

Advanced Process Mining

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Lecture 6: Predictive Process Monitoring



Lecture Overview



0	Organization and	Introduction
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- I Process Discovery
- II Process Conformance
- III Predictive Process Monitoring
- IV Event Log Preparation
- V Practical Tasks

Business Process Monitoring



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Definition of Business Process Monitoring

- Business Process Monitoring is the process-oriented monitoring of the most important or critical business processes of a company
- It includes the observation of all technical and business application-specific functions that are required for a smooth and reliable flow of the business processes

Goals of Business Process Monitoring

- To detect critical situations as early as possible
- To enable the customer's solution support organization to respond to and to solve problems as fast as possible

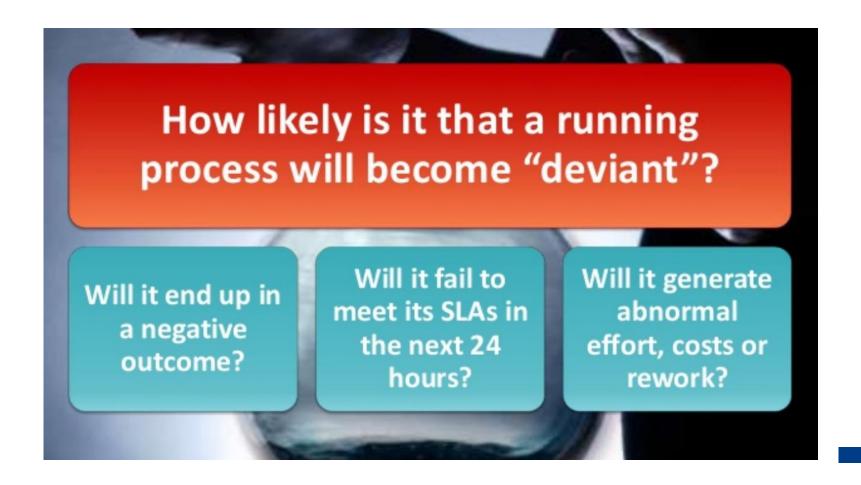
Monitoring types



- Proactive Monitoring: Activities are performed to prevent critical situations before they occur
- Reactive Monitoring: Once an error has occurred and detected, additional monitoring activities are required to analyse and solve the problem
- Automatic Monitoring: An automatic monitor triggers an alert immediately after an exceptional situation has occurred

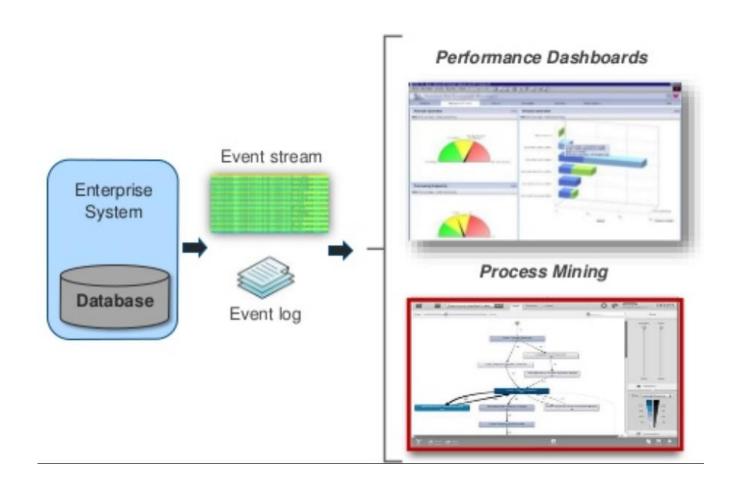
Business Process Monitoring





Business Process Monitoring

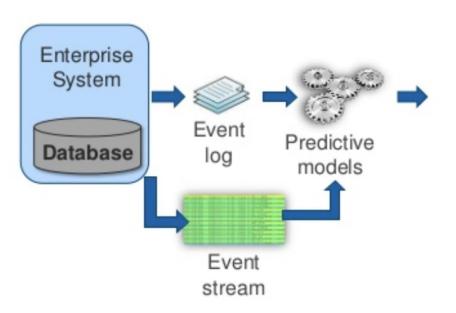




Predictive Process Monitoring



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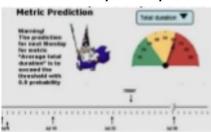
Aggregate predictive dashboards



