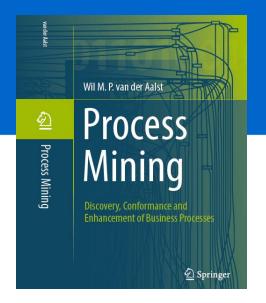
## Bits&Chips 2012 Embedded Systems Congrescentrum Brabanthallen, 's-Hertogenbosch, 8-11-2012

# Process Mining How are my systems used and when do they fail?

prof.dr.ir. Wil van der Aalst www.processmining.org



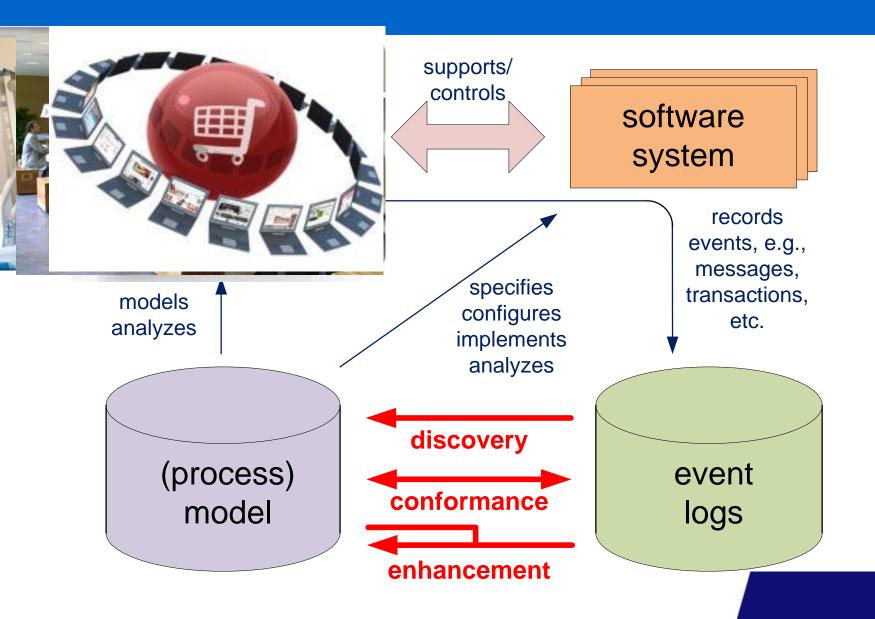


Technische Universiteit
Eindhoven
University of Technology

Where innovation starts



### **Process Mining**



## **Desire Lines in Big Data**

## **Big Data**

"All of the world's music can be stored on a \$600 disk drive."

0101001100

"Enterprises globally stored more than 7 exabytes of new data on disk drives in 2010, while consumers stored more than 6 exabytes of new data on devices such as PCs and notebooks."

"Indeed, we are generating so much data today that it is physically impossible to store it all. Health care providers, for instance, discard 90 percent of the data that they generate."

Source: "Big Data: The Next Frontier for Innovation, Competition, and Productivity" McKinsey Global Institute, 2011.

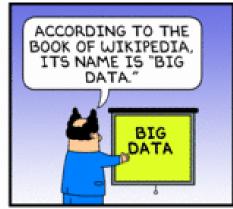
## What Happens in an Internet Minute?



