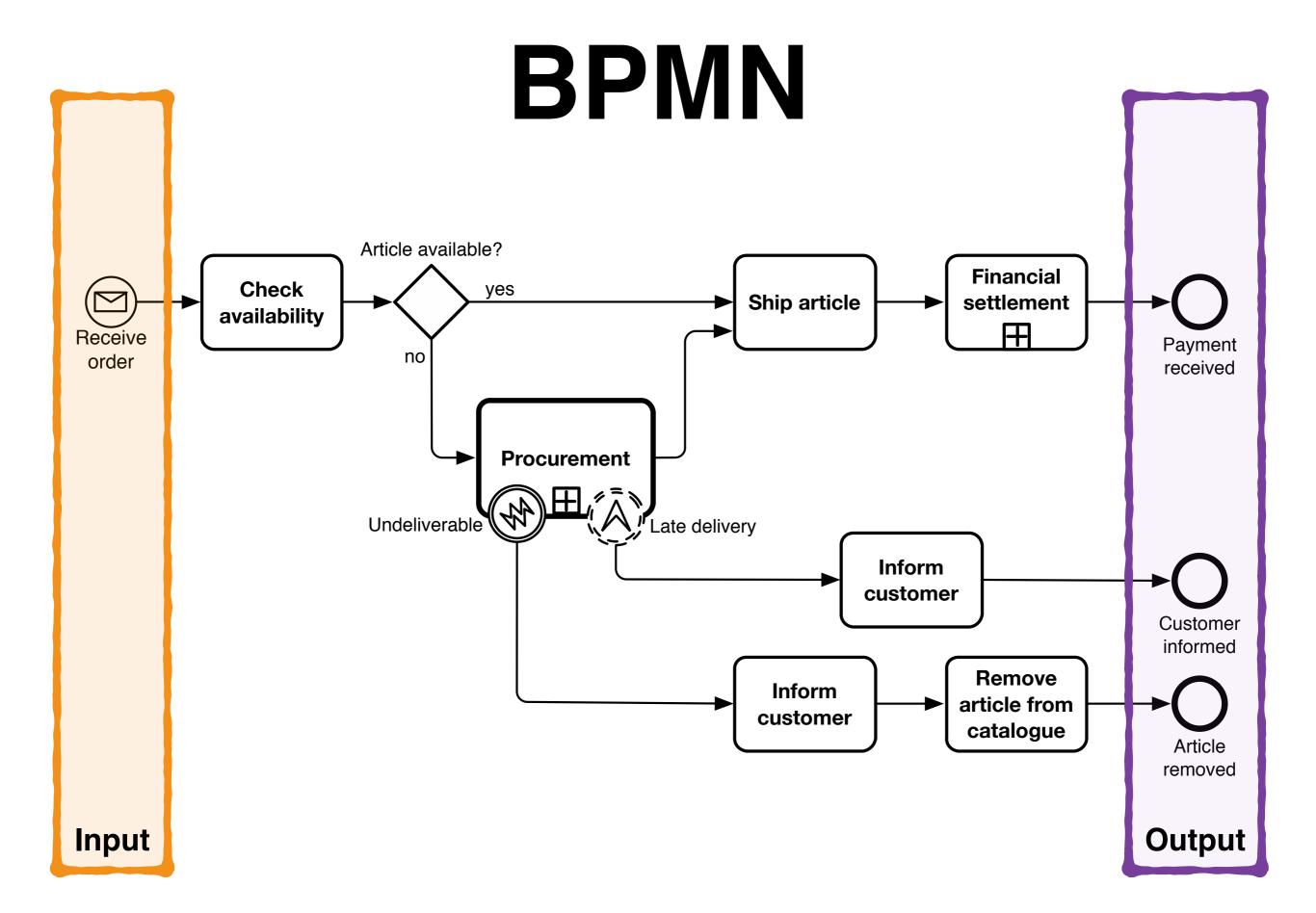
Explicit description of the process control-flow

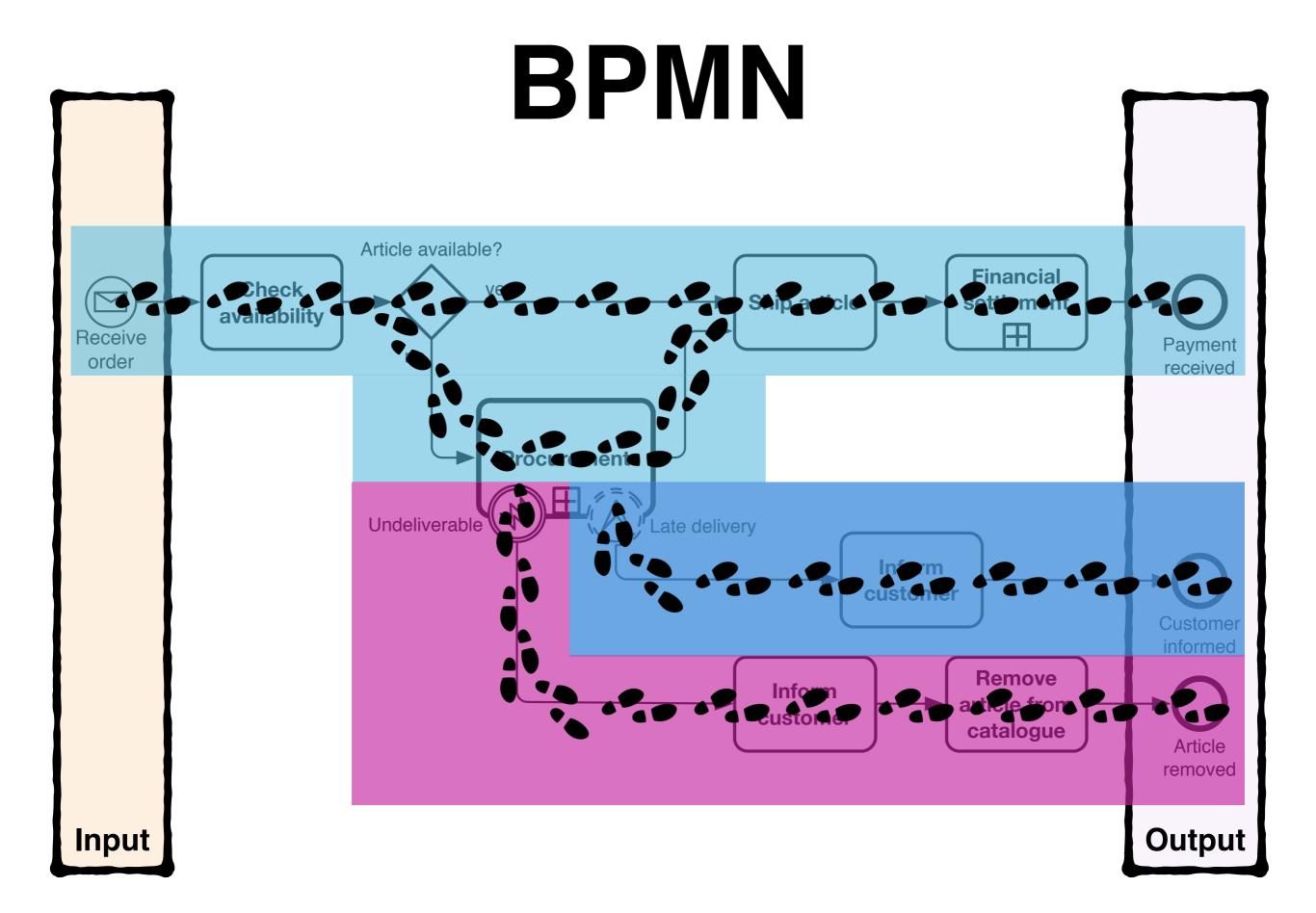
## Closed modeling

- All that is not explicitly modeled is forbidden
- Exceptions/uncommon behaviors have to be explicitly enumerated at design time