Detailed predictive dashboard

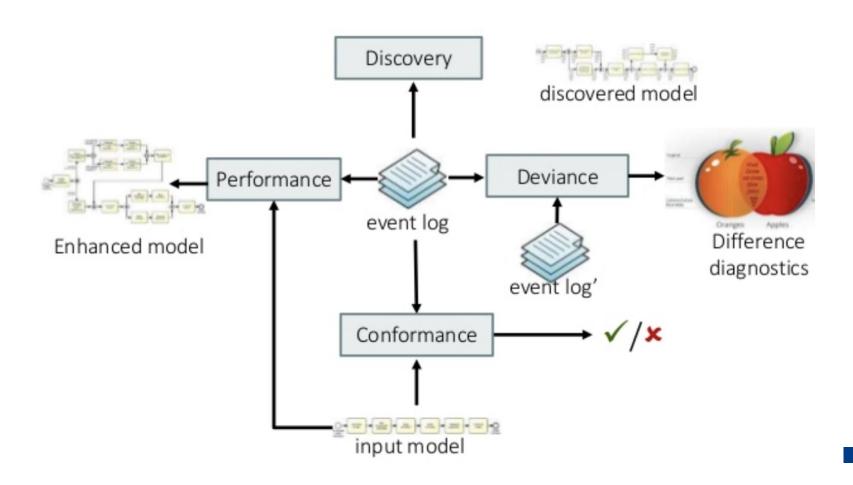


Alarm -based prescriptive dashboard



Offline Process Mining





Deviation Mining via Sequence Classification

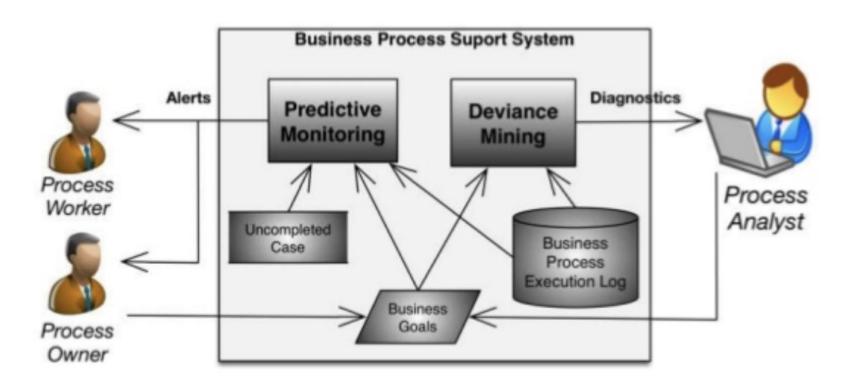


- Apply discriminative sequence mining methods to extract features characteristic of one class
- Build classification models (e.g. decision trees)
- Extract difference diagnostics from classification model

Rank	Rule
1	{(Open, 2}:anomalous
2	{(Closed, 2), (Postponed, 0), (Finished, 0)}:anomalous
3	{(Reopen, 2)}:anomalous
4	{(Closed, 1), (Rejected, 1), (Reopen, 0)}:anomalous
5	{(Reopen, Closed, 1)}:anomalous

Deviance Mining and Predictive Monitoring





Classical Setting for Business Process Monitoring(1)





Classical Setting for Business Process Monitoring(2)





We can tell at every moment:

- where each package is in the process,
- what is in each trolley,
- who is doing what

Classical Setting for Business Process Monitoring(3)



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We can apply process mining to

