

Process Mining and Intelligence Project

Emotion Based Music Selection

Ettore Ricci

Francesco Boldrini

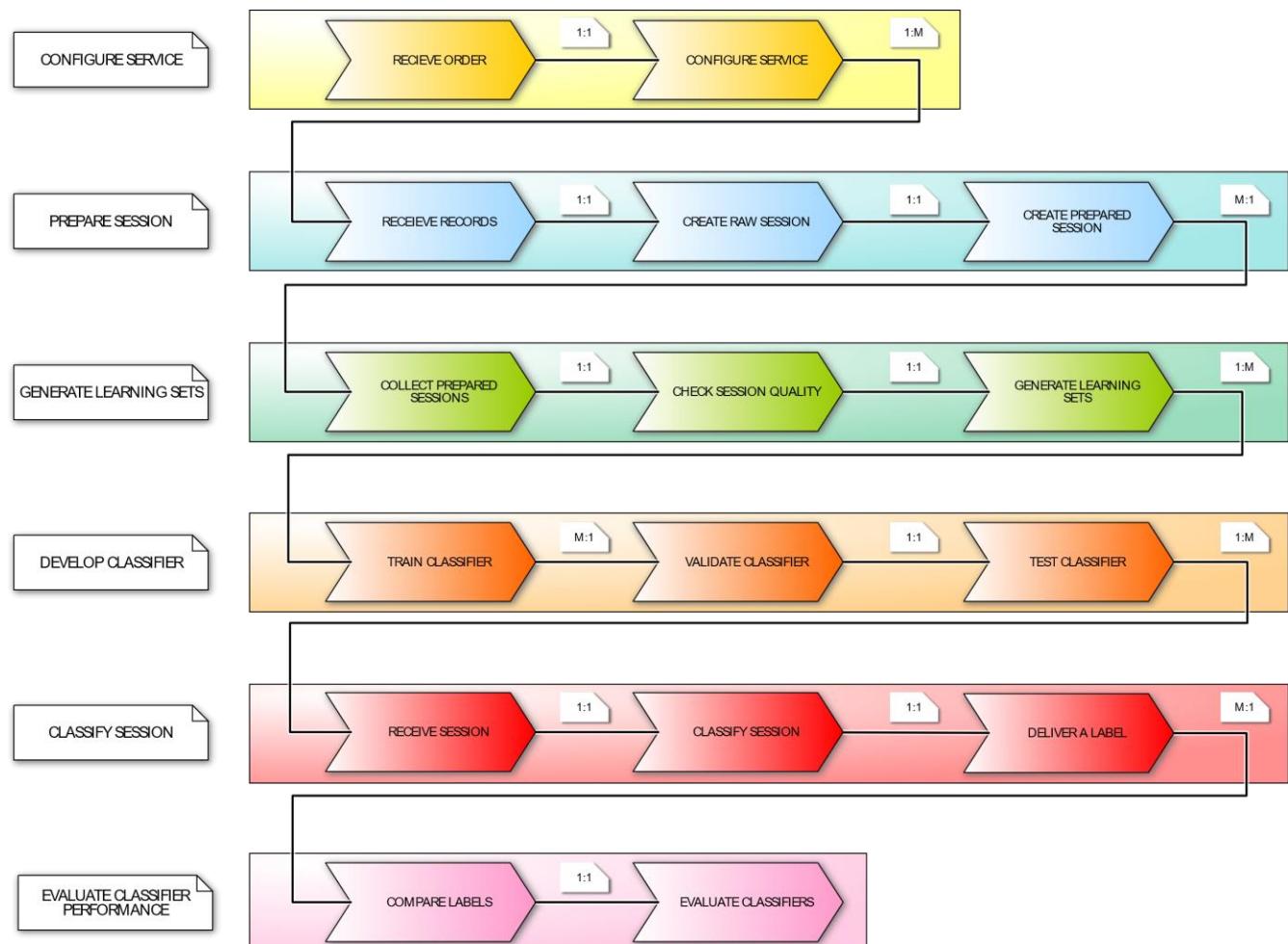
Paolo Palumbo

Zahra Omrani

BPMN modeling

Process landscape

[Ettore Ricci, Paolo Palumbo, Francesco Boldrini, Zahra Omrani]

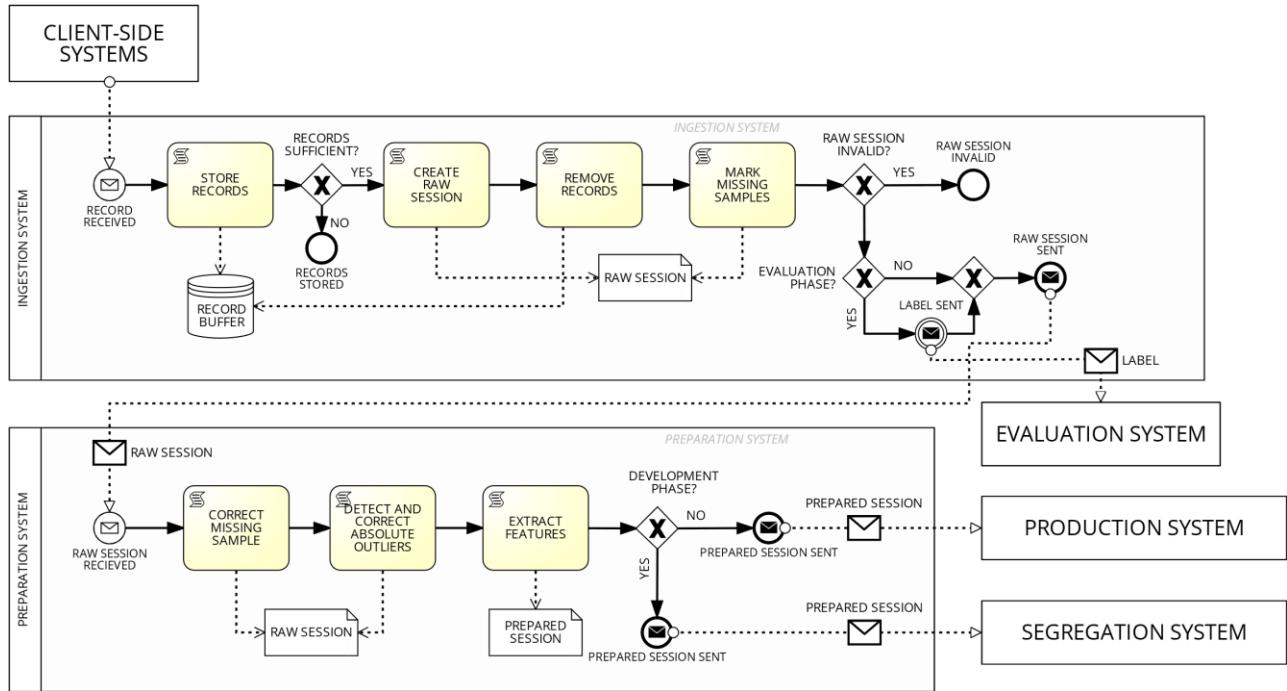


Process landscape

Process model

Prepare session

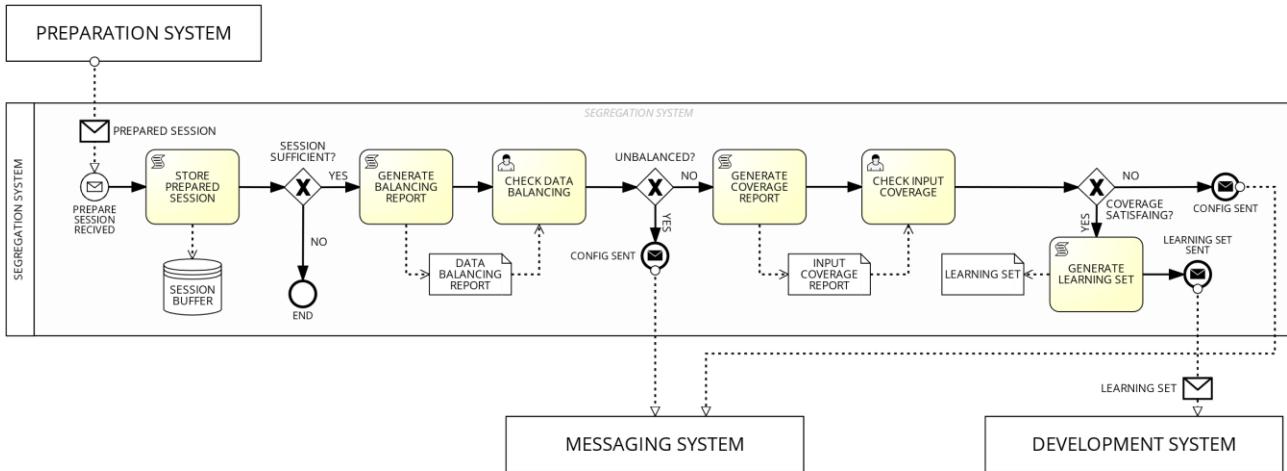
[Ettore Ricci, Paolo Palumbo]