### BPMN



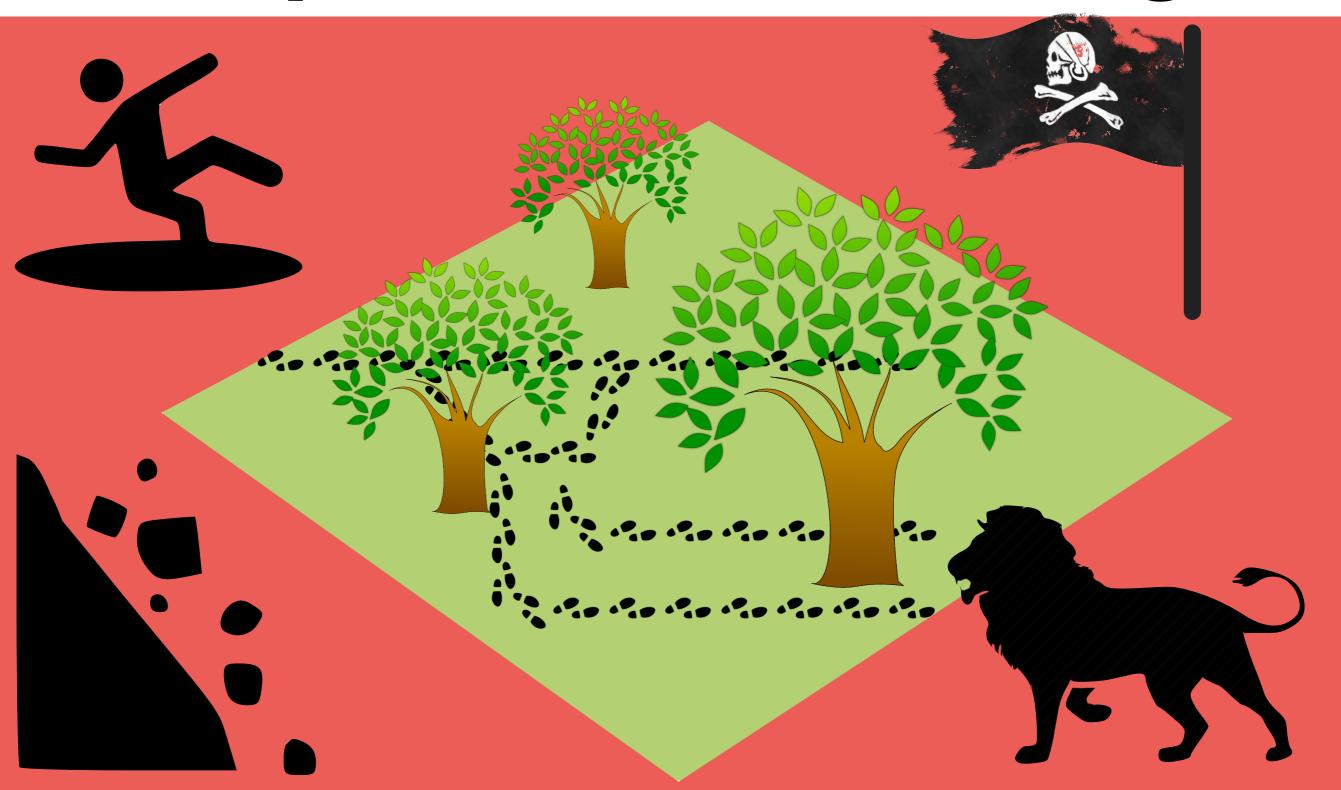




### Organizational boundaries



## Imperative modeling



# Example: rigid shopping

- Tasks:  $\Sigma = \{\text{pick item, close order, pay}\}$
- A customer starts the process by picking one or more items. She then decides to quit or close the order. In the latter case, she has to pay for the order.

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- A customer starts the process by picking one or more items. She then decides to quit or close the order. In the latter case, she has to pay for the order.
- Which traces are accepted?
  - <> (empty trace)
  - <i,i,i>
  - . <i,i,i,c,p>
  - <i,i,i,c,p,p>
  - <i,i,i,p,c>
  - <i,c,p,i,i,c,p>

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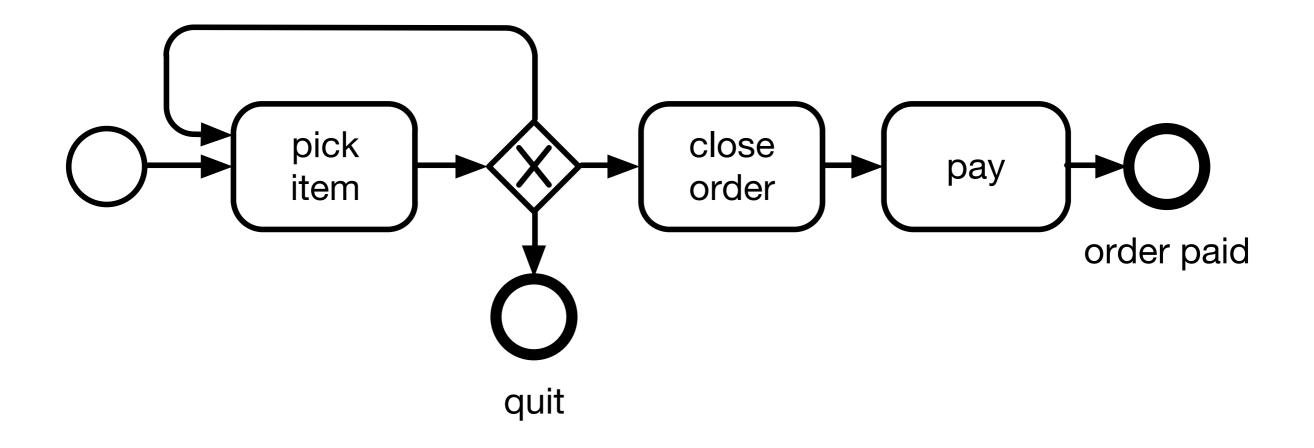
Can you model this process in BPMN?

How difficult is it?

A customer starts the process by picking one or more items. She then decides to quit or close the order. In the latter case, she has to pay for the order.

Can you model this process in BPMN?

How difficult is it?



# Example: flexible shopping

- Tasks:  $\Sigma = \{\text{pick item, close order, pay}\}$
- A customer may pick items, close several orders, and pay one or more orders at once.
- She may even close an empty order and pay for it, or pay multiple times (but this will result in a no-op)
- Business constraint: whenever you close an order, eventually you have to pay.

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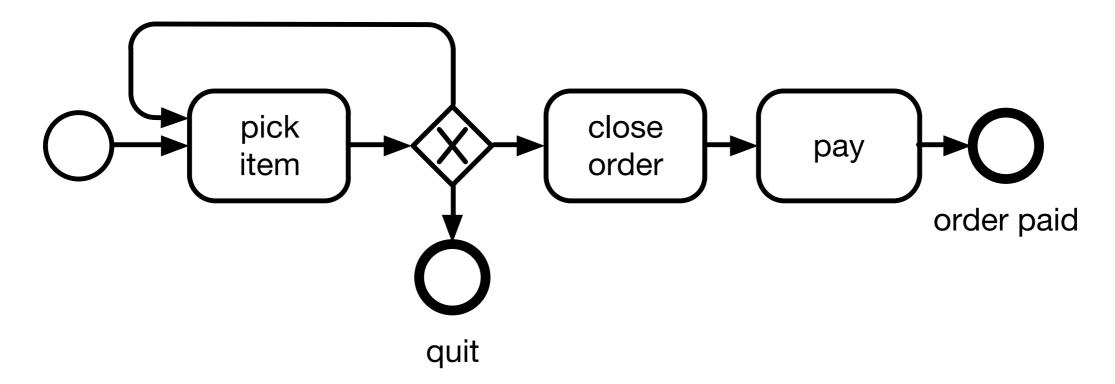
Can you model this process in BPMN?

How difficult is it?

You may pick items, close and pay an order. Whenever you close an order, eventually you have to pay.

Can you model this process in BPMN?

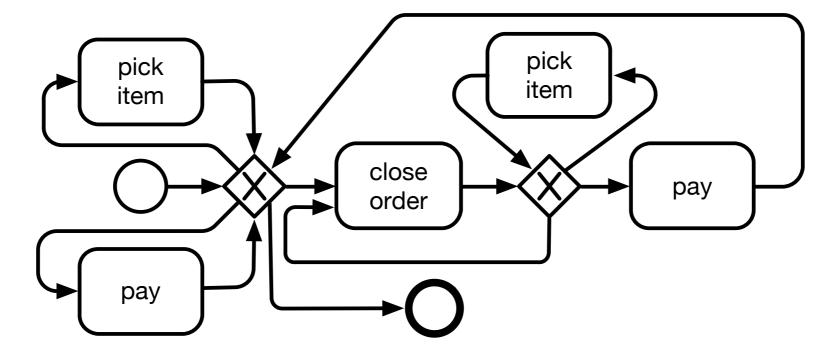
How difficult is it?



simple, correct, but not general enough!