## Big Data: Even Dilbert and the "pointy-haired boss" know about it ...





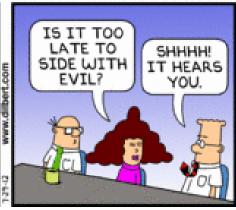


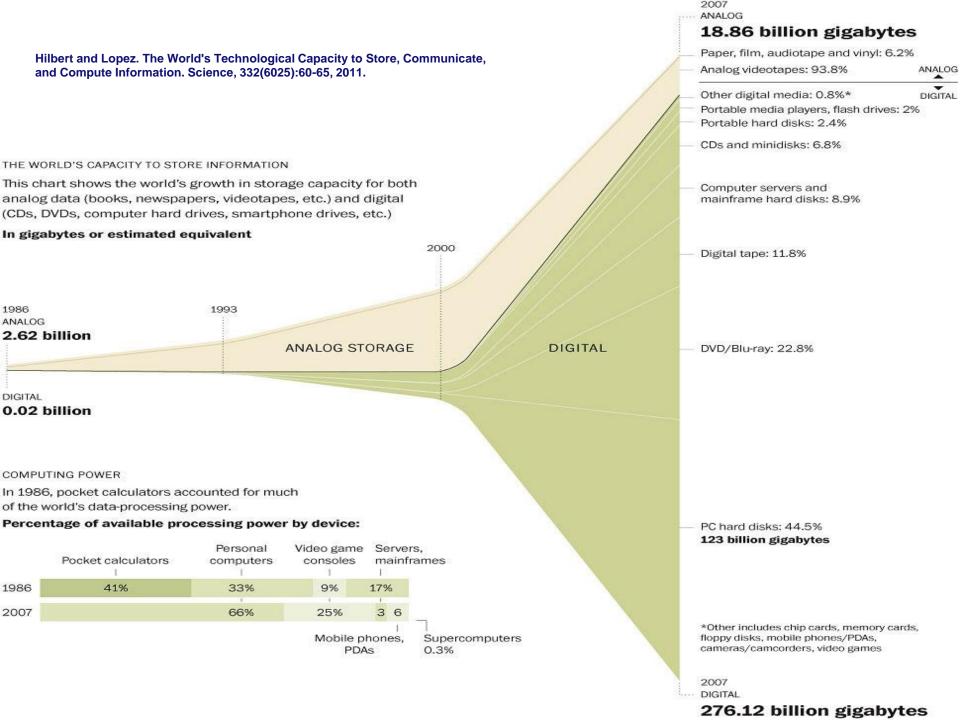










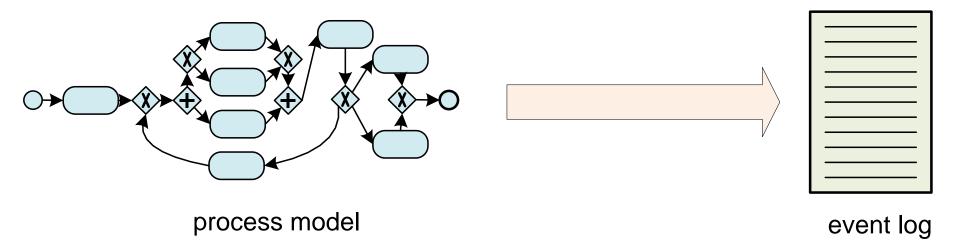




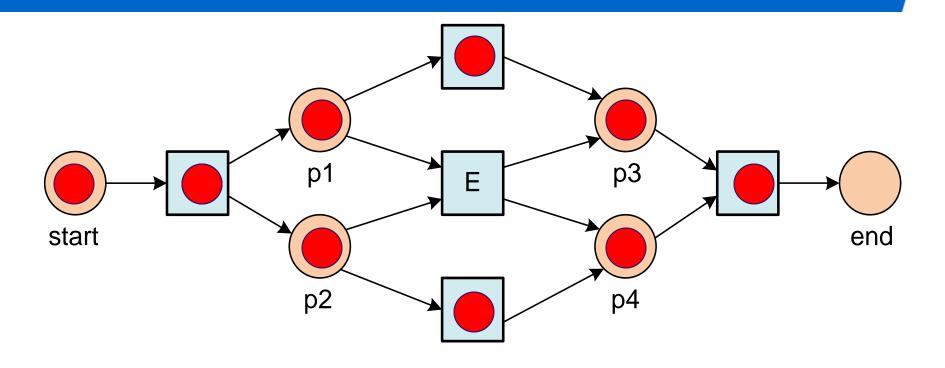


# On the different roles of (process) models ...

## **Play-Out**

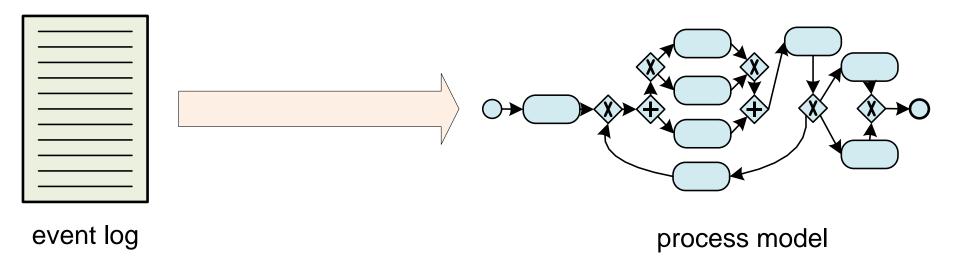


### Play-Out (Classical use of models)



ABCDAED AED ACBD ABCD ACBD

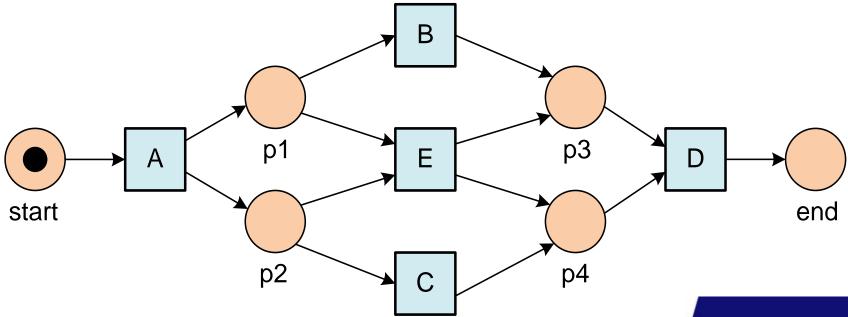
## Play-In



### Play-In

ABCD AED AED

ACBD ACBD

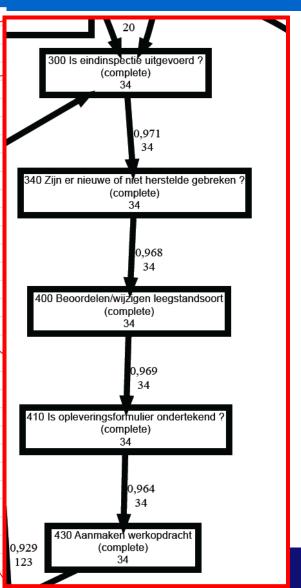


### **Example Process Discovery**

(Vestia, Dutch housing agency, 208 cases, 5987 events)