Business Diagram of the "Prepare session" process

Generate learning sets

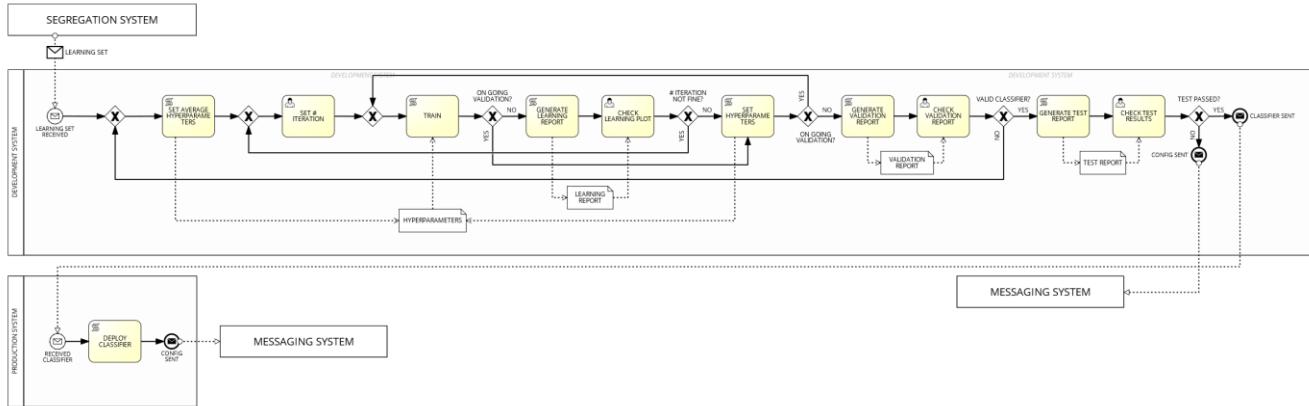
[Ettore Ricci, Paolo Palumbo]



Business Diagram of the "Generate learning sets" process

Develop classifier

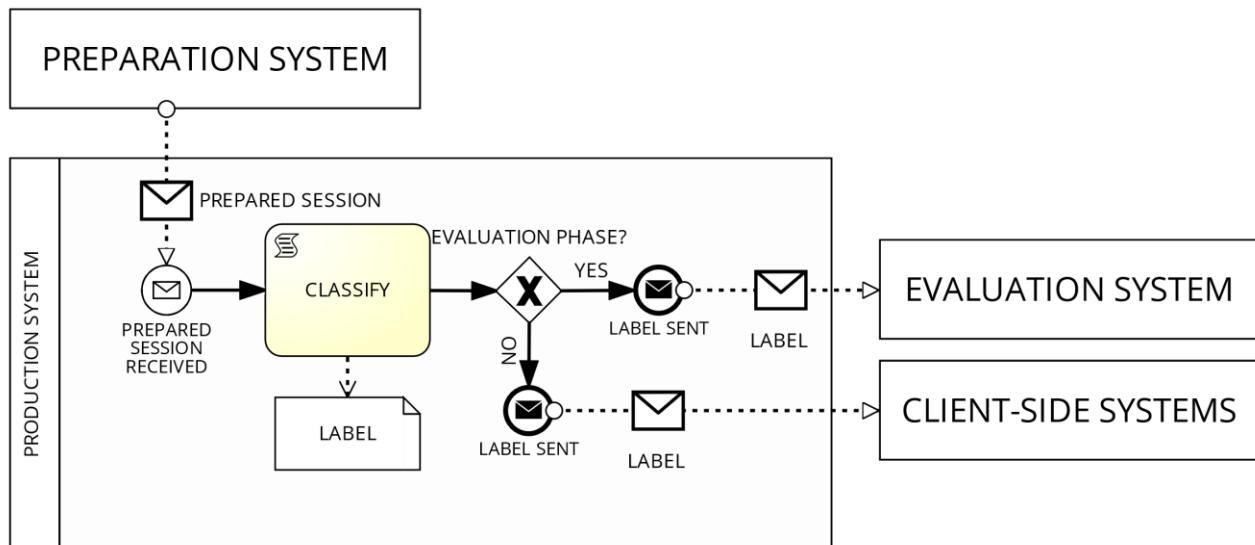
[Ettore Ricci, Paolo Palumbo]



Business Diagram of the "Develop classifier" process

Classify session

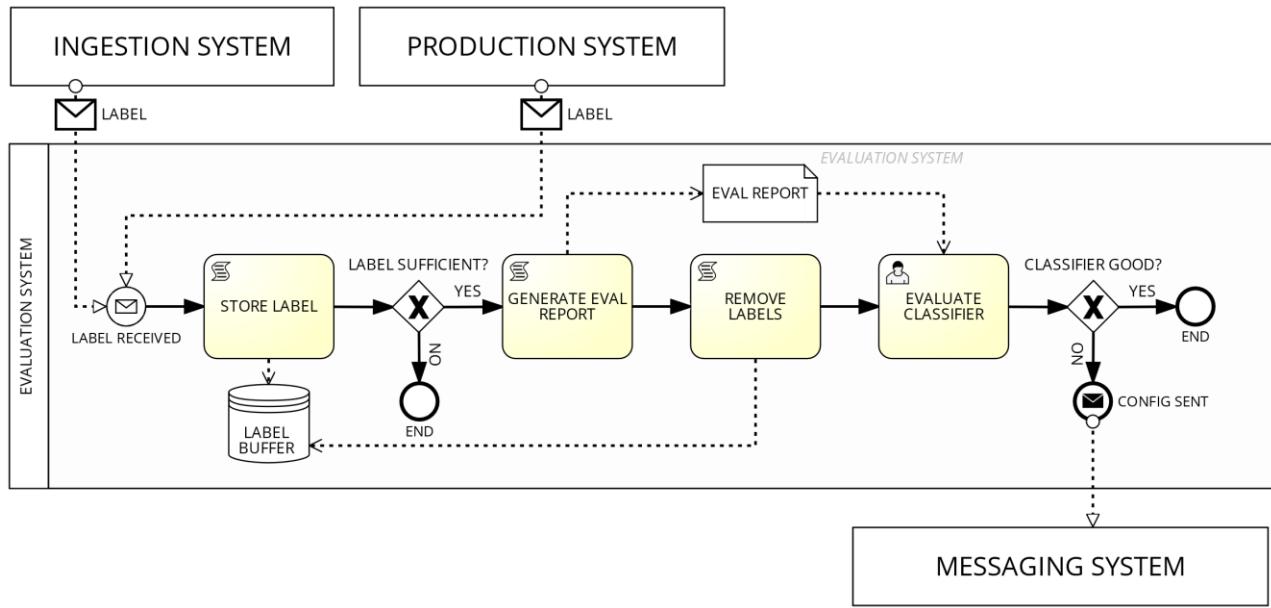
[Ettore Ricci, Paolo Palumbo]



Business Diagram of the "Classify session" process

Evaluate classifier performance

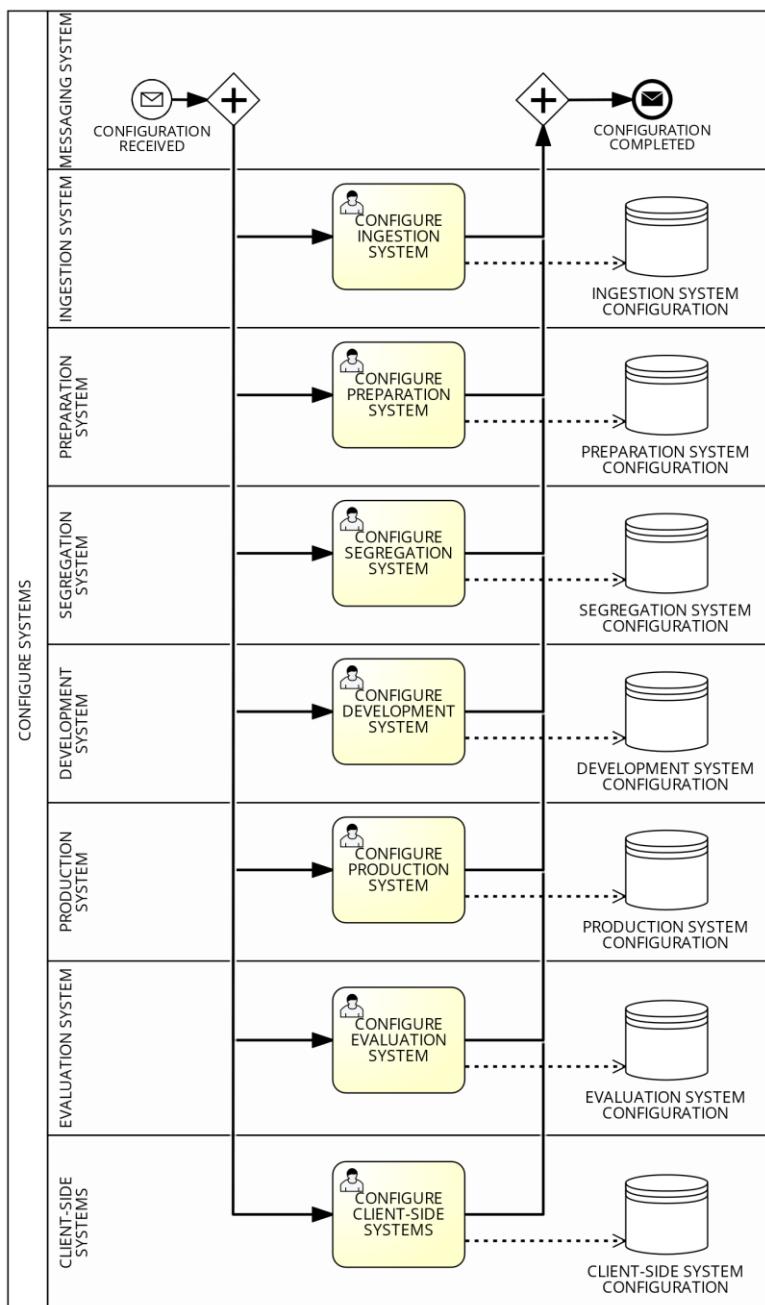
[Ettore Ricci, Paolo Palumbo]



Business Diagram of the "Evaluate classifier performance" process

Configure systems

[Ettore Ricci, Paolo Palumbo]