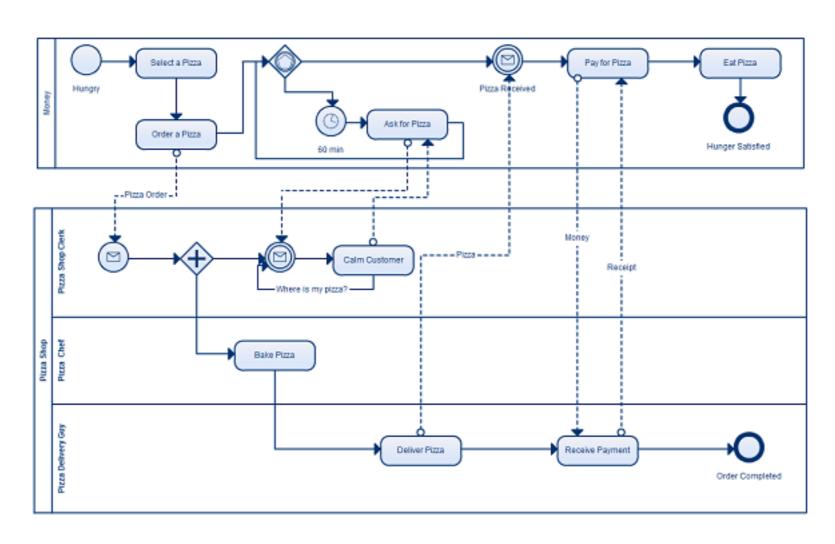
Can you model this process in BPMN?

How difficult is it?



correct, general, but unreadable!

### Favourite Italian food?



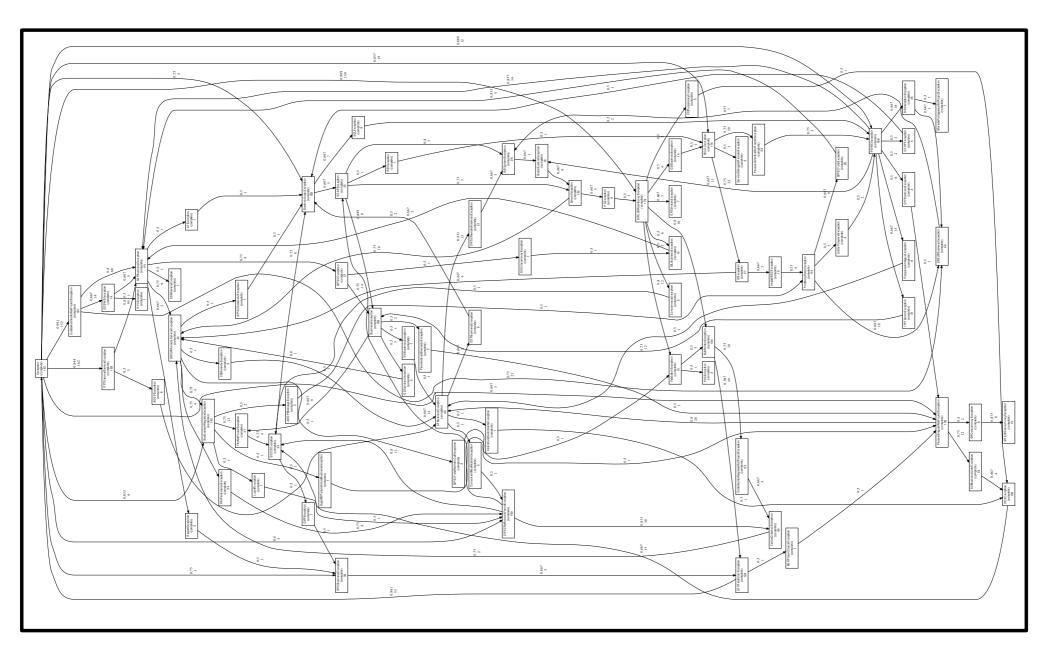
Order-to-delivery

### Favourite Italian food?



Order-to-delivery

### Favourite Italian food?



Healthcare

#### Favourite Italian food?

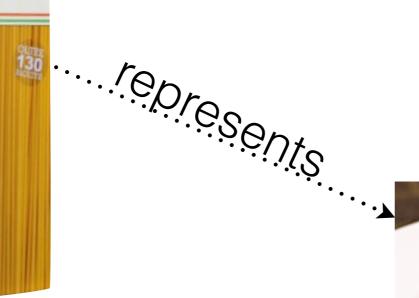


Healthcare

# Our goal



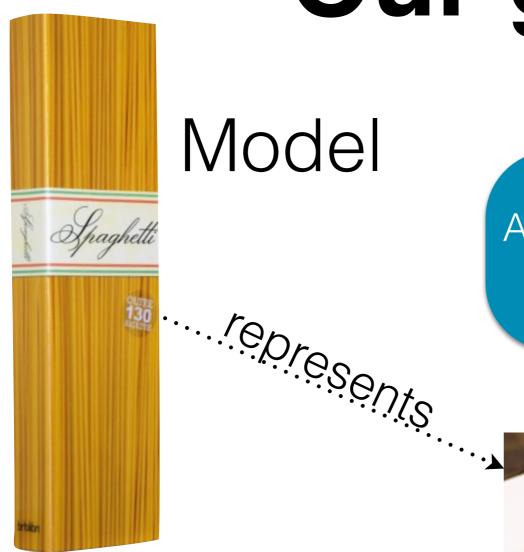
Model



Reality



# Our goal

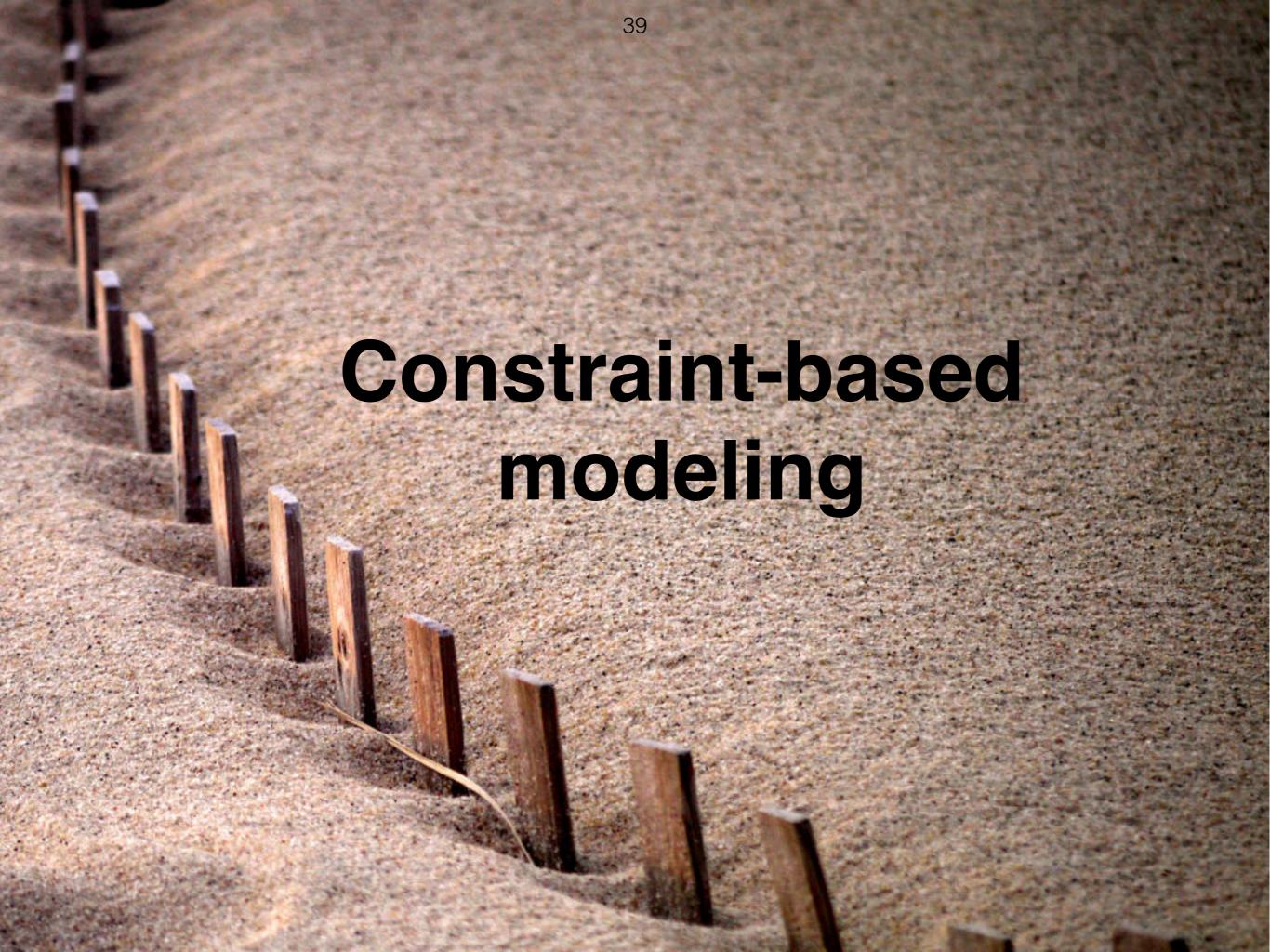


Model

A class of spaghetti models, not all of them!