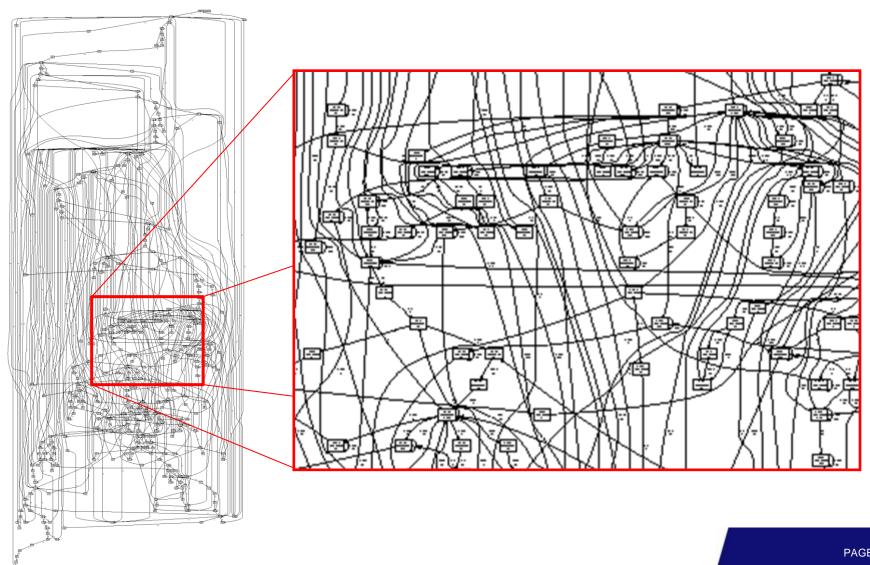






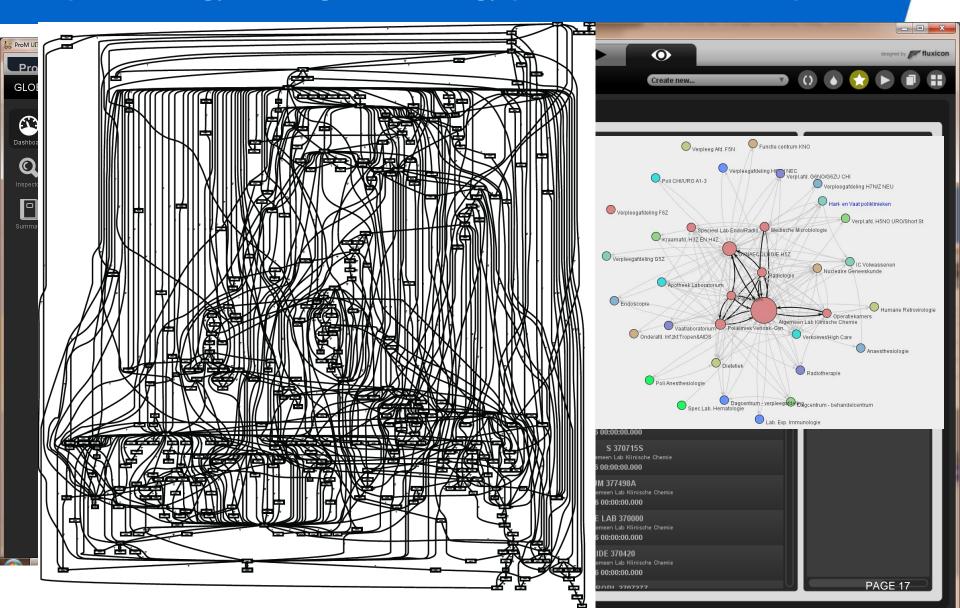
## **Example Process Discovery**

(ASML, test process lithography systems, 154966 events)

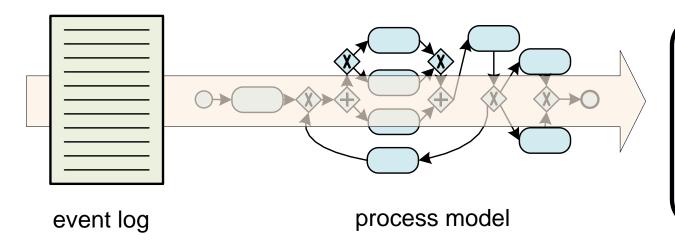


### **Example Process Discovery**

(AMC, 627 gynecological oncology patients, 24331 events)

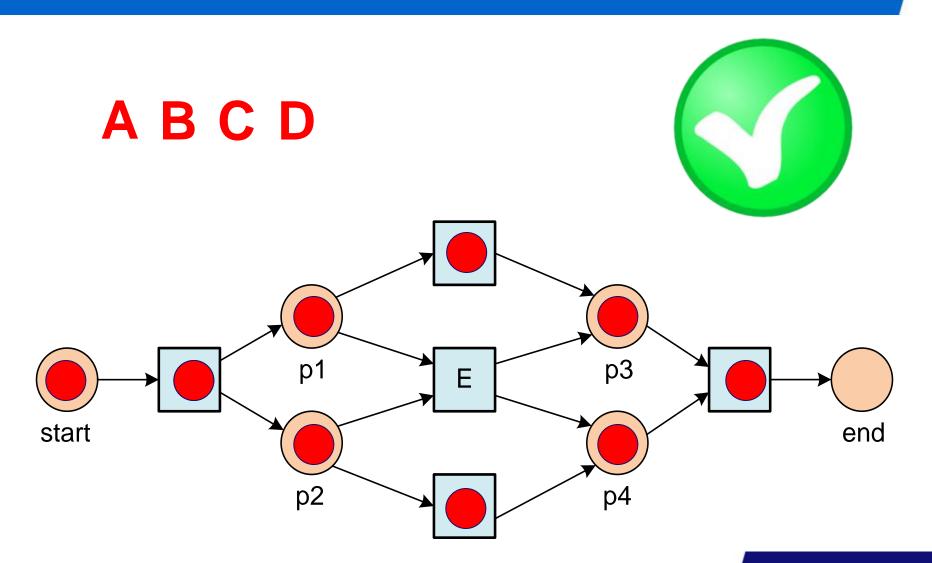


### Replay

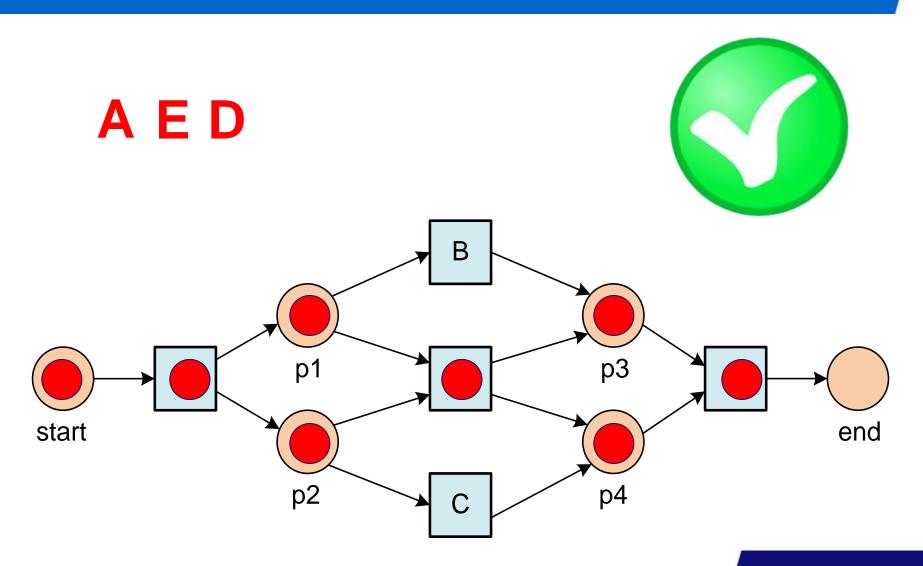


- extended model showing times, frequencies, etc.
- diagnostics
- predictions
- recommendations