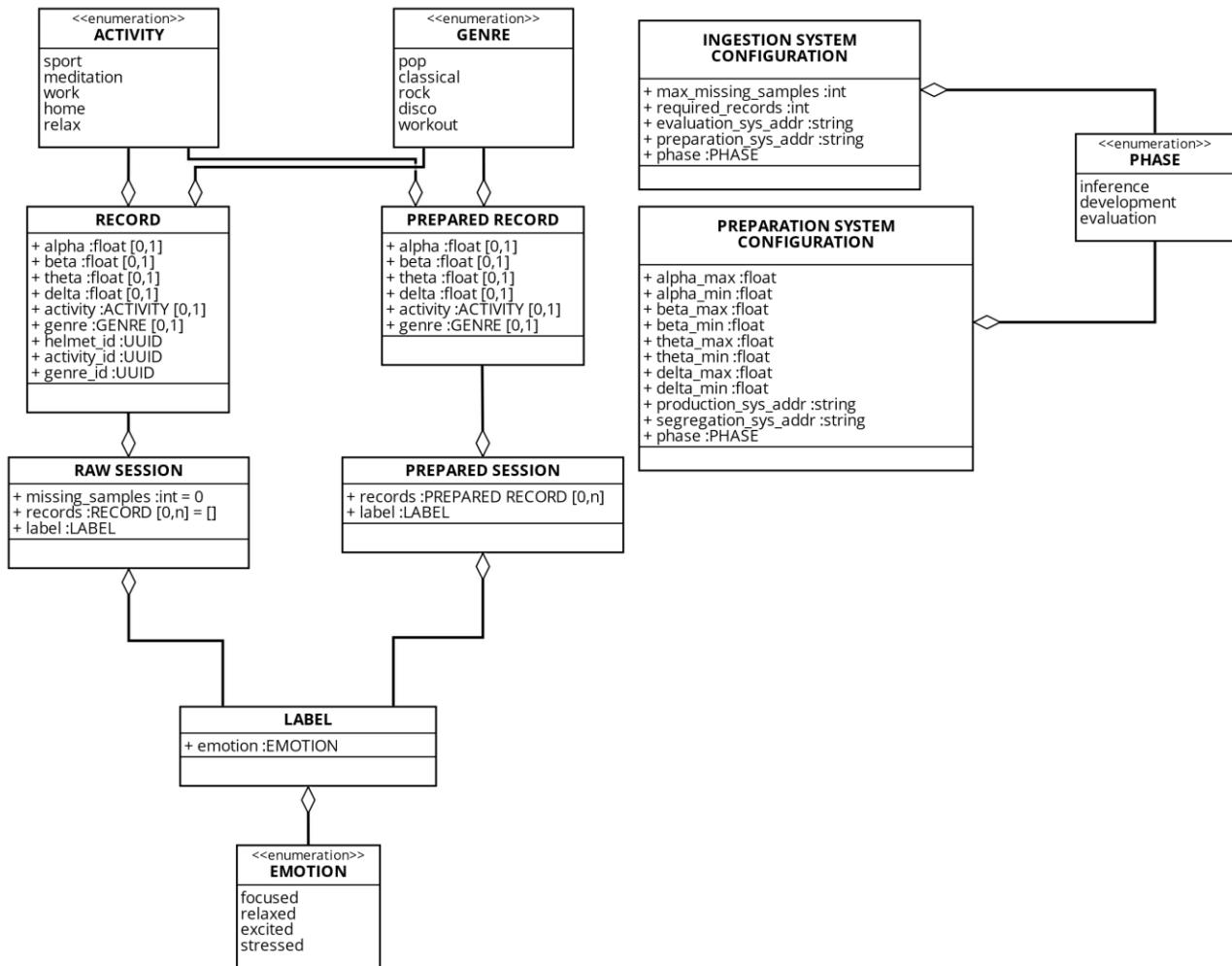
Business Diagram of the "Configure systems" process

Data modeling

Process model

Prepare session

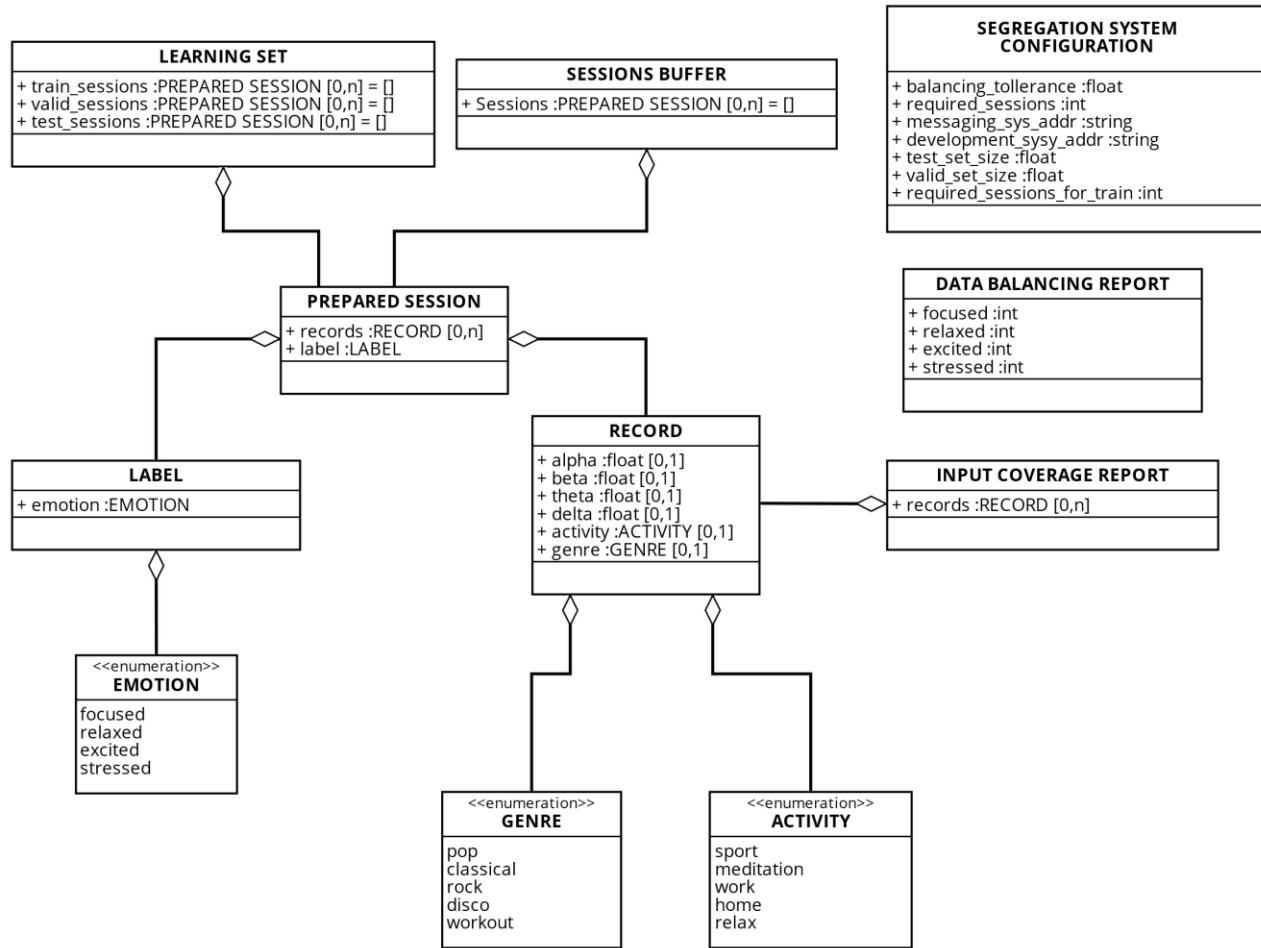
[Ettore Ricci]



Data Model of the "Prepare session" process

Generate learning sets

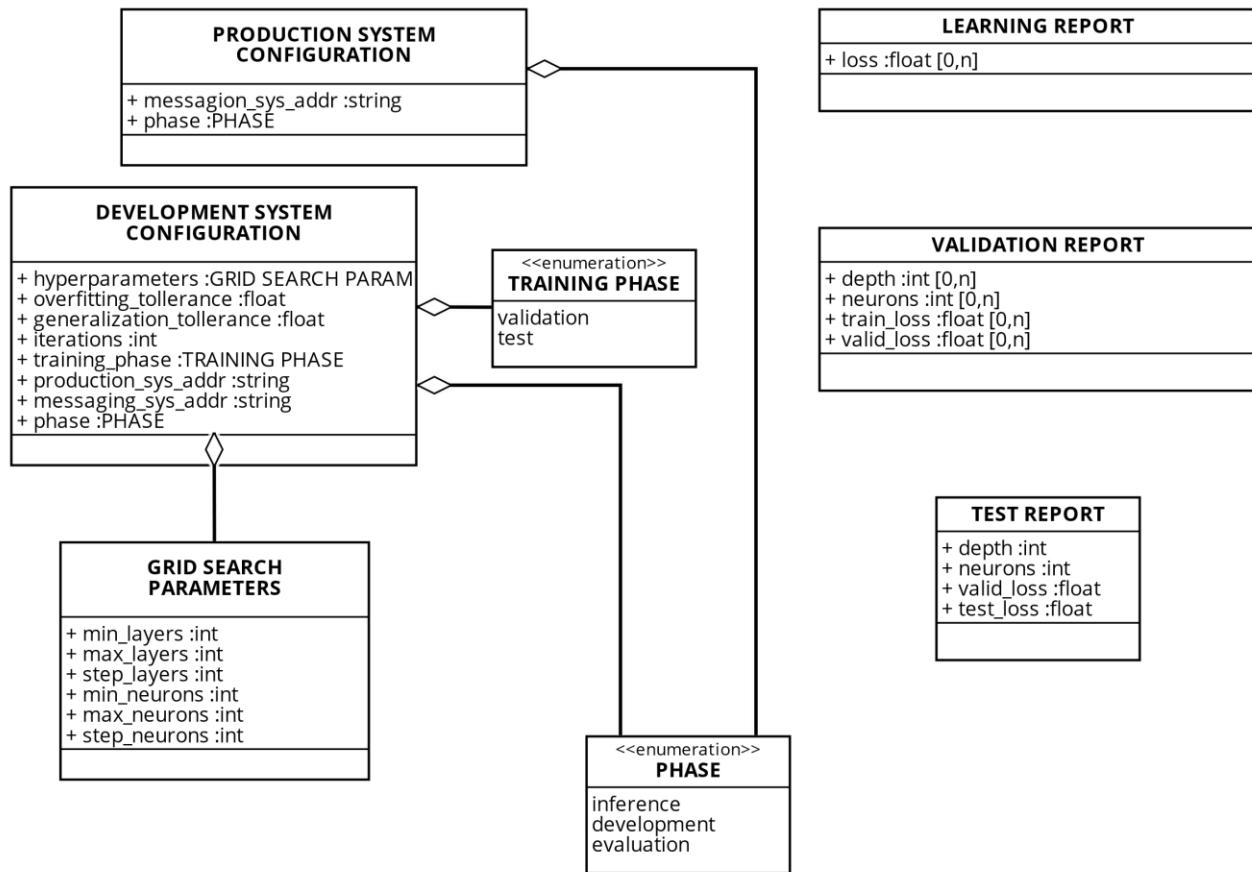
[Paolo Palumbo]