Reality





#### Idea

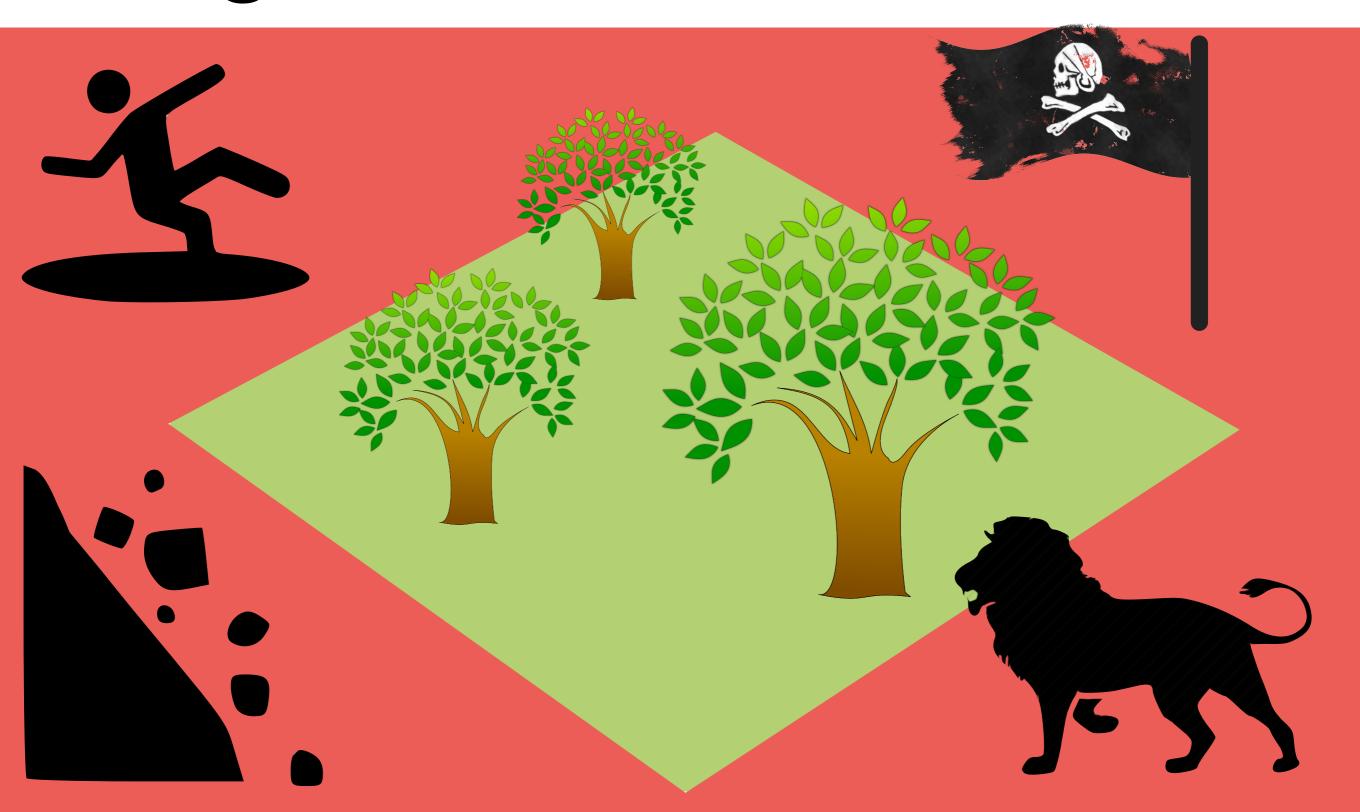
Focus: What has to be accomplished

 Explicit description of the relevant business constraints (behavioral constraints, best practices, norms, rules, ...)

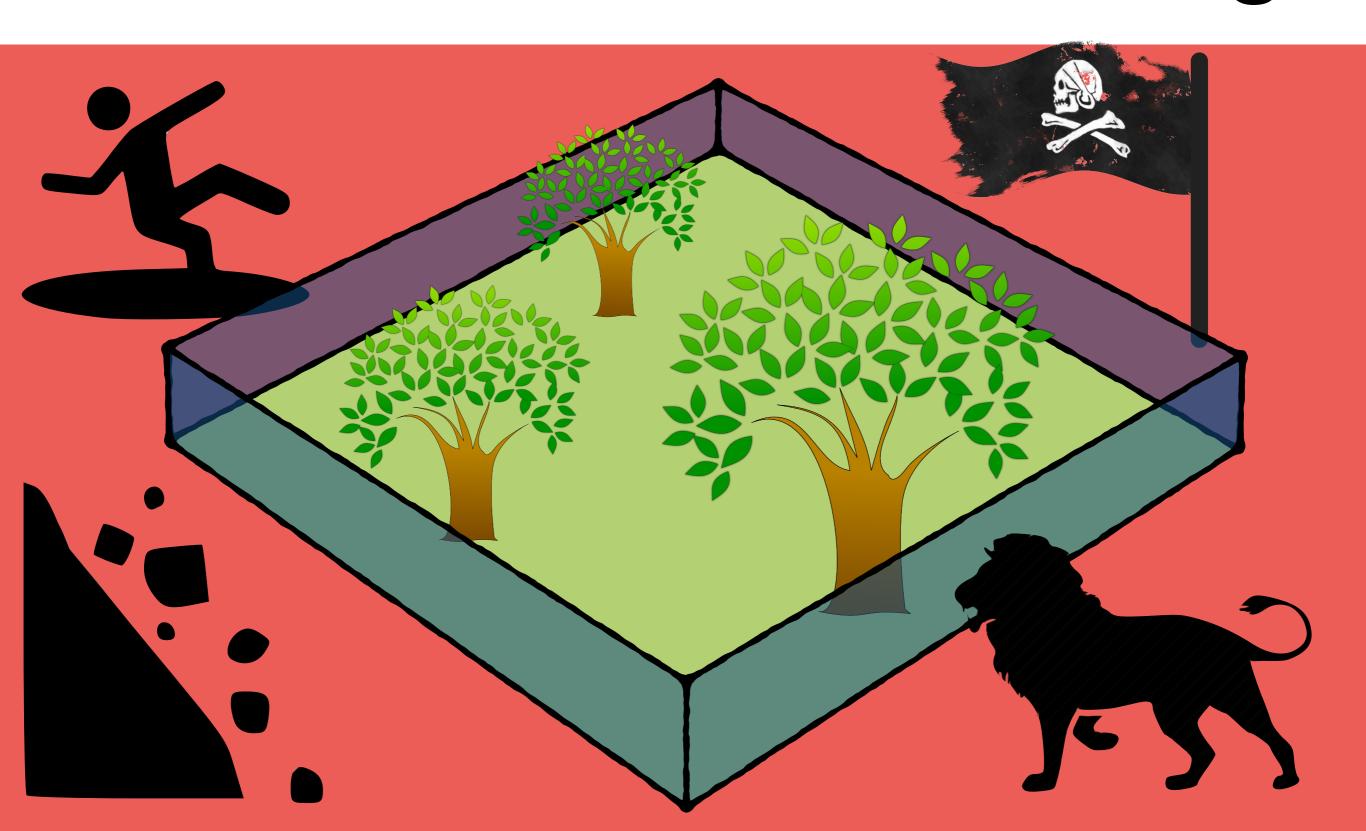
# Open modeling

- All behaviors are possible unless they are explicitly forbidden
- Control-flow left implicit