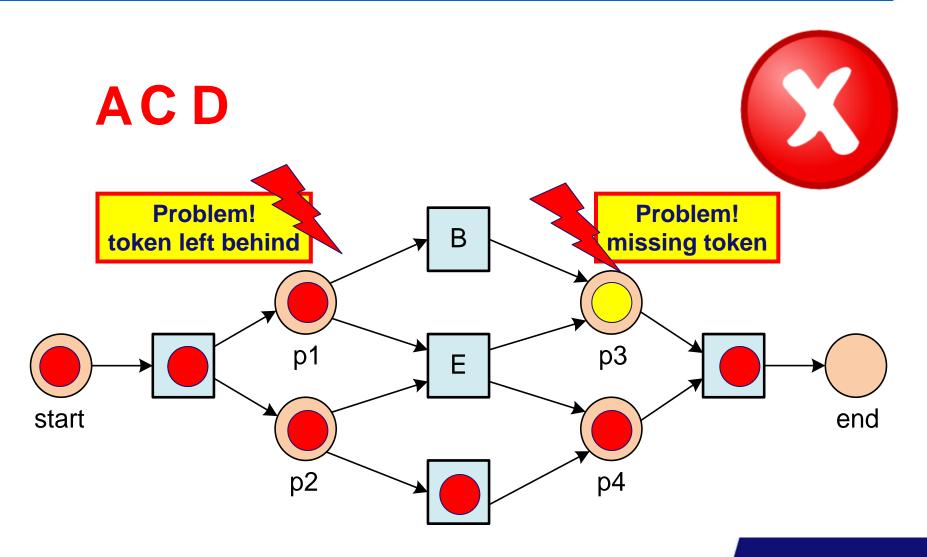
## Replay

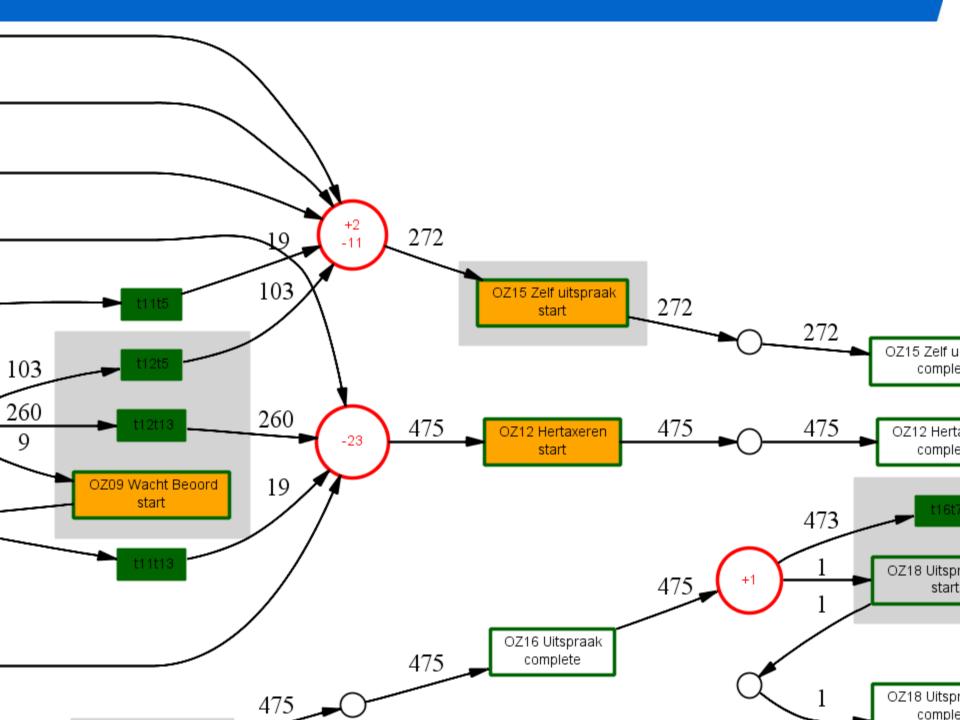


## Replay

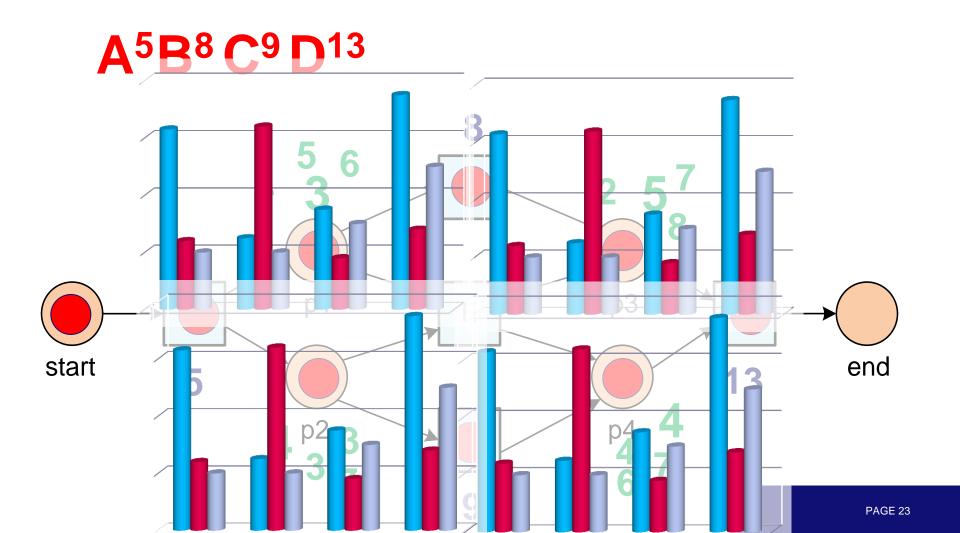


### Replay can detect problems



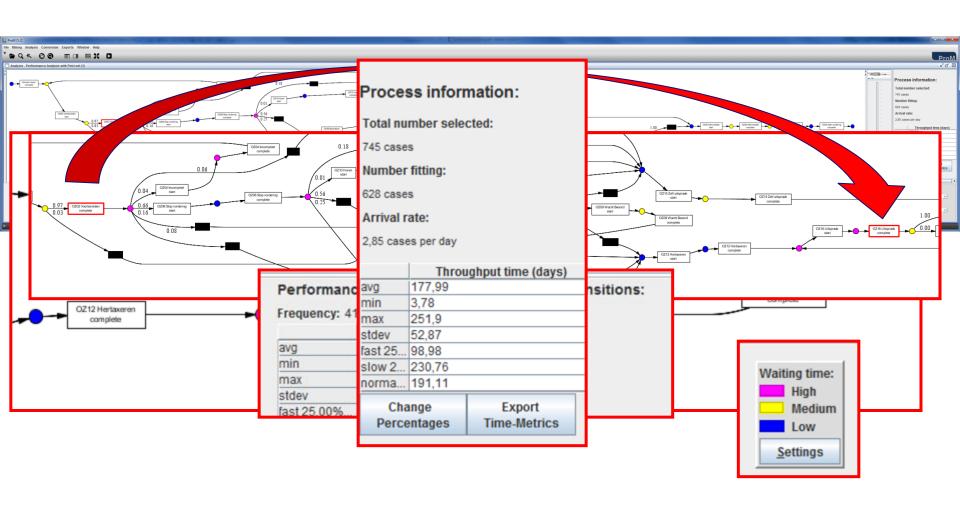


### Replay can extract timing information



### **Performance Analysis Using Replay**

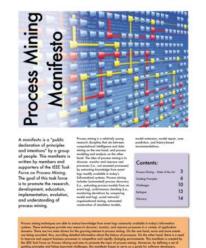
(WOZ objections Dutch municipality, 745 objections, 9583 event, f= 0.988)



## **Process Mining**

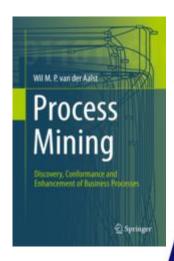
### **Advances in Process Mining**

- Many process discovery and conformance checking algorithms and tools are available (cf. the various ProM packages).
- Also commercial software based on these ideas:
   Disco (Fluxicon), Reflect (Futura/Percentive), BPMOne (Pallas Athena/Perceptive), ARIS Process Performance Manager (Software AG), Futura Reflect (Futura Technology), Interstage Automated Process Discovery (Fujitsu), QPR ProcessAnalyzer/Analysis (QPR Software), flow (fourspark), Discovery Analyst (StereoLOGIC), etc.