Data Model of the "Generate learning sets" process

Develop classifier

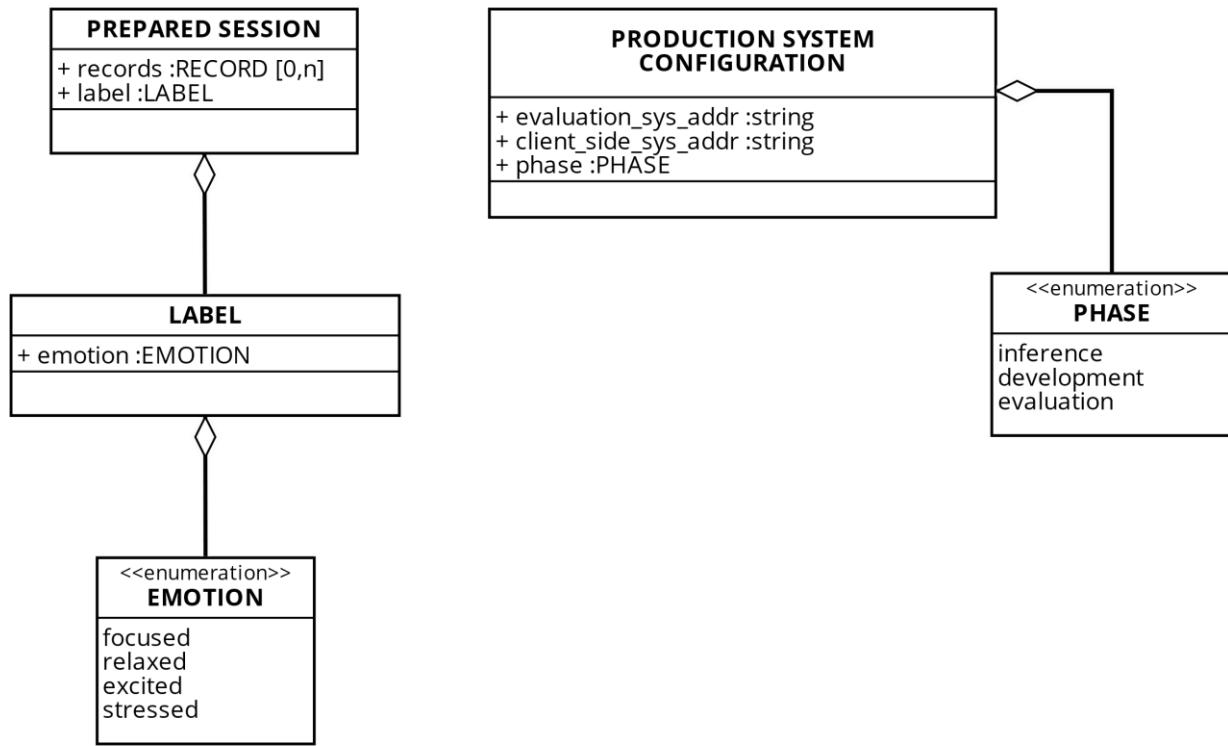
[*Paolo Palumbo*]



Data Model of the "Develop classifier" process

Classify session

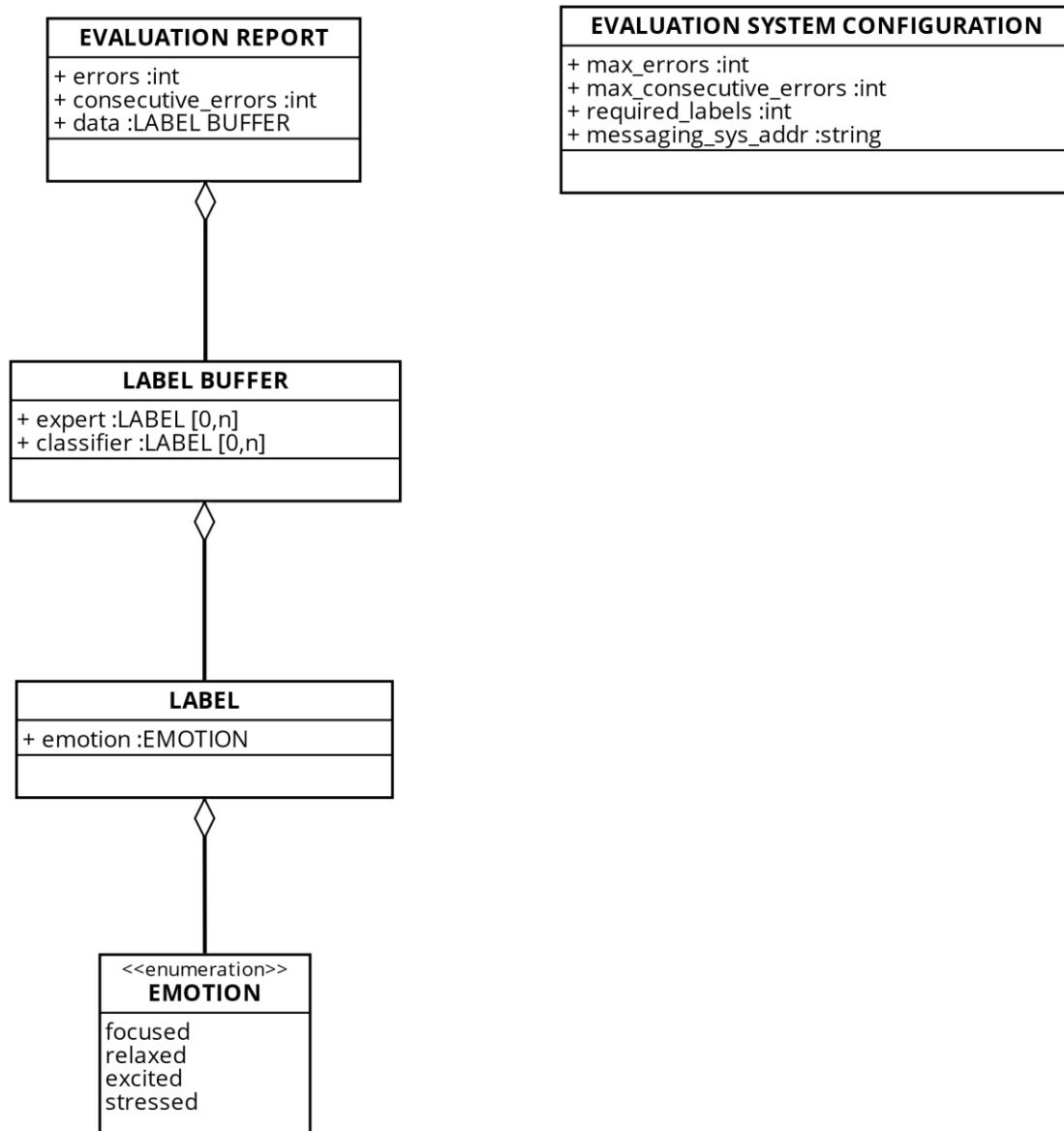
[Francesco Boldrini]



Data Model of the "Classify session" process

Evaluate classifier performance

[*Zahra Omrani*]