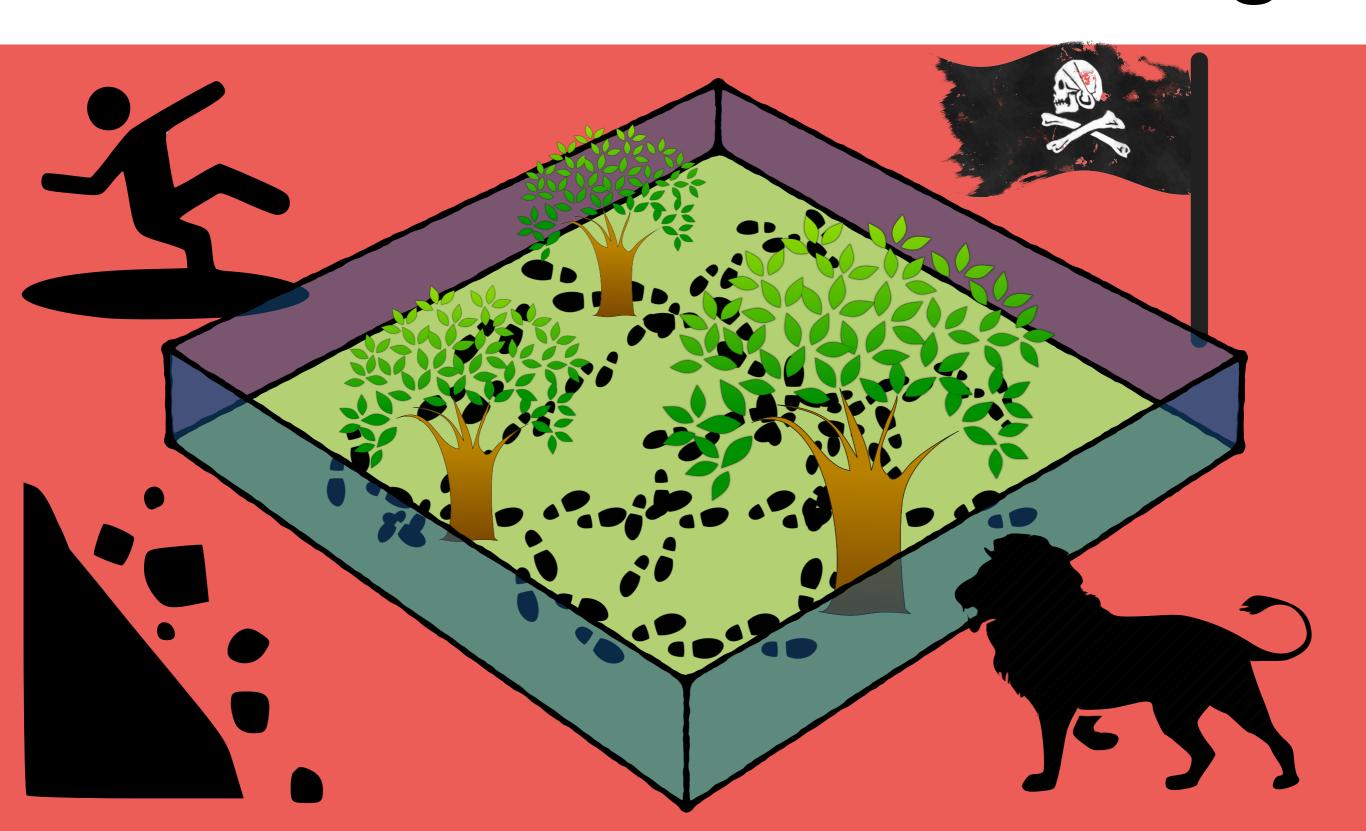
## Organizational boundaries



# Constraint-based modeling



#### Constraint-based modeling



#### Declare

- A constraint-based extensible language for flexible process modeling
- An execution engine for constraint-based processes
- Originally proposed by Pesic and van der Aalst
- Formalized by Pesic and myself (in our respective PhDs)

#### Constraints

- Much richer than the simple "precedence sequence flow connector" of imperative languages
- Supports:
  - Presence/absence of tasks (do/not do ... N times)
  - **Negative** relationships (it is forbidden to...)
  - Atemporal relationships (no order imposed)
  - Temporal relationships, with different strengths

# Global perspective

Constraints have to be globally satisfied along an entire process execution

 May stay quiescent or require events (not) to happen depending on the context

Example: if the conference format is changed, participants have to be notified

- The notification is not required to be immediate
- The format may change multiple times, or not change at all

# Composing constraints

Individual constraints are simply composed by conjunction (they must be all satisfied)

- Elegant compositional approach
- Creates interactions ("hidden dependencies") among constraints

Example: if you cancel the order, you cannot pay for it. If you confirm the order, you must pay for it.

—> It is not possible to confirm and cancel

# Flexible shopper

Pick item

**Close** order

**Pay** 

# Flexible shopper

Pick item



# Detour: are Petri nets declarative?

Pick item

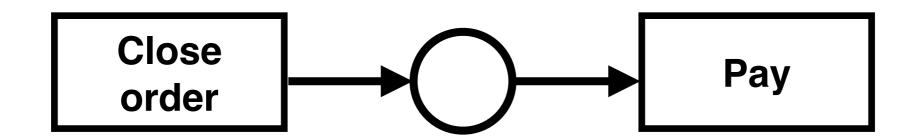
Close order

**Pay** 

A Petri net supporting any behavior on the given tasks!

# Detour: are Petri nets declarative?

Pick item

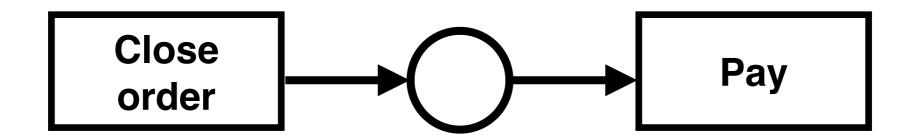


Adding places means adding temporal constraints...

But shall we interpret them backwards? Or forward?

# Detour: are Petri nets declarative?

Pick item



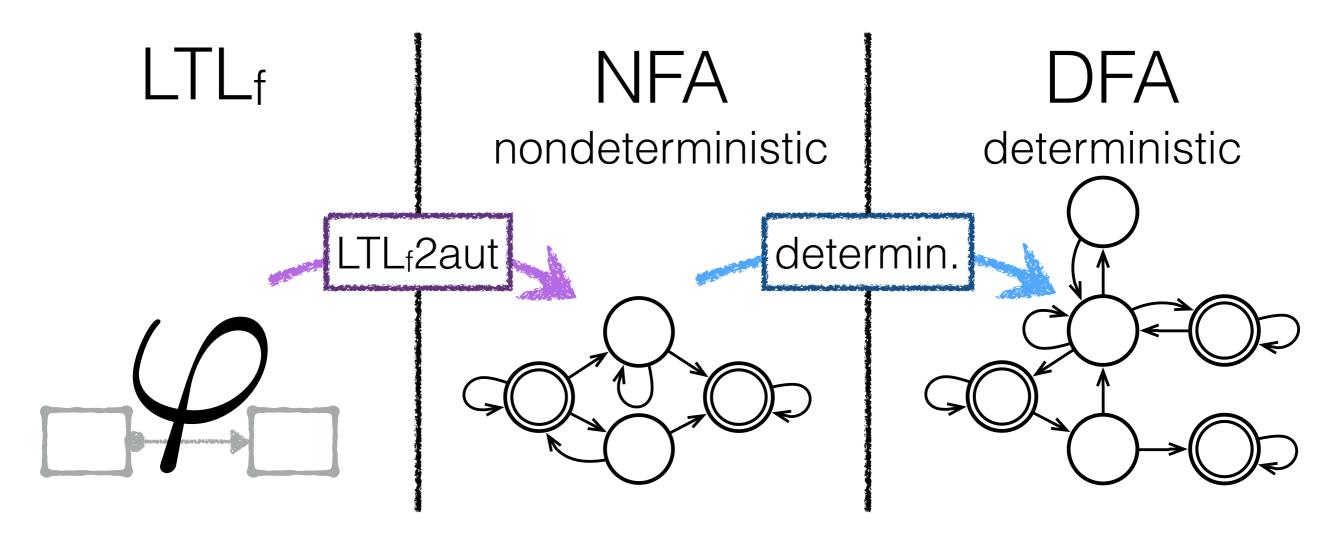
Atemporal and negative constraints cannot be added explicitly, nor incrementally.