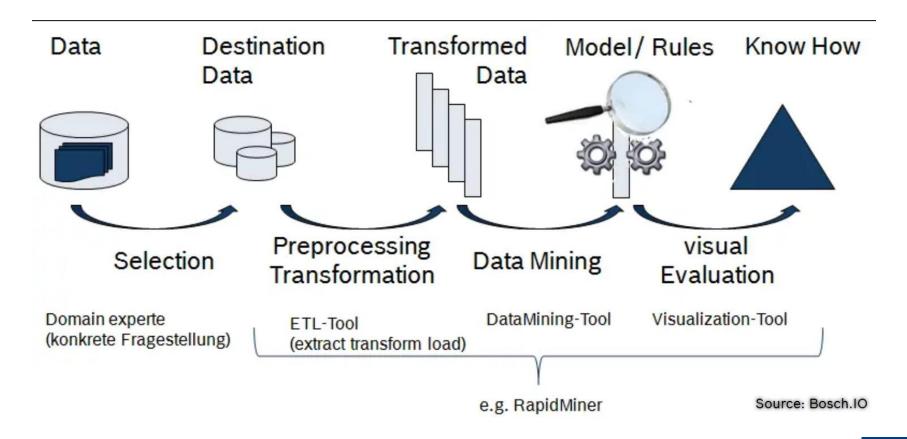
- understand the process,
- check compliance,
- generate predictions,
- ...

We can tell at every moment:

- where each package is in the process,
- what is in each trolley,
- who is doing what

Data Analysis and Predictive Monitoring





Quelle: https://blog.bosch-si.com/developer/large-scale-data-analysis-and-predictive-modeling-in-data-mining/

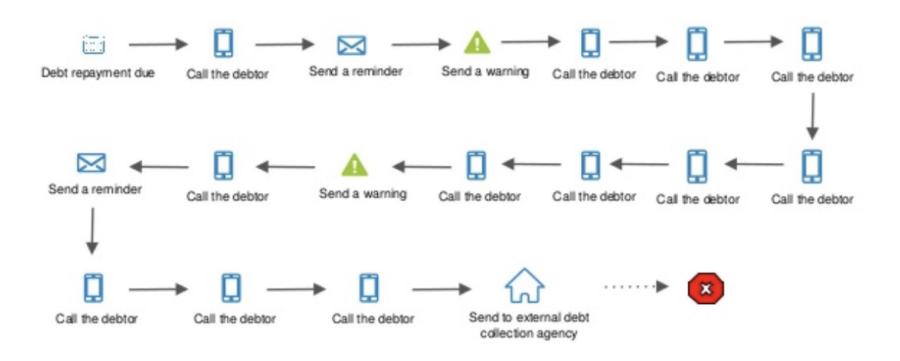
Predictive Monitoring Example: CIAIU Debt Recovery Process (1)



Send a reminder Payment received Debt repayment due Call the debtor

Predictive Monitoring Example: clalu Debt Recovery Process (2)





Time is important

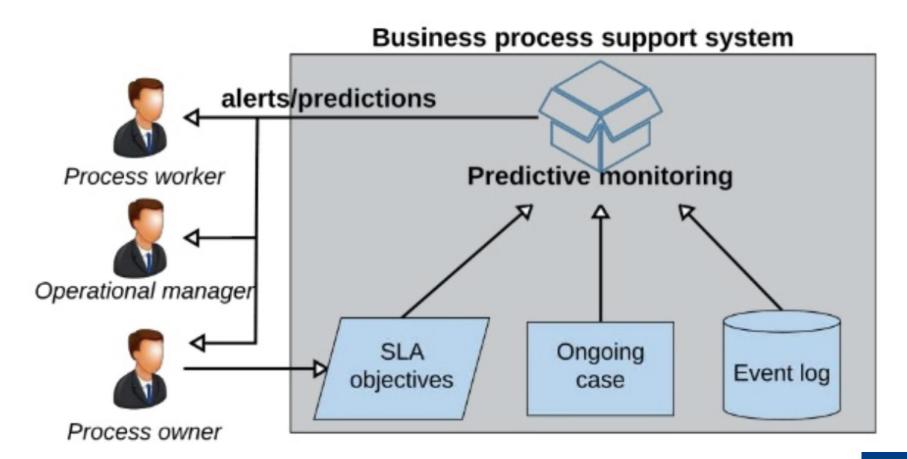




- Absolute time and its abstractions:
 - Between 9 pm and 6 am
 - On Sundays
 - •
- Relative time
 - Time from the last feeding
 - Time from falling asleep
- But also combinations of the two:
 - Time from the last feeding till 6am