Data Model of the "Evaluate classifier performance" process

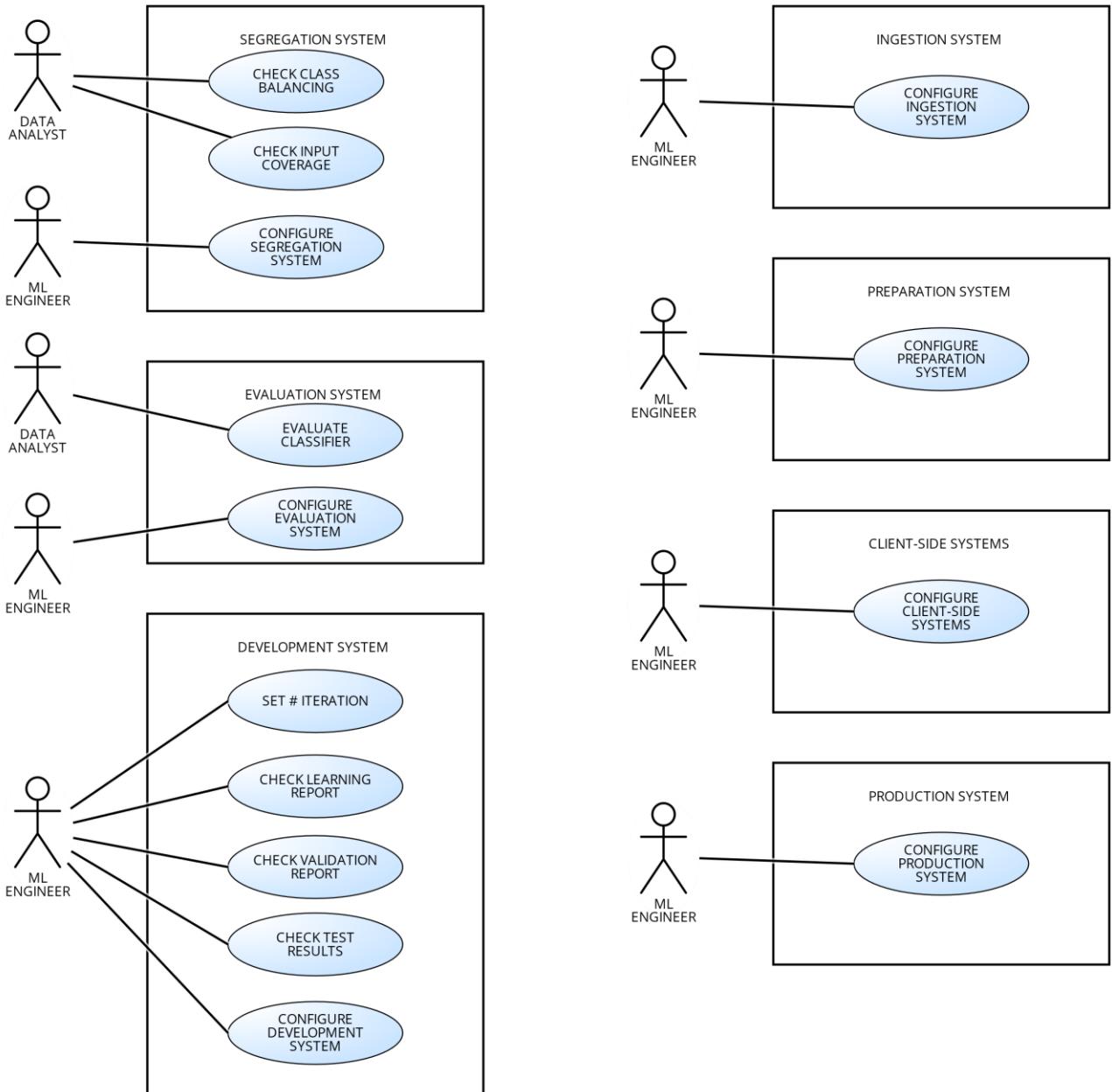
Task level modeling

Roles and salaries

[Ettore Ricci, Paolo Palumbo]

Salary and normalized salary for each position

Position	Description	Salary	Normalized Salary
Clerk	Handles administrative tasks, organizes documentation, and assists with data entry and labeling. Ensures smooth operations by coordinating communication and managing resources.	\$52,000.00	1.00
Data analyst	Prepares, analyzes, and visualizes data to extract insights. Collaborates on cleaning datasets, identifying trends, and supporting model validation.	\$60,000.00	1.15
ML engineer	Builds, tests, and deploys machine learning models, optimizing performance and scalability. Integrates AI solutions into production systems with a focus on efficiency.	\$130,000.00	2.50
Data scientist	Designs and experiments with AI models, applying advanced techniques to solve project challenges. Collaborates with experts to integrate domain knowledge and refine outputs.	\$123,000.00	2.37
Domain expert (Neurologist)	Provides medical expertise to guide AI development and validate results. Ensures solutions align with clinical standards and address neurological challenges.	\$267,000.00	5.13
Minimum		\$52,000.00	1.00



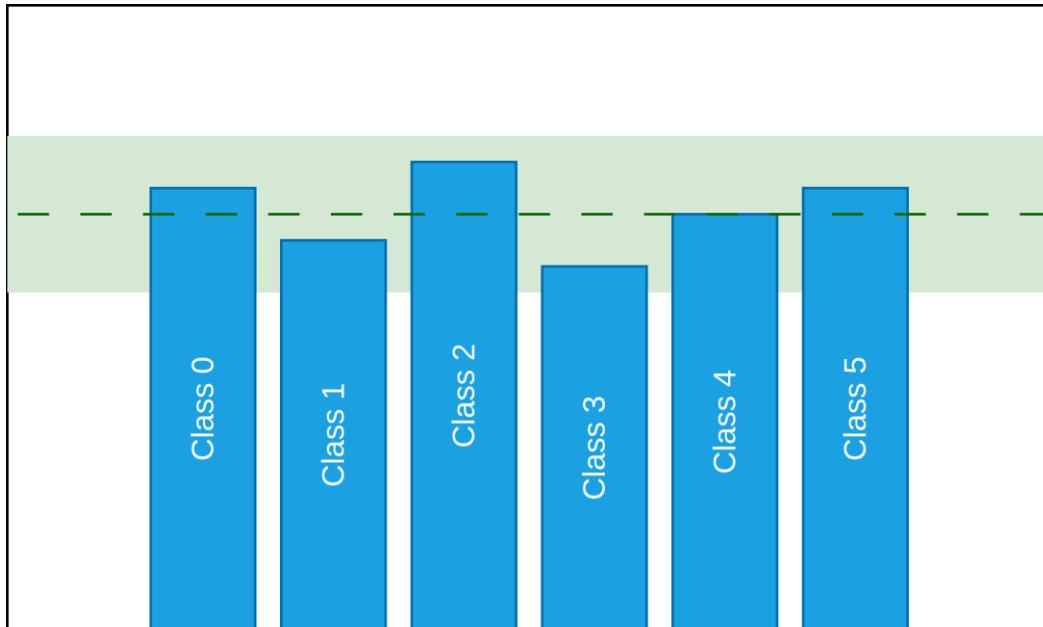
Use case diagram

Segregation system

Check data balancing

[*Ettore Ricci, Paolo Palumbo*]

The task is performed by a Data Analyst.



Balanced: yes

Tolerance: 5%

Balanced

Unbalanced

"Check data balancing" mock-up form

Detailed use case for "Check data balancing" task

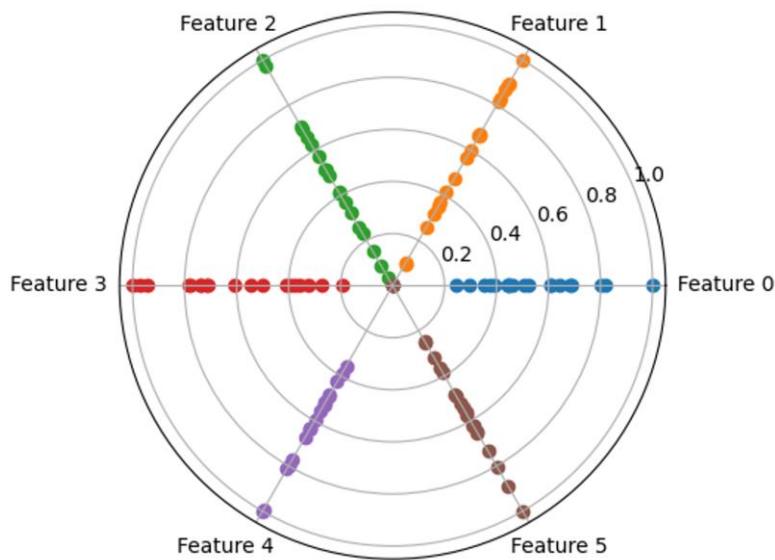
O - Occurrence, CL - Cognitive Level, S - Normalized Salary, SC - Step Cost

Step	O	CL	S	SC
1 ACTOR opens "Check data balancing" form.	1	1	1.15	1.15
2 SYSTEM shows the report.				
3 SYSTEM shows a hint whether the data is balanced or not.				
4 ACTOR checks the hint to see if the data is balanced or not.	1	2	1.15	2.30
5.1 IF the data is balanced.	0.2			
5.1.1 ACTOR clicks "Balanced" button.	0.2	1	1.15	0.23
5.2 ELSE	0.8			
5.2.1 ACTOR clicks "Unbalanced" button.	0.8	1	1.15	0.92
7 SYSTEM shows a confirmation dialog.				
8 ACTOR closes the form.	1	1	1.15	1.15
Human task cost				5.74

Check input coverage

[Ettore Ricci, Paolo Palumbo]

The task is performed by a Data Analyst.



Accept

Reject

"Check input coverage" mock-up form

Detailed use case for "Check input coverage" task

O - Occurrence, CL - Cognitive Level, S - Normalized Salary, SC - Step Cost

Step	O	CL	S	SC
1 ACTOR opens "Check input coverage" form.	1	1	1.15	1.15
2 SYSTEM shows a radar scatter plot of the input distribution.				
3 FOR EACH radius in the radar scatter plot:	6			
3.1 ACTOR checks if the distribution is uniform on the radius.	6	4	1.15	27.6
3.1.1 IF the distribution is not uniform as expected.	4			
3.1.1.1 THEN the input coverage is not satisfied.	4			
4.1 IF the input coverage is satisfied.	0.33			
4.1.1 ACTOR clicks "Accept" button.	0.33	1	1.15	0.38
4.2 ELSE	0.66			
4.2.1 ACTOR clicks "Reject" button.	0.66	1	1.15	0.76
5 SYSTEM shows a confirmation dialog.				
6 ACTOR closes the form.	1	1	1.15	1.15
Human task cost				31.04

Configure Segregation System

[Francesco Boldrini, Zahra Omrani]

This task is performed by a ML Engineer.