# From Declare to logics to automata

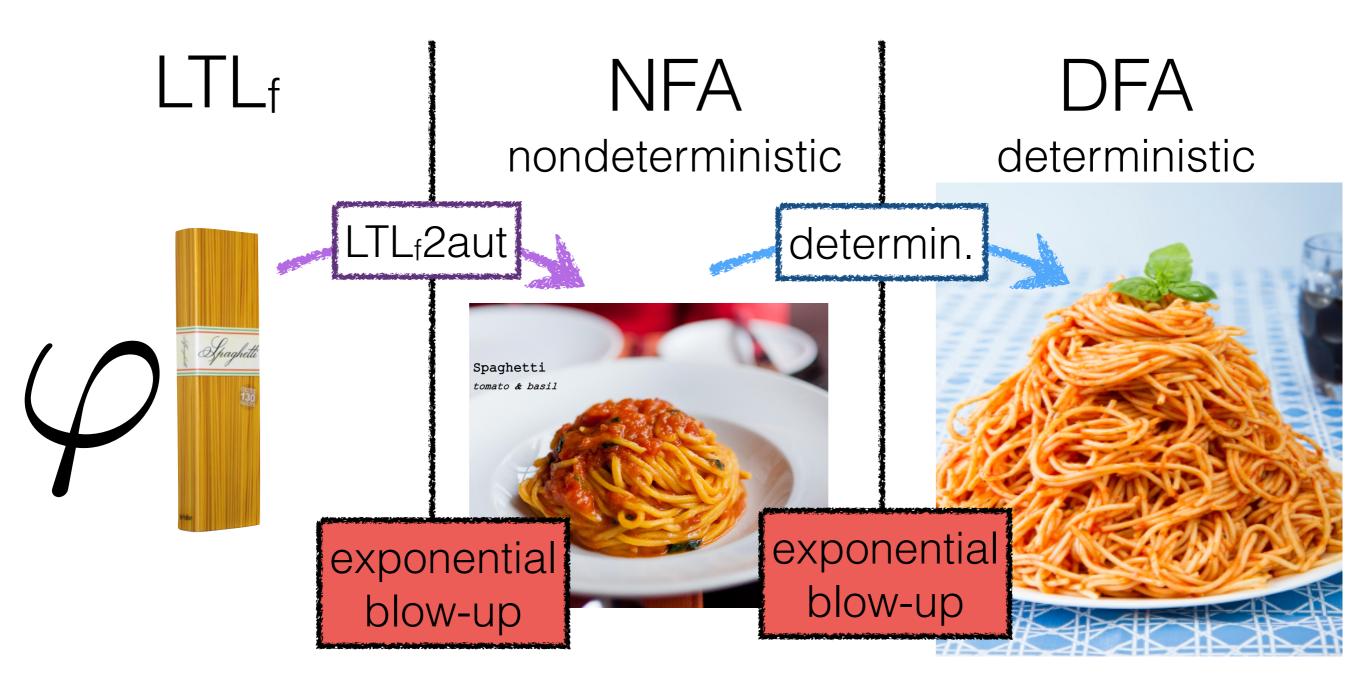
- Observation: Declare constraints are formalized using LTL over finite traces (LTL<sub>f</sub>)
- LTL<sub>f</sub> corresponds to the star-free fragment of regular expressions
- Intimate connection with finite-state automata

#### Declare -> automata

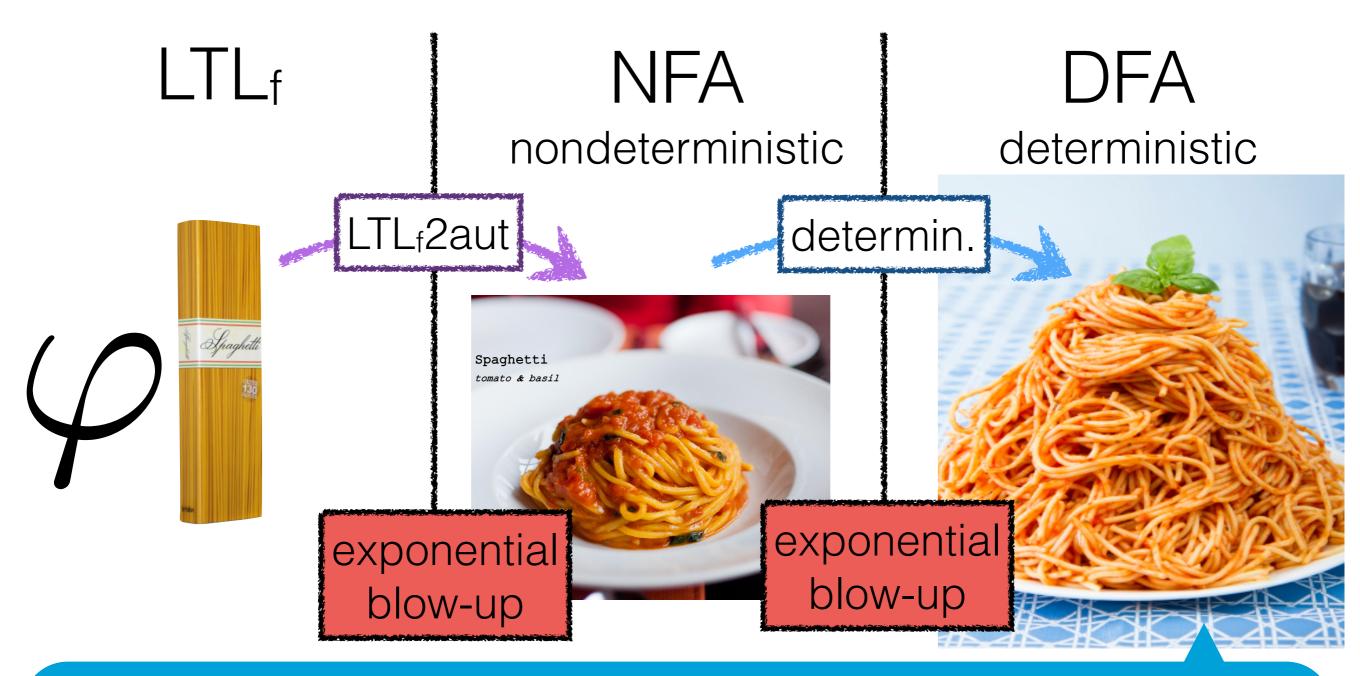


See keynote by Giuseppe De Giacomo (Tomorrow at BPM 2021)

## Our vision realized



## Our vision realized



Basis for reasoning, execution, process mining.

Built for each single constraints (fine-grained feedback) and for the entire model (constraint interplay).

# Some history: 3 waves

#### First wave

#### Declare, its formalization, verification, and execution

Constraint-Based
Workflow Management Systems:
Shifting Control to Users

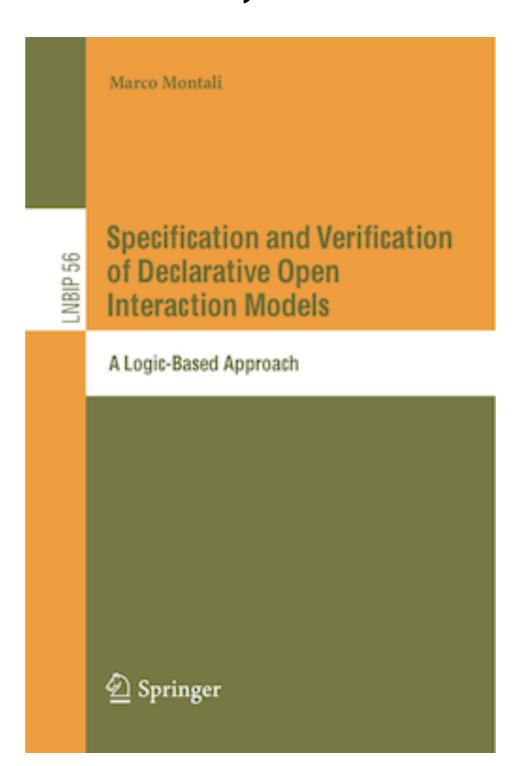
**PROEFSCHRIFT** 

ter verkrijging van de graad van doctor aan de Technische Universiteit Eindhoven, op gezag van de Rector Magnificus, prof.dr.ir. C.J. van Duijn, voor een commissie aangewezen door het College voor Promoties in het openbaar te verdedigen op woensdag 8 oktober 2008 om 16.00 uur

door

Maja Pešić

geboren te Belgrado, Servië



#### Second wave

#### Process discovery and operational support

- First work at BPM 2007 with related works in Leuven
- Claudio's PhD thesis and Fabrizio's works: more scalable algorithms
- Collaboration between Bologna, TU/e, Tartu on operational decision support (in particular, monitoring)

# **Second wave**Parallel line of research

- "Human factor" and understandability (Innsbruck, Copenhagen, St Gallen)
- Formal foundations using a variety of temporal logics over finite traces, with huge impact in AI (Rome, Rice, Bolzano) and BPM (Bolzano, Vienna)
- Model extensions and hybrid models (Tartu, Bolzano, Trento, Rome, Utrecht, Leuven, ...)
- Variety of groups investigating several facets of the topic

Parallel line of research on DCR graphs (Copenhagen)

#### Third wave: Rum...

... + multi-perspective models, real-life scaling

See closing part

#### Declarative Process Mining with RuM

#### Overview of RuM

#### RuM - Rule Mining Made Simple

- Based on the Declare language
- Incorporates various well-known algorithms
- Available at: <a href="https://rulemining.org/">https://rulemining.org/</a>
  - ~585 MB download
  - requires Java 11 JDK (available through the download link)



RuM will be used in the following sections

Please start the download now - We will provide assistance

Consider the goal to give a tutorial

Consider the goal to give a tutorial

- Consider the goal to give a tutorial
- The process starts by submitting a proposal

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init(submit proposal)

Init submit proposal

- Consider the goal to give a tutorial
- The process starts by submitting a proposal
- After submitting a proposal, a decision is received

init(submit proposal)

