

Training in a Virtual Learning Environment: A Process Mining Approach

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Scenario



- Workers' training
 - ✓ 360 Video
 - ✓ Virtual Reality
 - ✓ **Artificial Intelligence** → predicting the learning achievements of the trained workers

ARTIFICIAL
INTELLIGENCE



Novelties

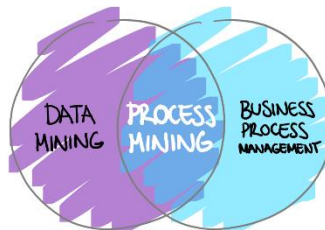
- The scenario is different from the one that is commonly investigated in the education studies ...
 - ✓ Students are **workers** instead of under-graduate or post-graduate students
 - ✓ The learning environment involves **Virtual Technologies** (360 Video+VR) instead of classrooms, traditional e-learning tools (e.g. MOOCs)
 - ✓ **Artificial Intelligence** to replace multiple-choice tests to determine the learning outcome achievements

Novelties

- Tracking the worker behaviour in the virtual learning environment and **logging** the **tracked data**



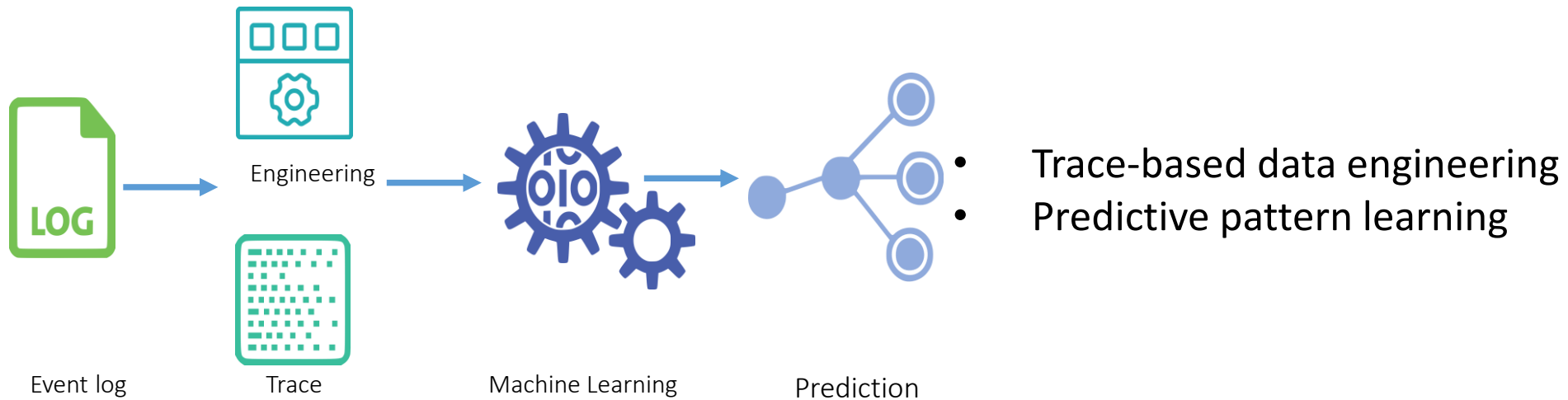
- Representing the logged data of the training session of each worker as a **trace** of a training process and designing a **predictive process mining** approach



Trace-defined training sessions

Event	Worker	Scene	Time	Completion	Elapsed time	Effort	
E1	W1	Scene1	2019-1-10,10:01:32	Completed	17442	18	
E2	W1	Scene2	2019-1-10,10:03:02	Completed	27413	27	
E3	W1	Scene3	2019-1-10,10:10:02	Completed	122709	48	
E4	W1	Scene4	2019-1-10,10:13:22	Completed	100567	21	
E5	W1	Scene5	2019-1-10,10:15:01	Uncompleted	458000	45	
E6	W1	Scene5	2019-1-10,10:15:10	Uncompleted	3770223	37	
E7	W1	Scene5	2019-1-10,10:15:19	Completed	557000	61	
	W1	PASSED					
E8	W2	Scene1	2019-1-10,10:30:31	Completed	18244	21	
...	

ViTE- Virtual Training Exam prediction



Data engineering

- Multi-perspective data engineering
 - ✓ Scene perspective
 - ✓ Control-flow perspective
 - ✓ Performance perspective

Data engineering – scene

Event	Worker	Scene	Time	Completion	Elapsed time	Effort
E1	W1	Scene1	2019-1-10,10:01:32	Completed	17442	18
E2	W1	Scene2	2019-1-10,10:03:02	Completed	27413	27
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E7	W1	Scene5	2019-1-10,10:15:19	Completed	557000	61

PASSED

Scene1	Scene2	Scene3	Scene4	Scene5
1	1	1	1	3

Scene counter

Scene1C	Scene2C	Scene3C	Scene4C	Scene5C
100%	100%	100%	100%	33.3%

Completion scene

Data engineering – control flow

Event	Worker	Scene	Time	Completion	Elapsed time	Effort
E1	W1	Scene1	2019-1-10,10:01:32	Completed	17442	18
E2	W1	Scene2	2019-1-10,10:03:02	Completed	27413	27
E3	W1	Scene3	2019-1-10,10:10:02	Completed	122709	48
E4	W1	Scene4	2019-1-10,10:13:22	Completed	100567	21
E5	W1	Scene5	2019-1-10,10:15:01	Uncompleted	458000	45
E6	W1	Scene5	2019-1-10,10:15:10	Uncompleted	3770223	37
E7	W1	Scene5	2019-1-10,10:15:19	Completed	557000	61