CONFIGURE SEGREGATION SYSTEM

CURRENT CONFIGURATION

Balancing Tollerance	10	Messaging Addr.	192.168.1.91
Required Sessions	10	Develop Addr.	192.168.1.94
Test Set Size	10		
Validation Set Size	10		
Required Sessions - Train.	10		

SAVE CONFIG

"Configure Segregation System" mock-up form

Detailed use case for "Configure Segregation System" task

O - Occurrence, CL - Cognitive Level, S - Normalized Salary, SC - Step Cost

Step	O	CL	S	SC
1 ACTOR opens the "Configure Segregation System" form.	1	1	2.50	2.50
2 SYSTEM displays the current configuration.				
3 ACTOR sets the <i>balancing_tolerance</i> .	1	4	2.50	10
4 ACTOR sets the <i>required_sessions</i> .	1	4	2.50	10
5 ACTOR sets the <i>messaging_sys_addr</i> .	1	1	2.50	2.50
6 ACTOR sets the <i>development_sys_addr</i> .	1	1	2.50	2.50
7 ACTOR sets the <i>test_set_size</i> .	1	4	2.50	10
8 ACTOR sets the <i>valid_set_size</i> .	1	4	2.50	10
9 ACTOR sets the <i>required_sessions_for_train</i> .	1	4	2.50	10
10 SYSTEM validates the configuration.				
10.1 IF the configuration is correct and properly formatted:				
10.1.1 SYSTEM displays a confirmation message.				
10.2 ELSE (if the configuration is incorrect):				
10.2.1 SYSTEM displays an error message and aborts the process.				

Step	O	CL	S	SC
11 ACTOR saves the form.	1	1	2.50	2.50
Human task cost				60

Development system

Set iteration number

[*Zahra Omrani*]

The task is performed by a ML engineer.

Set Iteration Number

Current Iteration Number 10

Enter New Iteration Number:

Iteration number updated successfully!

"Set iteration number" mock-up form

Detailed use case for "Set iteration number" task

O - Occurrence, CL - Cognitive Level, S - Normalized Salary, SC - Step Cost

Step	O	CL	S	SC
1 ACTOR opens "Set Iteration Number" form.	1	1	2.5	2.5
2 SYSTEM displays the current iteration number.				
3.1 IF it's the first configuration:				
3.1.1 ACTOR inputs the desired number of iterations based on task complexity and previous experience.	0.002	3	2.5	0.015
3.2 ELSE (subsequent configurations):				
3.2.1 ACTOR inputs the number based on the established learning curve.	0.998	1	2.5	2.495
4 ACTOR clicks "Submit" button to confirm the iteration number.	1	1	2.5	2.5
5 SYSTEM shows a confirmation dialog.				
6 ACTOR closes the form.	1	1	2.5	2.5
Human task cost				10.01

Check learning report

[Paolo Palumbo]

The task is performed by a ML engineer.



"Check learning report" mock-up form

Detailed use case for "Check training report" task

O - Occurrence, CL - Cognitive Level, S - Normalized Salary, SC - Step Cost

Step	O	CL	S	SC
1 ACTOR opens "Check training report" form.	1	1	2.50	2.50
2 SYSTEM shows the training loss curve.				
3 ACTOR checks the learning curve.	1	3	2.50	7.50
3.1 IF the loss is flat for at least half of the iterations:	0.4			
3.1.1 THEN ACTOR clicks "Overfit" button.	0.4	1	2.50	1.00
3.2 IF the loss is not flat at the end of the iterations:	0.4			
3.2.1 THEN ACTOR clicks "Underfit" button.	0.4	1	2.50	1.00
3.3 ELSE	0.2			
3.3.1 ACTOR clicks "Approved" button.	0.2	1	2.50	0.50
4 SYSTEM shows a confirmation dialog.				
5 ACTOR closes the form.	1	1	2.50	2.50
Human task cost				15

Check validation report

[Ettore Ricci]

This task is performed by a ML engineer.

ID	Depth	Neurons	Train MSE	Valid MSE	Delta MSE
1	954	3	4000	0.13	0.14 0.01
2	321	4	3000	0.23	0.24 0.01
3	5	3	1000	0.35	0.35 0.00
4	764	2	2000	0.24	0.45 0.21
5	202	3	2500	0.20	0.47 0.27
Reject					
Overfitting Tolerance:					0.10

"Check validation report" mock-up form

Detailed use case for "Check validation report" task

O - Occurrence, CL - Cognitive Level, S - Normalized Salary, SC - Step Cost

Step	O	CL	S	SC
1 ACTOR opens "Check validation report" form.	1	1	2.5	2.5
2 SYSTEM shows the best 5 models sorted by increasing Validation Loss.				
3 FOR EACH model in the list:	5			
3.1 ACTOR calculates model Validation Loss minus the Training Loss	1	3	2.5	7.5
3.2 IF 3.1 is less than the Overfitting Tolerance and the Best Model is not selected.	1	3	2.5	7.5
3.2.1 THEN ACTOR selects the model as the Best Model.	1	1	2.5	2.5
4 FOR EACH model in the list aside from the previous:	4			
4.1 ACTOR calculates model Validation Loss minus the Training Loss	1	3	2.5	7.5

Step	O	CL	S	SC
4.2 IF 4.1 is less than the Overfitting Tolerance and the Second Best Model is not selected.	1	3	2.5	7.5
4.2.1 THEN select the model as the Second Best Model.	0.25	1	2.5	0.625
5 ACTOR calculates if the Validation Loss of the Second Best Model is one order of magnitude greater than the Validation Loss of the Best Model.	1	3	2.5	7.5
6.1 IF the Best Model is not selected.	0.05	1	2.5	0.125
6.1.1 ACTOR clicks "Reject" button.	0.05	1	2.5	0.125
6.2 ELSE IF the Second Best Model is not selected or 5 is true	0.3	3	2.5	2.25
6.2.1 ACTOR clicks on the Best Model.	0.3	1	2.5	0.75
6.3 ELSE	0.65	3	2.5	4.875
6.3.1 ACTOR clicks on the least complex model among the Best Model and the Second Best Model.	0.65	3	2.5	4.875
7 SYSTEM shows a confirmation dialog.				
8 ACTOR closes the form.	1	1	2.5	2.5
Human task cost				175.5

Check test results

[Ettore Ricci]

This task is performed by a ML engineer.

ID	Depth	Neurons	Valid MSE	Test MSE	Delta MSE
954	3	4000	0.14	0.15	0.01
Overfitting Tolerance:					0.10
Accept			Reject		

"Check test results" mock-up form

Detailed use case for "Check test results" task

O - Occurrence, CL - Cognitive Level, S - Normalized Salary, SC - Step Cost

Step	O	CL	S	SC
1 ACTOR opens "Check test results" form.	1	1	2.5	2.5
2 SYSTEM shows the test results.				

Step	O	CL	S	SC
3 ACTOR checks if the difference between the test results and the validation results is within overfitting tolerance.	1	2	2.5	5
4.1 IF the test results is not satisfactory.	0.01			
4.1.1 ACTOR clicks "Reject" button.	0.01	1	2.5	0.025
4.2 ELSE	0.99			
4.2.1 ACTOR clicks "Approve" button.	0.99	1	2.5	2.475
5 SYSTEM shows a confirmation dialog.				
6 ACTOR closes the form.	1	1	2.5	2.5
Human task cost				12.5

Configure Development System

[Francesco Boldrini, Zahra Omrani]

This task is performed by a ML Engineer.

CONFIGURE DEVELOPMENT SYSTEM

CURRENT CONFIGURATION

Overfitting Tolerance	10	Iterations	10
Generalization Tolerance	1	Min Layers	10
Production Addr.	192.168.1.90	Step Layers	10
Messaging Addr.	192.168.1.91	Max Layers	10
Min Neurons	10	Step Neurons	10
Select Phase	Inference development evaluation	Select Training Phase	Validation Test
SAVE CONFIG			

"Configure Development System" mock-up form

Detailed use case for "Configure Development" task

O - Occurrence, CL - Cognitive Level, S - Normalized Salary, SC - Step Cost

Step	O	CL	S	SC
1 ACTOR opens the "Configure Development System" form.	1	1	2.50	2.50
2 SYSTEM displays current configuration.				
3 ACTOR sets the <i>min_Layers</i> .	1	4	2.50	10
4 ACTOR sets the <i>max_Layers</i> .	1	4	2.50	10

Step	O	CL	S	SC
5 ACTOR sets the <i>min_neurons</i> .	1	4	2.50	10
6 ACTOR sets the <i>step_Layers</i> .	1	4	2.50	10
7 ACTOR sets the <i>step_neurons</i> .	1	4	2.50	10
8 ACTOR sets the <i>max_neurons</i> .	1	4	2.50	10
9 ACTOR sets the <i>overfitting_tolerance</i> parameter.	1	4	2.50	10
10 ACTOR sets the <i>generalization_tolerance</i> parameter.	1	4	2.50	10
11 ACTOR sets the <i>iterations</i> parameter.	1	4	2.50	10
12 ACTOR choose the <i>training_phase</i> parameter from the drop down (<i>validation, test</i>).	1	1	2.50	2.50
13 ACTOR sets the <i>production_sys_addr</i> parameter.	1	1	2.50	2.50
14 ACTOR sets the <i>messaging_sys_addr</i> parameter.	1	1	2.50	2.50
15 ACTOR choose the <i>phase</i> parameter from the drop down (<i>inference, develop, evaluation</i>).	1	1	2.50	2.50
16.1 SYSTEM IF config is correct and correctly formatted.				
16.1.1 SYSTEM shows a confirmation message.				
16.2 ELSE				
16.2.1 SYSTEM shows error message and aborts.				
17 ACTOR saves the form.	1	1	2.50	2.50
Human task cost				105

Evaluation system

Evaluate classifier performance

[*Zahra Omrani*]

This task is performed by a Data Analyst.