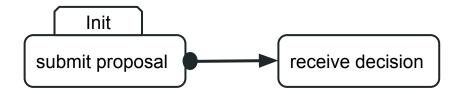
Init submit proposal

### Declare Example

- Consider the goal to give a tutorial
- The process starts by submitting a proposal
- After submitting a proposal, a decision is received

init(submit proposal)

response(submit proposal, receive decision)



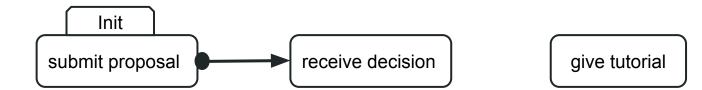
give tutorial

### Declare Example

- Consider the goal to give a tutorial
- The process starts by submitting a proposal
- After submitting a proposal, a decision is received
- The tutorial can be given if the decision is positive

init(submit proposal)

response(submit proposal, receive decision)



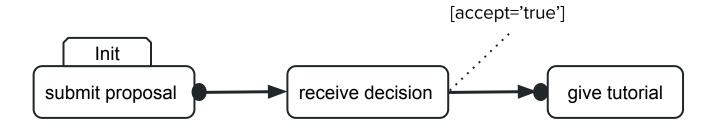
### Declare Example

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init(submit proposal)

response(submit proposal, receive decision)

precedence(give tutorial, receive decision [accept='true'])



## Hands-on session with RuM

### Scenario

### Overarching goal:

Create a synthetic event log that mimics behavior found in a real event log

### Inputs:

- 2 event logs of sepsis cases
  - Training set and test set
  - o 391 traces each (50/50 split)
  - Completed cases only

### Overarching goal

Create a synthetic event log that mimics behavior found in a real event log

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Create a synthetic event log that mimics behavior found in a real event log

### Main steps

1. Discover a model



### Overarching goal

Create a synthetic event log that mimics behavior found in a real event log

### Main steps



### Overarching goal

Create a synthetic event log that mimics behavior found in a real event log

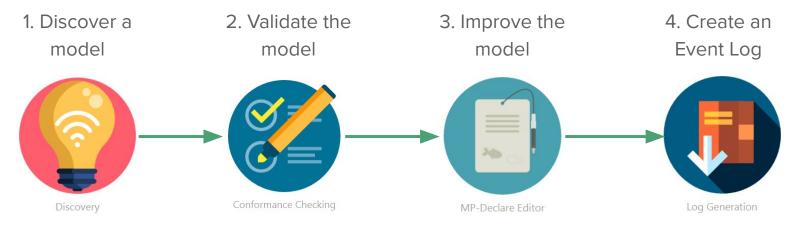
### Main steps



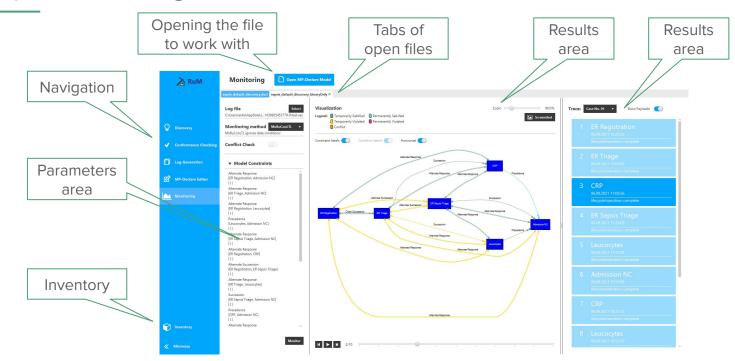
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### Main steps

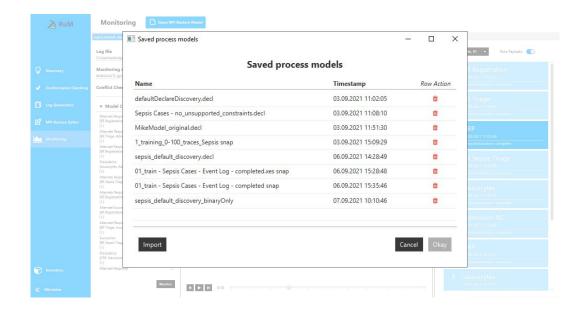


## Tips for Using RuM



## The Inventory System

- For fast access to models and event logs
  - Reuse of models and event logs across RuM
- Allows for snapshots
  - A temporary model or event log for analysis



### Task 1: Discover a Model



#### Inputs:

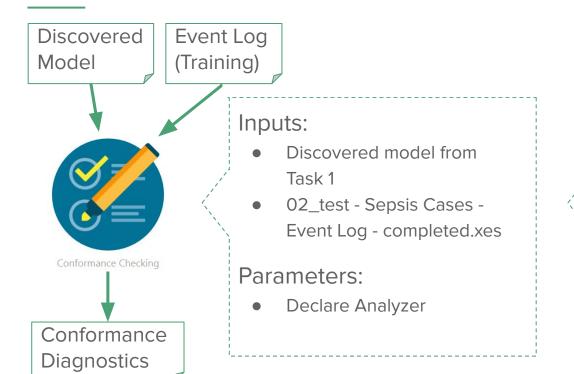
01\_train - Sepsis Cases Event Log - completed.xes

#### Parameters:

- Declare Miner
- Support: 90%
- All reductions
- Unary templates
- Positive Binary Templates
- Activity support: 95%

- Template descriptions can be found in the templates panel
- The textual view can be helpful for understanding the model
- Activity support can be changed after the discovery
- Activity support can be written with the keyboard

### Task 2: Validate the Model



- Results are displayed in groups
- Each group can be opened
- Each line in a group can be clicked
- Results can be explored by traces and by constraints
- Results can be sorted.

### Task 3: Improve the Model



#### Inputs:

Discovered model from Task 1

#### What to modify:

- Remove the constraints that caused 50 or more violations
  - (based on the Task 2)

- You can switch back and forth between the different views of RuM
- Constraints can be sorted by clicking on the column title
- Constraints can be modified using the 'Row actions' column

### Task 4: Create an Event Log



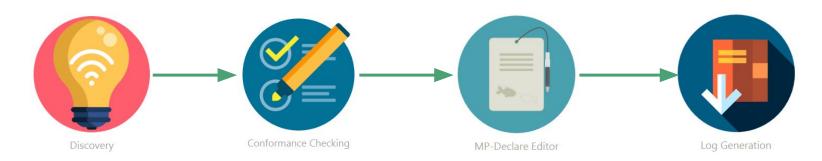
#### Inputs:

Improved model from Task 3

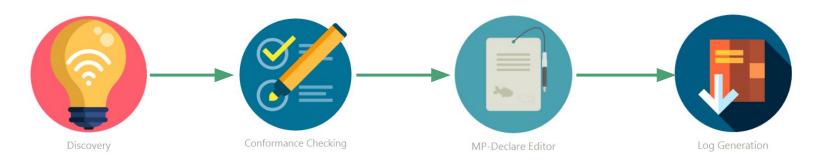
#### Parameters:

- AlloyLogGenerator
- Minimum events: 5
- Maximum events: 10
- Positive traces: 50
- Positive vacuous: 0%
- Negative traces: 0
- Negative vacuous: 0%

- The created event log can be seen on the left
- It is possible to explore each trace individually
- It is possible to explore the data payloads



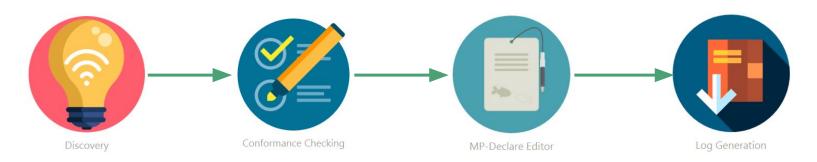
What more could have been done?



#### What more could have been done?

#### Discover a model

- MINERful
- Various templates
- Support thresholds
- Data conditions



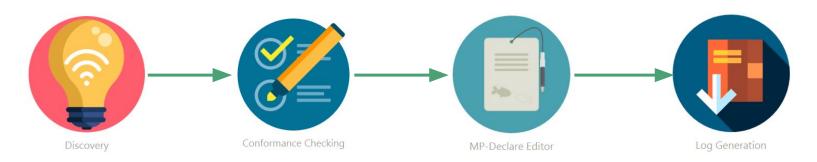
#### What more could have been done?