- We applied process mining in over 100 organizations.

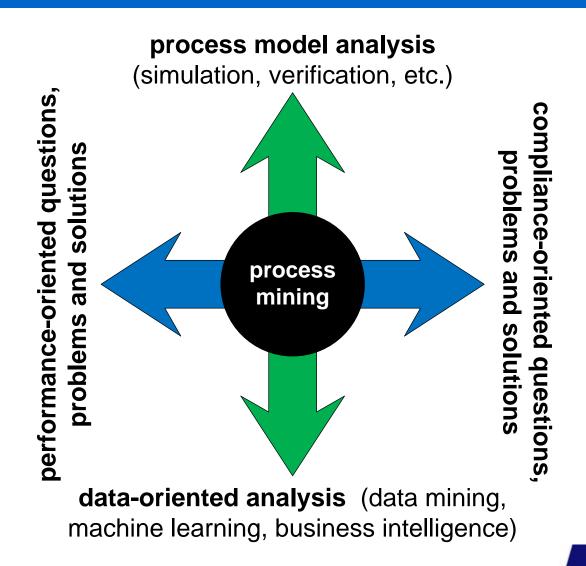


More than 75 people involving more than 50 organizations created the Process Mining Manifesto in the context of the IEEE Task Force on Process Mining.

Available in 13 languages



### Missing link



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### **Process Discovery (small selection)**

automata-based learning

heuristic mining

genetic mining

stochastic task graphs

fuzzy mining

mining block structures

α algorithm

α# algorithm

The second of th

distributed genetic mining

language-based regions

state-based regions

LTL mining

neural networks

hidden Markov models

multi-phase mining

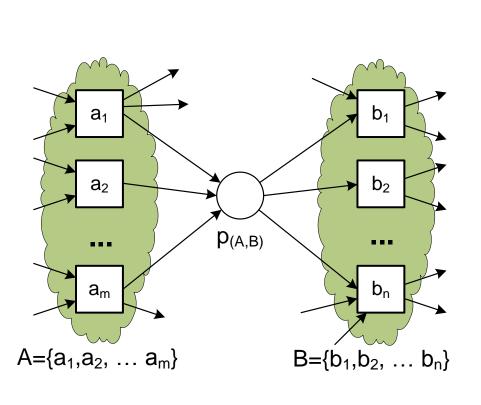
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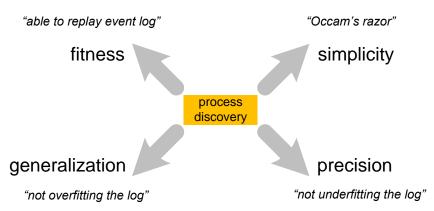
conformal process graph
partial-order based mining