Evaluate Classifier Performance (X)

Session ID	Expert Label	Classifier Label	Error
0	1	2	Yes
1	1	3	Yes
2	2	1	Yes
3	3	3	No

Max number of errors tolerated (th1): 4
Max number of consecutive error tolerated (th 2) :2

th1 satisfied
 $3 < 4$

th 2 exceeded
 $3 > 2$

Pass

Fail

"Evaluate Classifier Performance" mock-up form

Detailed use case for "Evaluate Classifier Performance" task

O - Occurrence, CL - Cognitive Level, S - Normalized Salary, SC - Step Cost

Step	O	CL	S	SC
1 ACTOR opens the "Evaluate Classifier Performance" form.	1	1	1.15	1.15
2 SYSTEM displays a table of sessions with Expert Label (ground truth) and Classifier Label (predicted label). The difference between the labels (if any) represents an error.				

Step	O	CL	S	SC
3.1 ACTOR checks the total errors threshold color.	1	2	1.15	2.30
3.2 ACTOR checks the consecutive errors threshold color	1	2	1.15	2.30
3.3 IF at least one threshold is red				
3.3.1 ACTOR clicks the "Fail" button.	0.14	1	1.15	0.161
3.4 ELSE				
3.4.1 ACTOR clicks the "Pass" button.	0.86	1	1.15	0.989
4 SYSTEM shows a confirmation dialog.				
5 ACTOR closes the form.	1	1	1.15	1.15
Human task cost				8.05

Configure Evaluation System

[Francesco Boldrini, Zahra Omrani]

This task is performed by a ML Engineer.

The mock-up shows a blue rounded rectangle containing a configuration interface. At the top center is the title 'CONFIGURE EVALUATION SYSTEM'. Below it is a section titled 'CURRENT CONFIGURATION' with four parameter pairs: 'Max Errors' set to 10, 'Max Consecutive Errors' set to 10, 'Required Labels' set to 10, and 'Messaging Addr.' set to 192.168.1.91. At the bottom left is a green button labeled 'SAVE CONFIG'.

"Configure Evaluation System" mock-up form

Detailed use case for "Configure Evaluation" task

O - Occurrence, CL - Cognitive Level, S - Normalized Salary, SC - Step Cost

Step	O	CL	S	SC
1 ACTOR opens the "Configure Evaluation System" form.	1	1	2.50	2.50
2 SYSTEM displays current configuration.				
3 ACTOR sets the <i>max_errors</i> parameter.	1	4	2.50	10
4 ACTOR sets the <i>max_consecutive_errors</i> parameter.	1	4	2.50	10

Step	O	CL	S	SC
5 ACTOR sets the <i>required_Labels</i> parameter.	1	4	2.50	10
6 ACTOR sets the <i>messaging_sys_addr</i> parameter.	1	1	2.50	2.50
7.1 SYSTEM IF config is correct and correctly formatted.				
7.1.1 SYSTEM shows a confirmation message.				
7.2 ELSE				
7.2.1 SYSTEM shows error message and aborts.				
9 ACTOR saves the form.	1	1	2.50	2.50
Human task cost				37.50

Client-Side Systems

Configure Client-Side Systems

[Francesco Boldrini, Zahra Omrani]

This task is performed by a ML Engineer.

The mock-up form is titled "CONFIGURE CLIENT-SIDE SYSTEM". It displays "CURRENT CONFIGURATION" with an input field for "Ingestion Addr." containing the value "192.168.1.95". At the bottom is a green "SAVE CONFIG" button.

"Configure Client-Side Systems" mock-up form

Detailed use case for "Configure Client-Side Systems" task

O - Occurrence, CL - Cognitive Level, S - Normalized Salary, SC - Step Cost

Step	O	CL	S	SC
1 ACTOR opens the "Configure Client-Side System" form.	1	1	2.50	2.50
2 SYSTEM displays current configuration.				
3 ACTOR sets the <i>ingestion_sys_addr</i> parameter.	1	1	2.50	2.50

Step	O	CL	S	SC
4.1 SYSTEM IF config is correct and correctly formatted.				
4.1.1 SYSTEM shows a confirmation message.				
4.2 ELSE				
4.2.1 SYSTEM shows error message and aborts.				
5 ACTOR saves the form.	1	1	2.50	2.50
	Human task cost			7.50

Production System

Configure Production Systems

[Francesco Boldrini, Zahra Omrani]

This task is performed by a ML Engineer.

CONFIGURE PRODUCTION SYSTEM

CURRENT CONFIGURATION

Messaging Addr.	192.168.1.91
Evaluation Addr.	192.168.1.92
Client-Side Addr.	192.168.1.93

Select Phase > Inference
development
evaluation

SAVE CONFIG

"Configure Production System" mock-up form

Detailed use case for "Configure Production" task

O - Occurrence, CL - Cognitive Level, S - Normalized Salary, SC - Step Cost

Step	O	CL	S	SC
1 ACTOR opens the "Configure Production System" form.	1	1	2.50	2.50
2 SYSTEM displays current configuration.				
3 ACTOR sets the <i>production_sys_addr</i> parameter.	1	1	2.50	2.50
4 ACTOR sets the <i>messaging_sys_addr</i> parameter.	1	1	2.50	2.50

Step	O	CL	S	SC
5 ACTOR choose the <i>phase</i> parameter from the drop down (<i>inference</i> , <i>develop</i> , <i>evaluation</i>).	1	1	2.50	2.50
6.1 SYSTEM IF config is correct and correctly formatted.				
6.1.1 SYSTEM shows a confirmation message.				
6.2 ELSE				
6.2.1 SYSTEM shows error message and aborts.				
7 ACTOR saves the form.	1	1	2.50	2.50
	Human task cost		12.50	

Ingestion System

Configure Ingestion System

[Francesco Boldrini, Zahra Omrani]

This task is performed by a ML Engineer.

CONFIGURE INGESTION SYSTEM

CURRENT CONFIGURATION

Max Missing Samples	10
Required Records	10
Evaluation Addr.	192.168.1.92
Preparation Addr.	192.168.1.93
Select Phase	> Inference development evaluation

SAVE CONFIG

"Configure Ingestion System" mock-up form

Detailed use case for "Configure Ingestion System" task

Step	O	CL	S	SC
1 ACTOR opens the "Configure Ingestion System" form.	1	1	2.50	2.50
2 SYSTEM displays the current configuration.				
3 ACTOR sets the <i>max_missing_samples</i> .	1	4	2.50	10
4 ACTOR sets the <i>required_records</i> .	1	4	2.50	10

Step	O	CL	S	SC
5 ACTOR sets the <i>evaluation_sys_addr</i> .	1	1	2.50	2.50
6 ACTOR sets the <i>preparation_sys_addr</i> .	1	1	2.50	2.50
7 ACTOR selects the <i>phase</i> from the dropdown (<i>inference, development, evaluation</i>).	1	1	2.50	2.50
8.1 SYSTEM IF the configurations are correct and properly formatted:				
8.1.1 SYSTEM displays a confirmation message.				
8.2 ELSE (if the configurations are incorrect):				
8.2.1 SYSTEM displays an error message and aborts the process.				
9 ACTOR saves the form.	1	1	2.50	2.50
	Human task cost			
	32.5			

Preparation System

Configure Preparation System

[Francesco Boldrini, Zahra Omrani]

This task is performed by a ML Engineer.

CONFIGURE PREPARATION SYSTEM

CURRENT CONFIGURATION

Alpha Max	10	Beta Max	10
Alpha Min	1	Beta Min	1
Production Addr.	192.168.1.90	Theta Max	20
Segregation Addr.	192.168.1.89	Theta Min	1
Select Phase	> inference development evaluation	Delta Max	2
SAVE CONFIG		Delta Min	10

"Configure Preparation System" mock-up form

Detailed use case for "Configure Preparation System" task

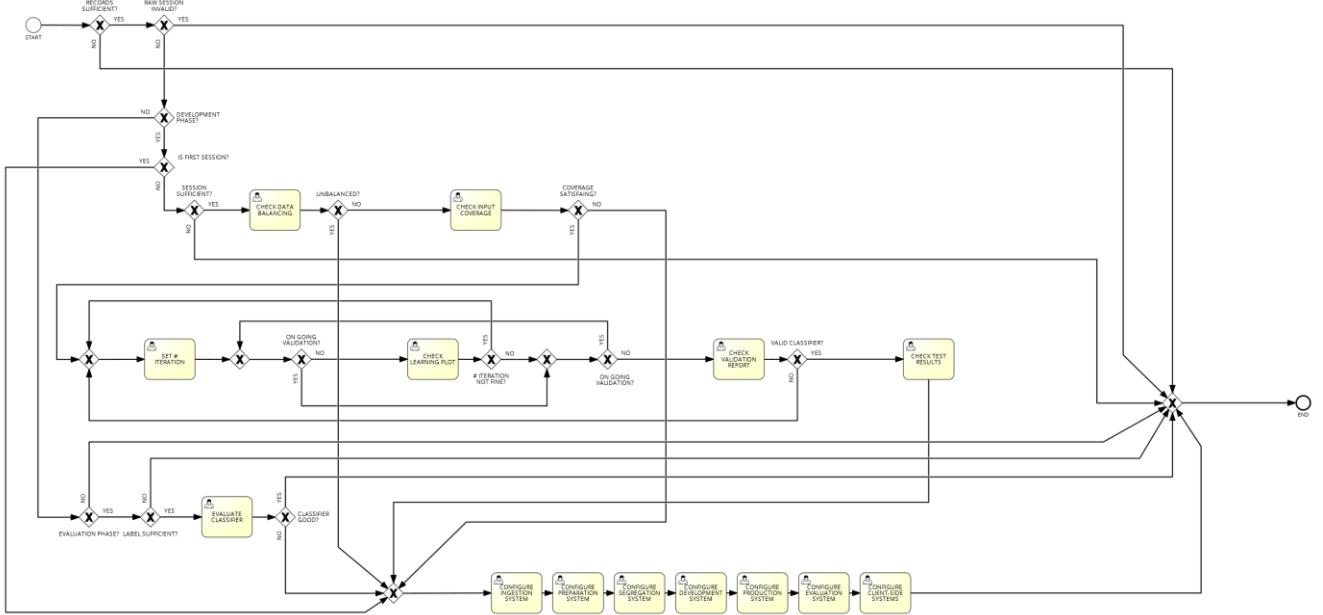
Step	O	CL	S	SC
1 ACTOR opens the "Configure Preparation System" form.	1	1	2.50	2.50
2 SYSTEM displays the current configuration.				