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#### Validate the model

- Alignments
- Monitoring
- Export fulfilled
- Assess statistics



#### What more could have been done?

#### Discover a model

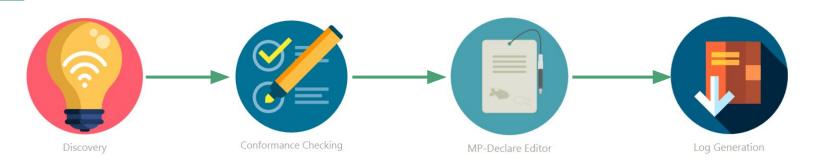
- MINERful
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#### Validate the model

- Alignments
- Monitoring
- Export fulfilled
- Assess statistics

#### Improve the model

- Adding constraints
- Adding activities
- Data conditions
- Textual input
- Voice input



#### What more could have been done?

#### Discover a model

- MINERful
- Various templates
- Support thresholds
- Data conditions
- Time conditions

#### Validate the model

- Alignments
- Monitoring
- Export subsets
- Assess statistics

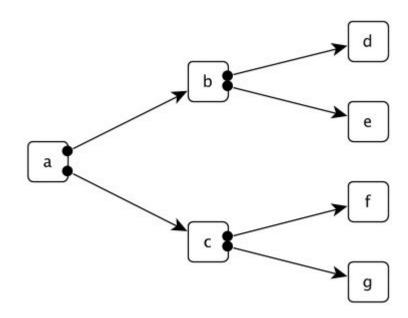
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- Adding constraints
- Adding activities
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- Textual input
- Voice input

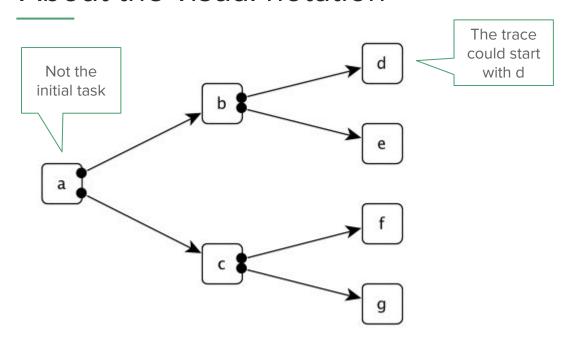
#### Create an event log

- MINERful Log
   Generator
- Negative traces
- Vacuous traces
- Data payloads

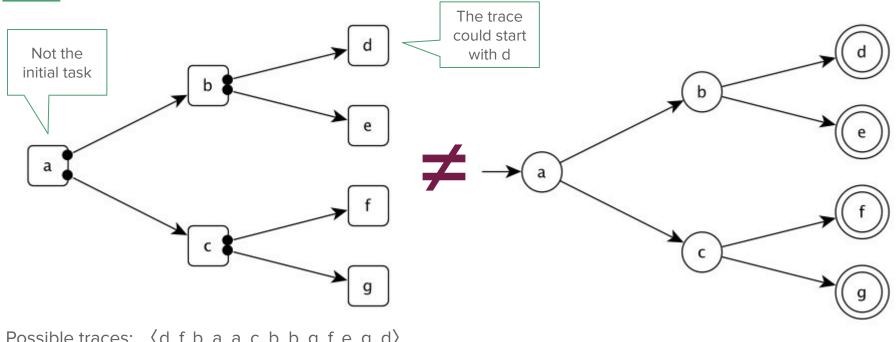
# Some Research Opportunities



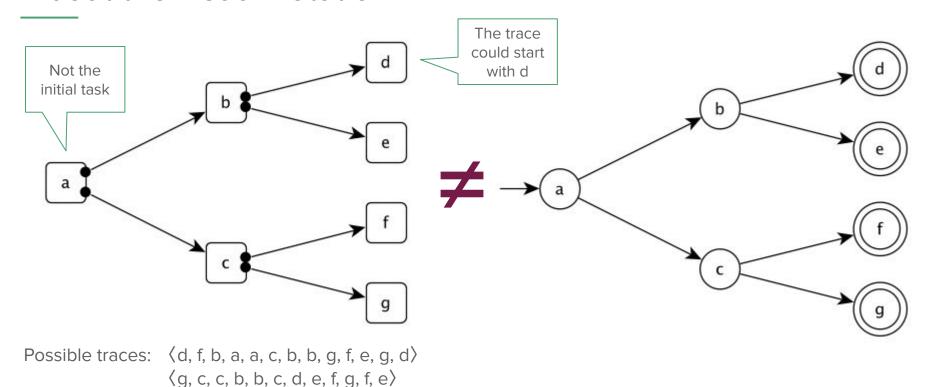
Possible traces:  $\langle d, f, b, a, a, c, b, b, g, f, e, g, d \rangle$   $\langle g, c, c, b, b, c, d, e, f, g, f, e \rangle$ 



Possible traces:  $\langle d, f, b, a, a, c, b, b, g, f, e, g, d \rangle$   $\langle g, c, c, b, b, c, d, e, f, g, f, e \rangle$ 



Possible traces:  $\langle d, f, b, a, a, c, b, b, g, f, e, g, d \rangle$   $\langle g, c, c, b, b, c, d, e, f, g, f, e \rangle$ 



BPM 2021 | September 8th, 2021

## Hybrid (mixed-paradigm) models

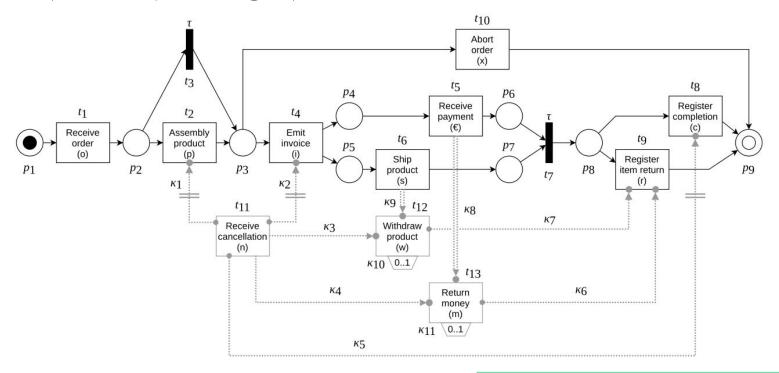
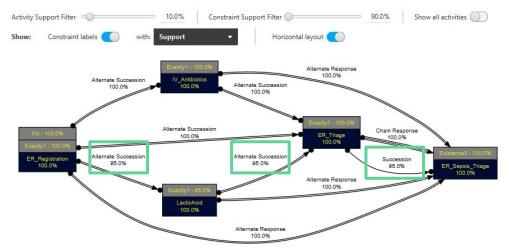


Image from: van Dongen et al. 2021, https://doi.org/10.1016/j.is.2020.101685

## Crisp and probabilistic constraints

 Measures of mined constraints can lie beneath 100%...

 ... but if they lie above the threshold, they are treated as if they were undeniable



## Crisp and probabilistic and optional constraints

 Measures of mined constraints can lie beneath 100%...

• ... but if they lie above the threshold, they are treated as if they were undeniable

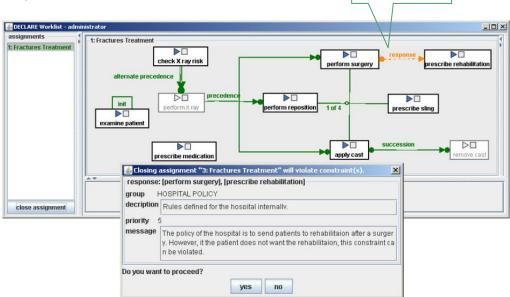


Image from: Pesic 2008, <a href="https://doi.org/10.6100/IR638413">https://doi.org/10.6100/IR638413</a>

Optional

constraint

## Object-centric processes

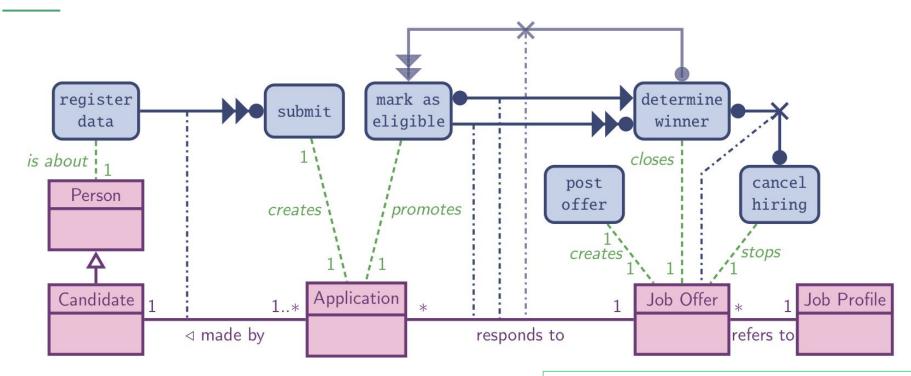


Image from: Artale et al 2019, https://doi.org/10.3233/978-1-61499-955-3-257

## Your ideas?



# RuM: Declarative Process Mining, Distilled

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https://rulemining.org/