**ILP** mining

α++ algorithm

# Petri net view: Just discover the places ...

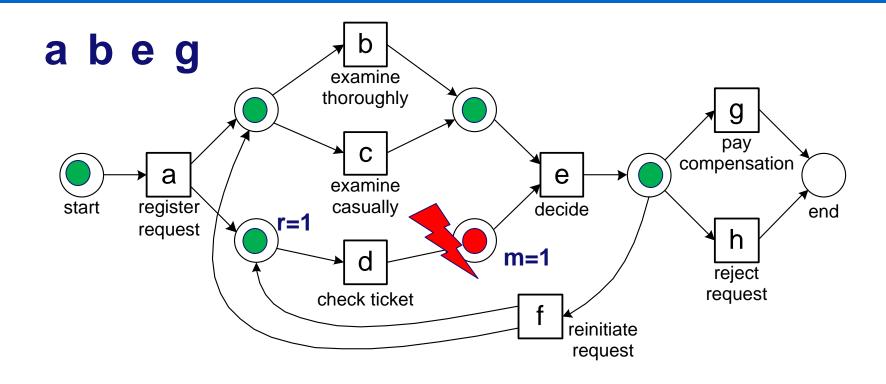




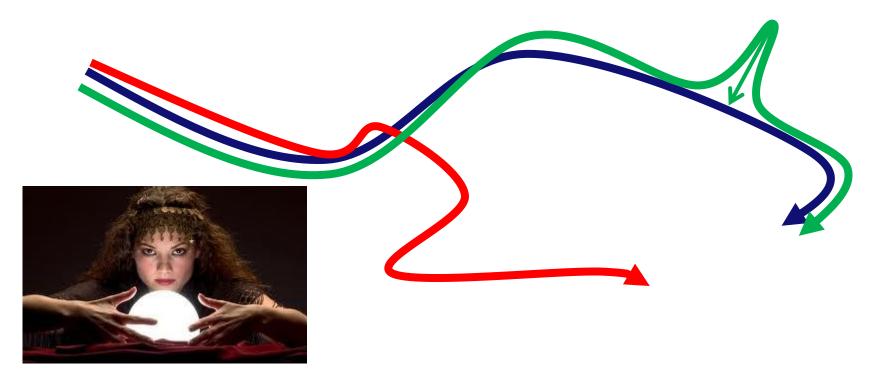
#### Adding a place limits behavior:

- overfitting ≈ adding too many places
- underfitting ≈ adding too few places

### Replaying trace "abeg"



$$fitness(\sigma, N) = \frac{1}{2} \left( 1 - \frac{1}{6} \right) + \frac{1}{2} \left( 1 - \frac{1}{6} \right) = 0.83333$$

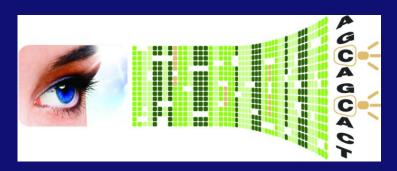


### Alignments are essential!

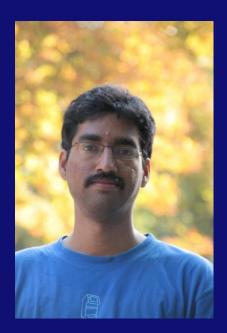
- conformance checking to diagnose deviations
- squeezing reality into the model to do model-based analysis
- not just control-flow (also data, resources, time, ...)!
- also at run-time (predictions and recommendations)!

## **Example Philips Healthcare**

(work of JC Bose)

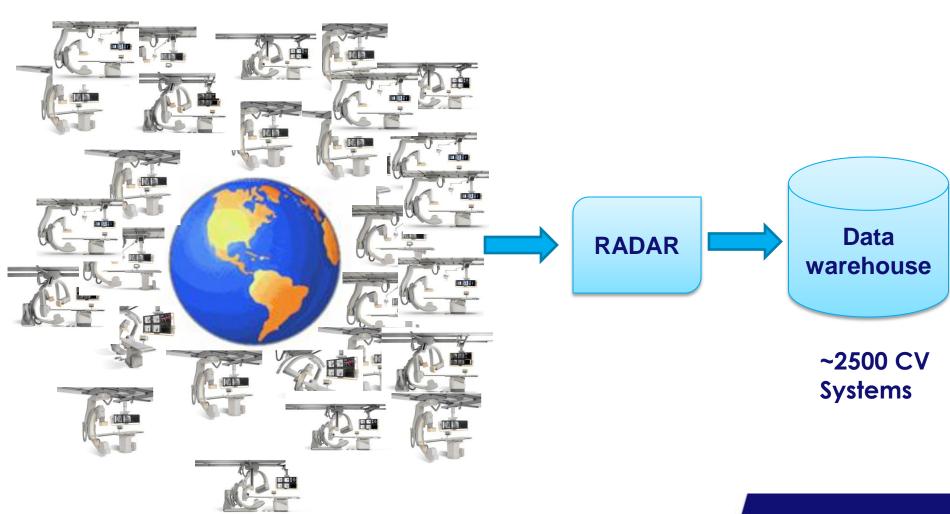






http://www.win.tue.nl/~jcbose/

### **Data Collection Process**



### What is Being Logged?

- Errors
- Warnings
- Information
- User commands
- Run information





### **Challenges**

### Event Granularity

- Fine grained and too detailed
- Often differs from the granularity at which an end-user/analyst is interested in
- Lack of any standard/guideline on logging specifications

### Case Heterogeneity

- Heterogeneous mix of usage scenarios
- Wide range of functionality and flexibility in usage

#### Voluminous Data

- Hundreds of events can be triggered within a short span of time (even within a second)
- Challenges the availability and scalability of analysis techniques