Predictive Business Process Monitoring



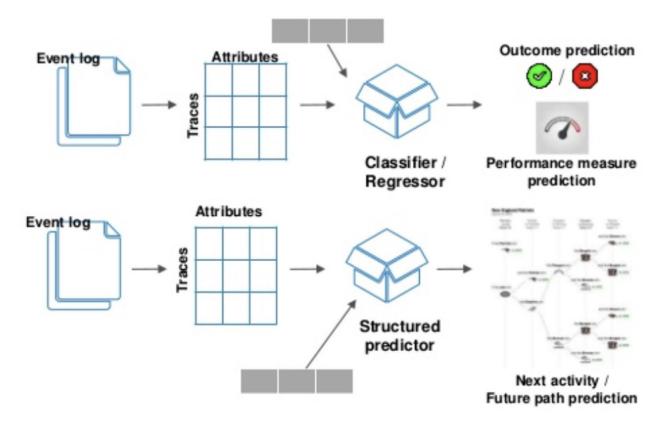


Business Process Monitoring



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Predictive Process Monitoring: General Approach



Predictive Monitoring Example



Case id	Event id	Properties				•
		Timestamp	Activity	Resource	Cost	
	35654423	30-12-2010:11.02	register request	Pete	50	***
	35654424	31-12-2010:10.06	examine thoroughly	Suc	400	
	35654425	05-01-2011:15.12	check ticket	Mike	100	***
	35654426	06-01-2011:11.18	decide	Sara	200	
	35654427	07-01-2011:14.24	reject request	Pete	200	
	35654483	30-12-2010:11.32	register request	Mike	50	***
	35654485	30-12-2010:12.12	check ticket	Mike	100	***
	35654487	30-12-2010:14.16	examine casually	Pete	400	***
	35654488	05-01-2011:11.22	decide	Sara	200	***
	35654489	08-01-2011:12.05	pay compensation	Ellen	200	***
	35654521	30-12-2010:14.32	register request	Pete	50	***
	35654522	30-12-2010:15.06	examine casually	Mike	400	***
	35654524	30-12-2010:16.34	check ticket	Ellen	100	***
	35654525	06-01-2011:09.18	decide	Sara	200	***
	35654526	06-01-2011:12.18	reinitiate request	Sara	200	***
	35654527	06-01-2011:13.06	examine thoroughly	Sean	400	
	35654530	08-01-2011:11.43	check ticket	Pete	100	***
	35654531	09-01-2011:09.55	decide	Sara	200	***
	35654533	15-01-2011:10.45	pay compensation	Ellen	200	***

Sequence encoding(1)