PASSED

Control-flow counter

Scene1C→Scene1C	Scene1C→Scene1U	Scene1C→Scene2C	...	Scene5U→Scene5U
0	0	1		1



Data engineering – performance

Event	Worker	Scene	Time	Completion	Elapsed time	Effort
E1	W1	Scene1	2019-1-10,10:01:32	Completed	17442	18
E2	W1	Scene2	2019-1-10,10:03:02	Completed	27413	27
E3	W1	Scene3	2019-1-10,10:10:02	Completed	122709	48
E4	W1	Scene4	2019-1-10,10:13:22	Completed	100567	21
E5	W1	Scene5	2019-1-10,10:15:01	Uncompleted	458000	45
E6	W1	Scene5	2019-1-10,10:15:10	Uncompleted	3770223	37
E7	W1	Scene5	2019-1-10,10:15:19	Completed	557000	61

PASSED

count	sum	min	max	median	mean
7	1660153	17442	557000	122709	237164.71

Trace-level time

Scene1Sum	Scene1Min	Scene1Max	...	Scene5Mean
17442	17442	17442		1595074.3

Scene-level time

Data engineering – performance

Event	Worker	Scene	Time	Completion	Elapsed time	Effort
E1	W1	Scene1	2019-1-10,10:01:32	Completed	17442	18
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E7	W1	Scene5	2019-1-10,10:15:19	Completed	557000	61

PASSED

Trace-level effort

sum	min	max	median	mean
257	18	61	37	36.71

Scene-level effort

Scene1Sum	Scene1Min	Scene1Max	Scene1Median	...	Scene5Mean
18	18	18	18		47.6

Predictive pattern learning

- Any supervised classification algorithm can be selected, in order to learn a pattern to predict the training outcome based on the features constructed in the data engineering process
- The Principal Component Analysis is performed to deal with the curse of dimensionality

Empirical validation – data

- Virtual training platform developed by MTM Project S.R.L with the content provided by MTM PROJECT S.R.L and Cinemagica S.R.L.
- 97 maintenance technicians have performed a virtual training session with 5 scenes

Empirical validation – data

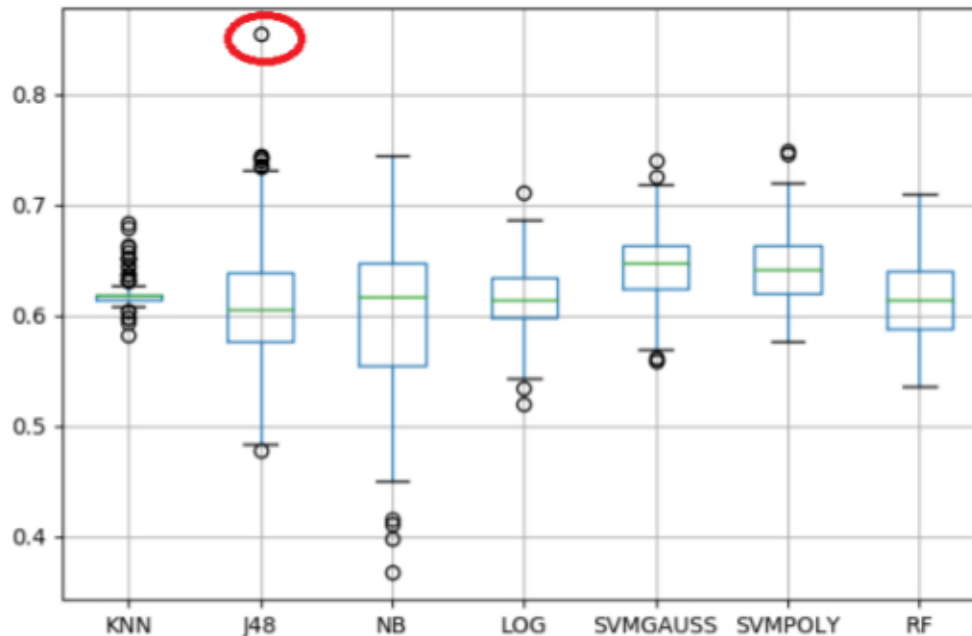
- A training team has manually evaluated the learning achievements of workers after the virtual training test with a multi-choice test
 - ✓ 71 technicians → passed
 - ✓ 26 technicians → rejected

Empirical validation – methodology

- Leave-one-out validation
- Classification algorithm
 - ✓ KNN, J48, Naive Bayes – NB, Logistic Regression – Log, SVM (Gaussian kernel – GAUSS, polynomial kernel – POLY), Random Forest – RF
- Accuracy performance evaluation
 - ✓ F-score

Empirical validation – results by classification algorithm

F-score



Empirical validation – results by data engineering schema

J48

configuration	F-score
scene	0.6186
transition	0.7181
performance	0.6337
scene, transition	0.5830
scene, performance	0.6701
transition, performance	0.7075
scene, transition, performance	0.8537

Final remarks

Kometa + ViTE	Accuracy	F-score
	0.8556	0.8537

- The accuracy of ViTE is in step with accuracy values in the recent literature for task of outcome prediction in education ranging between 0.7 and 0.81

Future work

- Extending the validation by considering data collected through **new training sessions**
- Framing the approach in a **streaming environment**, in order to learn a predictive pattern that may change over time as new training sessions are completed
- Exploring the use of **transfer learning** to transfer the predictive pattern learned with a training session to a session having a new topic

Thank you for your attention

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