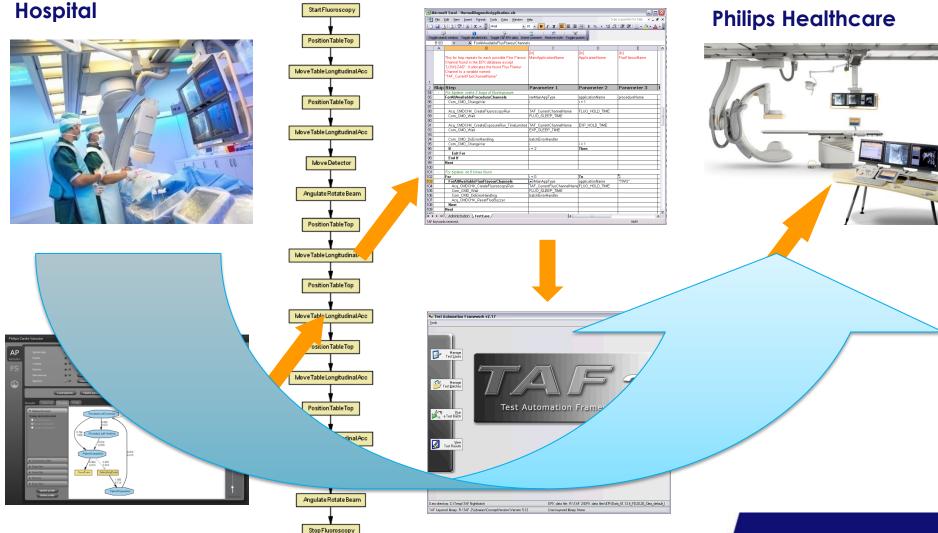
### Unreliable Timestamps

- Ordering of events is not always reliable
- An outcome of the current logging mechanism
  - Local clocks and local buffers
- Clocks are not synchronized

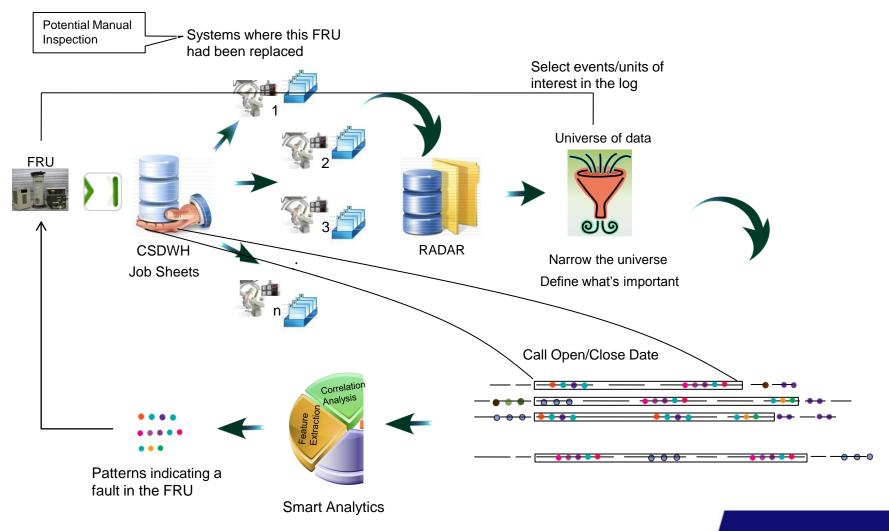
### Scoping

- Single log file captures everything related to a particular system on a single day
- Appropriate scoping is essential based on the perspective of analysis

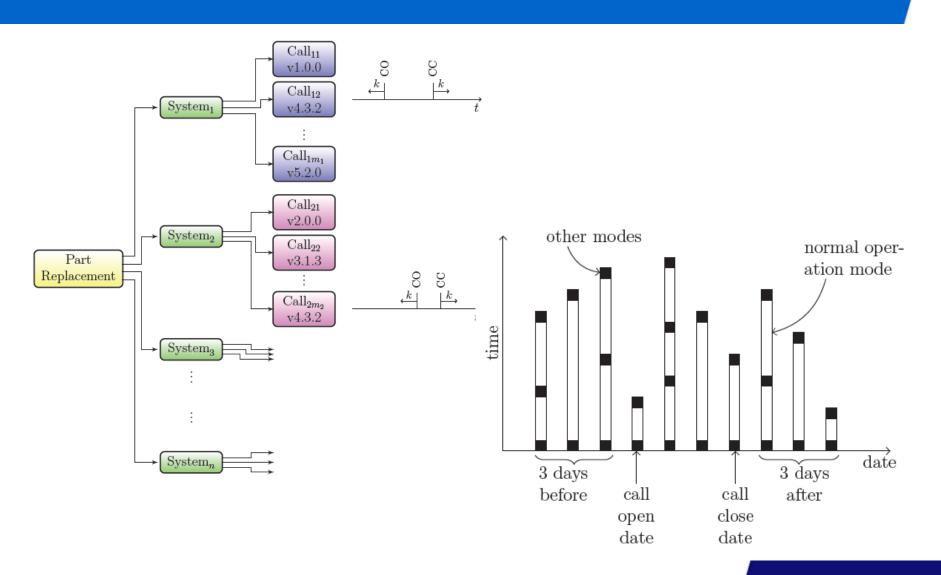
## Test Automation Framework: Testing CV systems under realistic circumstances