Description Description		
C001 19-10-2016 Calculate risk score Bob Mortgage 100 000 C001 20-10-2016 Make offer Mike Mortgage 100 000 C001 25-10-2016 Make offer Mike Mortgage 100 000 C002 20-10-2016 Check completeness Sue Car 15 000 C002 20-10-2016 Check credit history Sue Car 15 000 C002 22-10-2016 Calculate risk score Elsa Car 15 000 C002 24-10-2016 Reject application Elsa Car 16 000 C003 02-11-2016 Check completeness Maria Mortgage 30 000 C003 04-11-2016 Ask for additional data Maria Mortgage 30 000 C003 10-11-2016 Check credit history Maria Mortgage 30 000		
C001 20-10-2016 Make offer Mike Mortgage 100 000 C001 25-10-2016 Make offer Mike Mortgage 100 000 C002 20-10-2016 Check completeness Sue Car 15 000 C002 20-10-2016 Check credit history Sue Car 15 000 C002 22-10-2016 Calculate risk score Elsa Car 15 000 C002 24-10-2016 Reject application Elsa Car 16 000 C003 02-11-2016 Check completeness Maria Mortgage 30 000 C003 04-11-2016 Ask for additional data Maria Mortgage 30 000 C003 10-11-2016 Check credit history Maria Mortgage 30 000	-	
C001 25-10-2016 Make offer Mike Mortgage 100 000 C002 20-10-2016 Check completeness Sue Car 15 000 C002 20-10-2016 Check credit history Sue Car 15 000 C002 22-10-2016 Calculate risk score Elsa Car 15 000 C002 24-10-2016 Reject application Elsa Car 15 000 C003 02-11-2016 Check completeness Maria Mortgage 30 000 C003 04-11-2016 Ask for additional data Maria Mortgage 30 000 C003 10-11-2016 Check credit history Maria Mortgage 30 000	-	
C002 20-10-2016 Check completeness Sue Car 15 000 C002 20-10-2016 Check credit history Sue Car 15 000 C002 22-10-2016 Calculate risk score Elsa Car 15 000 C002 24-10-2016 Reject application Elsa Car 16 000 C003 02-11-2016 Check completeness Maria Mortgage 30 000 C003 04-11-2016 Ask for additional data Maria Mortgage 30 000 C003 10-11-2016 Check credit history Maria Mortgage 30 000	70 000	
C002 20-10-2016 Check credit history Sue Car 15 000 C002 22-10-2016 Calculate risk score Elsa Car 15 000 C002 24-10-2016 Reject application Elsa Car 16 000 C003 02-11-2016 Check completeness Maria Mortgage 30 000 C003 04-11-2016 Ask for additional data Maria Mortgage 30 000 C003 10-11-2016 Check credit history Maria Mortgage 30 000	80 000	
C002 22-10-2016 Calculate risk score Elsa Car 15 000 C002 24-10-2016 Reject application Elsa Car 15 000 C003 02-11-2016 Check completeness Maria Mortgage 30 000 C003 04-11-2016 Ask for additional data Maria Mortgage 30 000 C003 10-11-2016 Check credit history Maria Mortgage 30 000		
C002 24-10-2016 Reject application Elsa Car 16 000 C003 02-11-2016 Check completeness Maria Mortgage 30 000 C003 04-11-2016 Ask for additional data Maria Mortgage 30 000 C003 10-11-2016 Check credit history Maria Mortgage 30 000	-	
C003 02-11-2016 Check completeness Maria Mortgage 30 000 C003 04-11-2016 Ask for additional data Maria Mortgage 30 000 C003 10-11-2016 Check credit history Maria Mortgage 30 000	-	
C003 04-11-2016 Ask for additional data Maria Mortgage 30 000 C003 10-11-2016 Check credit history Maria Mortgage 30 000		
C003 10-11-2016 Check credit history Maria Mortgage 30 000	-	
-	-	
Feature vector x	141	
	Teopet y	
x ² 1 0 0 1 0 0 14 0 0 0 .	1	
	. 1	
	0	
4 0 0 1 0 1 0 10 0 0 0	1	

Sequence encoding(2)



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Index-based encoding

id	loan goal	activity_1	1111	activity.m	amount_1	***	$amount_m$	label
1	mortgage	check completeness		make offer	0		80 000	1
2		check completeness			0	110	0	0

Case ID	Timestamp	Activity	Resource	Loan-goal	Requested amt	Offered ami
C001	18-10-2016	Check completeness	Sue	Mortgage	100 000	
C0001	19-10-2016	Check credit history	Sue	Mortgage	100 000	-
C001	19-10-2016	Calculate risk score	Bob	Mortgage	100 000	- 2
C001	20-10-2016	Make offer	Mise	Mortgage	100 000	70 000
C001	25-10-2016	Maise offer	Mise	Mortgage	100 000	80 000
C002	20-10-2016	Check completeness	Sue	Car	15 000	
C0002	20-10-2016	Check credit history	Sue	Car	15 000	-
C002	22-10-2016	Calculate risk score	Elsa	Car	15-000	-
C002	24-10-2016	Reject application	Elsa	Car	15 000	-
C003	02-11-2016	Check completeness	Maria	Mortgage	30 000	-
C003	04-11-2016	Ask for additional data	Maria	Mortgage	30 000	-
C003	10-11-2016	Check credit history	Maria	Mortgage	30 000	
	-				-	-