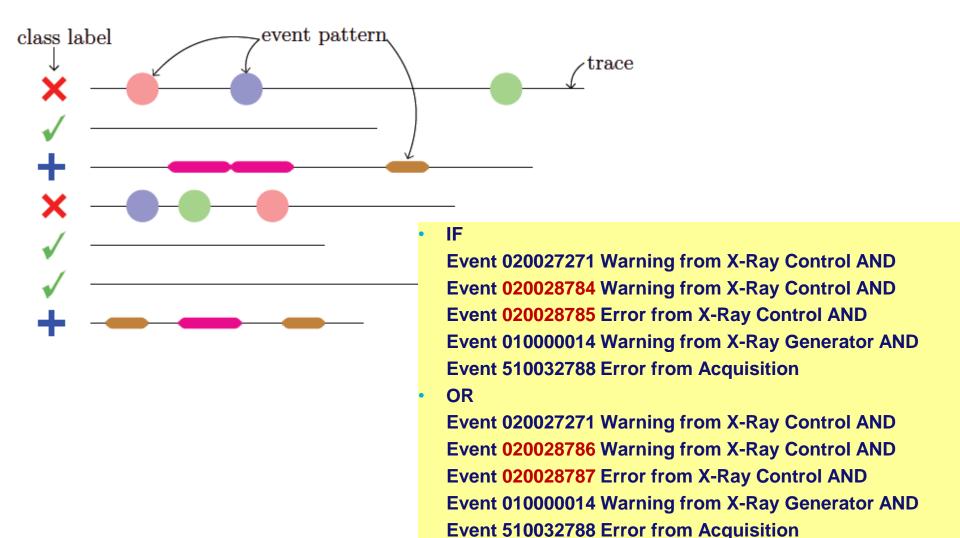
### **Signature Discovery**



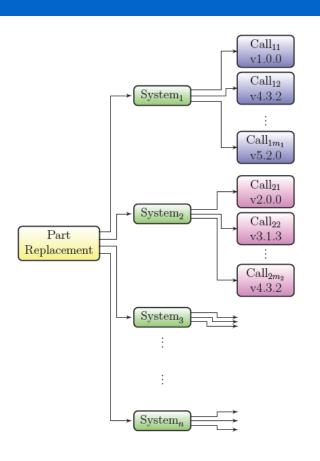
### **Building a classification problem**

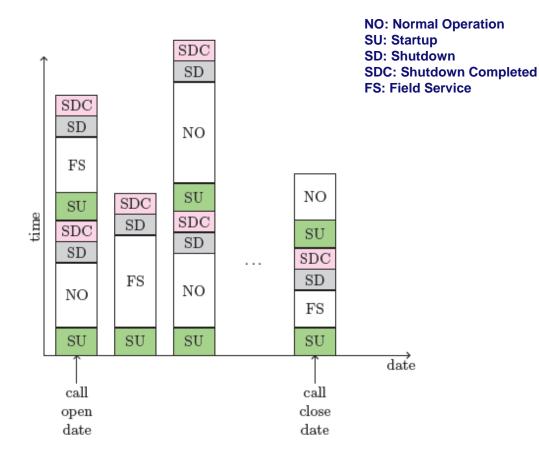


### **Signature Discovery**



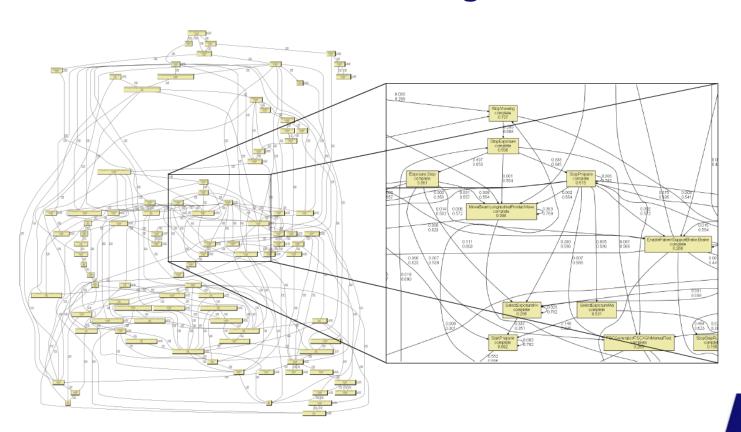
### Field Service Engineer Workflow Analysis



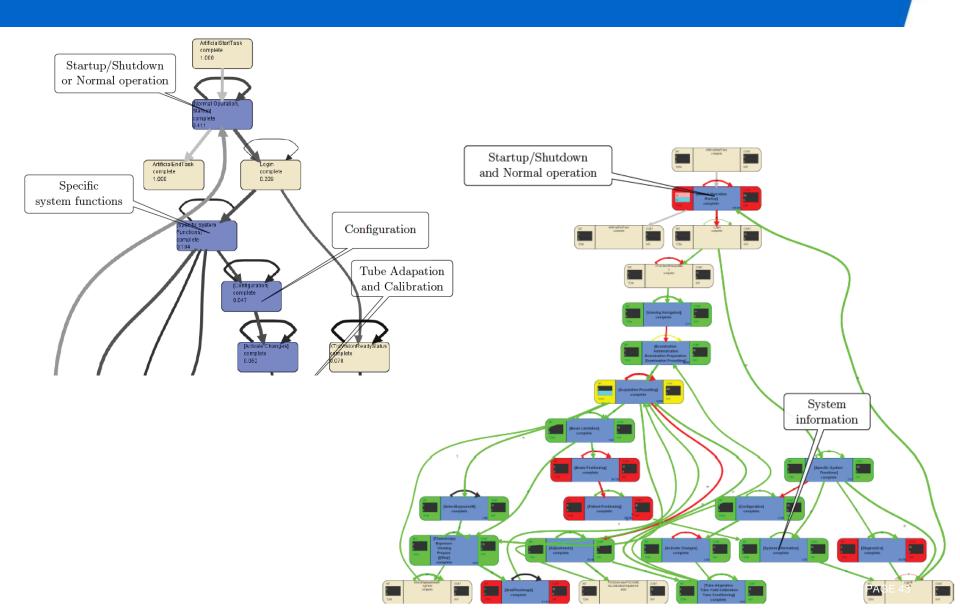


# Conformance Checking and Discovering the Actual Workflow

- Many deviations between instructions and real work of engineers.
- Event data are not at the right abstraction level.



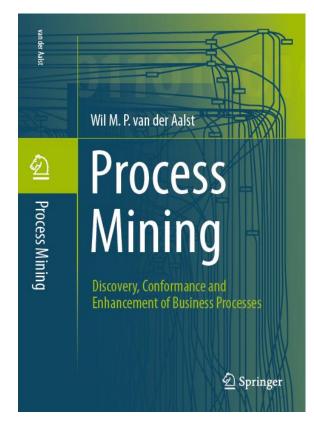
### Using abstractions and time information



## Conclusion

### Conclusion

- Process Mining (PM): connecting event data and process models
- PM shows how systems are <u>really</u> being used.
- PM can be used for <u>conformance</u> <u>checking</u> and <u>discovery</u>.
- There are mature PM tools (e.g., ProM, Disco, Reflect).
- If there is <u>enough high-quality</u> data, PM can also be used for prediction and recommendation.
- Interested: Contact us!



www.processmining.org