Aggregation encoding

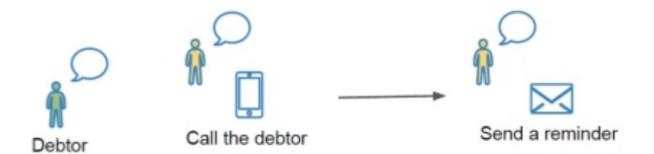
id	loan goal	${\rm count}({\rm check\ compl-ness})$	1100	$\operatorname{count}(\operatorname{make\ offer})$	$\min(\mathrm{amt})$	***	$\max(\mathrm{amt})$	label
1	mortgage	1	450	1	0	***	80 000	1
2	car	1	1111	0	0	110	0	0

▷ LSTM

id	event	loan goal	event	amount	label
	1	mortgage	check completeness	0	
1	210	441.	***	100	1
	m	mortgage	make offer	80 000	
_	1	car	check completeness	0	
2		***	111	1111	0
	m	car	calculate risk score	0	

Predictive Monitoring with Unstructured Data





	Event1	Event2	Resource1	Resource2	Debtor	Summary1	Summary2
Trace1	Call the debtor	Send a reminder	Sue	Bob	Mark	?	?

Text Mining



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The last ten years has seen a surge of interest in design science research in information systems, organizations, process modelling and software engineering. In this talk I present a framework for design science that shows how in design science research, we iterate over designing new artifacts for a context, and empirically investigating these artifacts in this context. To be relevant, the artifacts should potentially contribute to organizational goals, and to be empirically sound, research to validate new artifacts should provide insight into the effects of using these artifacts in an organizational context. The logic of both of these activities, design and empirical research, is that of rational decision making. I show how this logic can be used to structure reports about our technical or empirical research. This gives us checklists for the design cycle used in technical research and for the empirical cycle used in empirical research. Finally, I will discuss in more detail what the role of theories in design science research is, and how we use theory to state research questions and to generalize the research results.

0.2, 0.1, 0.8, 0.5, ..., 0.1

The tutorial first introduces the PPM including its activities: problem understanding, method finding, modeling, reconciliation, and validation.

0.4, 0.8, 1.0, 0.2, ..., 0.4

What is a good business process model and how do you get value from it? We have for many years worked with SEQUAL, a general framework for understanding the quality of models and modelling languages, which covers all main aspects relative to quality of models. The tramework has been widely cited since the first version was presented in the nineties, and the tutorial will focus on the most recent version of the framework (2016), specialised for quality of business process models, with a focus on how to achieve value through long-term usage of business process models in organizations.

0.9, 0.0, 0.4, 0.5, ..., 0.2

Business process models have gained significant importance due to their critical role for managing business processes. Process models not only play a fundamental role for obtaining a common understanding of an organization's business processes, but are also important assets for improving business processes and to support the development of information systems. In this tutorial we will focus on the process of process modeling (PPM) and shed light on how process models are created.

0.2, 0.3, 0.7, 0.6, ..., 0.6

Text-Extended Index-Based Encoding





	Event1	Event2	Resource1	Resource2	Debtor	t,,	***	t _{in}	t ₂₁	***	t _{2n}
Trace1		Send a reminder	Sue	Bob	Mark	0.2	***	0.1	0.4	***	0.4

- Bag-of-N-grams
- Weighted bag-of-N-grams
- Latent Dirichlet Allocation (LDA)
- Paragraph Vector (PV)

LSTM-Based Predictive Process Monitoring