Step	O	CL	S	SC
3 ACTOR sets the <i>alpha_max</i> .	1	4	2.50	10
4 ACTOR sets the <i>alpha_min</i> .	1	4	2.50	10
5 ACTOR sets the <i>beta_max</i> .	1	4	2.50	10
6 ACTOR sets the <i>beta_min</i> .	1	4	2.50	10
7 ACTOR sets the <i>theta_max</i> .	1	4	2.50	10
8 ACTOR sets the <i>theta_min</i> .	1	4	2.50	10
9 ACTOR sets the <i>delta_max</i> .	1	4	2.50	10
10 ACTOR sets the <i>delta_min</i> .	1	4	2.50	10
11 ACTOR sets the <i>production_sys_addr</i> .	1	1	2.50	2.50
12 ACTOR sets the <i>segregation_sys_addr</i> .	1	1	2.50	2.50
13 ACTOR selects the phase from the dropdown (<i>inference, develop, evaluation</i>).	1	1	2.50	2.50
14 SYSTEM IF the configuration is correct and properly formatted:				
14.1 SYSTEM displays a confirmation message.				
14.2 ELSE (if the configuration is incorrect):				
14.2.1 SYSTEM displays an error message and aborts the process.				
15 ACTOR saves the form.	1	1	2.50	2.50
Human task cost				92.5

Simulation

Collapsed workflow

[Ettore Ricci, Paolo Palumbo, Francesco Boldrini]



Collapsed workflow

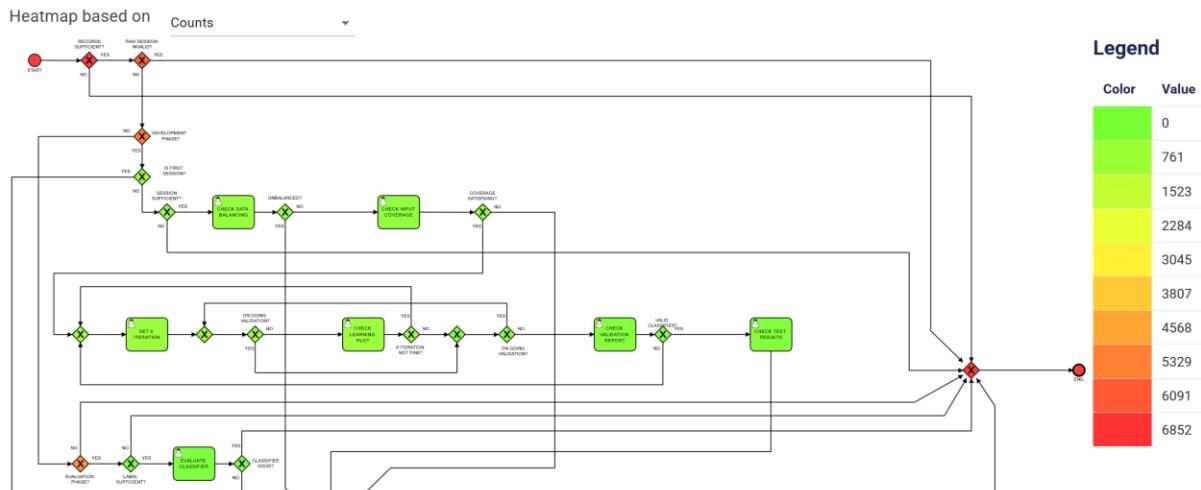
AS-IS Simulation

[Francesco Boldrini, Zahra Omrani]

NB: we set the total number of initial process instances to 6852, as with our assumptions for the two initial gates, where we discard 10% of the sessions twice, we need 6852 sessions to start, to work with

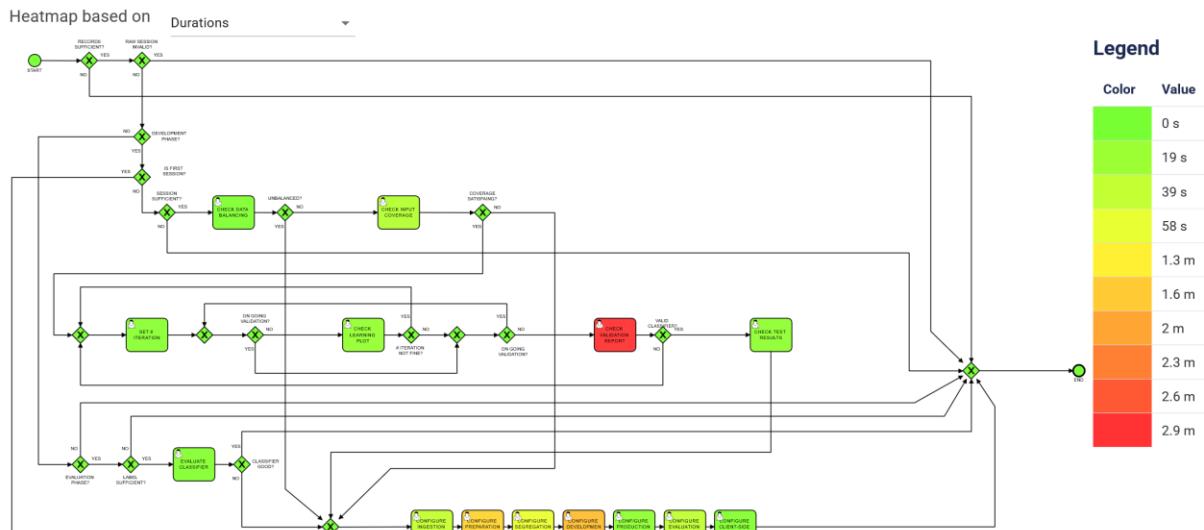
the documentation's assumptions of the 5550 good sessions for all phases altogether

Heatmap



AS-IS Heatmap of the counts of the parameters

Heatmap



AS-IS Heatmap of the time spent in each passage [Durations]

Scenario Statistics

	Minimum	Maximum	Average
Process instance cycle times including off-timetable hours	0 seconds	20.3 minutes	36.9 seconds
Process instance cycle times excluding off-timetable hours	0 seconds	20.3 minutes	36.9 seconds
Process instance costs	0 EUR	0 EUR	0 EUR

Activity Durations, Costs, Waiting times, Deviations from Thresholds

Name	Waiting time				Duration				Duration over threshold				Cost				Cost over threshold			
	Count	Min	Avg	Max	Min	Avg	Max	Min	Avg	Max	Min	Avg	Max	Min	Avg	Max	Min	Avg	Max	
CHECK DATA BALANCING	528	0 s	0 s	0 s	5.5 s	5.7 s	6 s	0 s	0 s	0 s	0	0	0	0	0	0	0	0	0	
CHECK INPUT COVERAGE	434	0 s	0 s	0 s	29.5 s	31 s	32.6 s	0 s	0 s	0 s	0	0	0	0	0	0	0	0	0	
CHECK LEARNING PLOT	604	0 s	0 s	0 s	14.3 s	15 s	15.7 s	0 s	0 s	0 s	0	0	0	0	0	0	0	0	0	
CHECK TEST RESULTS	150	0 s	0 s	0 s	11.9 s	12.5 s	13.1 s	0 s	0 s	0 s	0	0	0	0	0	0	0	0	0	
CHECK VALIDATION REPORT	159	0 s	0 s	0 s	2.8 m	2.9 m	3.1 m	0 s	0 s	0 s	0	0	0	0	0	0	0	0	0	
CONFIGURE CLIENT-SIDE SYSTEMS	537	0 s	0 s	0 s	7.1 s	7.5 s	7.9 s	0 s	0 s	0 s	0	0	0	0	0	0	0	0	0	
CONFIGURE DEVELOPMENT SYSTEM	537	0 s	0 s	0 s	1.7 m	1.7 m	1.8 m	0 s	0 s	0 s	0	0	0	0	0	0	0	0	0	
CONFIGURE EVALUATION SYSTEM	537	0 s	0 s	0 s	35.6 s	37.6 s	39.4 s	0 s	0 s	0 s	0	0	0	0	0	0	0	0	0	
CONFIGURE INGESTION SYSTEM	537	0 s	0 s	0 s	38 s	40 s	42 s	0 s	0 s	0 s	0	0	0	0	0	0	0	0	0	
CONFIGURE PREPARATION SYSTEM	537	0 s	0 s	0 s	1.5 m	1.5 m	1.6 m	0 s	0 s	0 s	0	0	0	0	0	0	0	0	0	
CONFIGURE PRODUCTION SYSTEM	537	0 s	0 s	0 s	11.9 s	12.5 s	13.1 s	0 s	0 s	0 s	0	0	0	0	0	0	0	0	0	
CONFIGURE SEGREGATION SYSTEM	537	0 s	0 s	0 s	57 s	59.9 s	1 m	0 s	0 s	0 s	0	0	0	0	0	0	0	0	0	
EVALUATE CLASSIFIER	29	0 s	0 s	0 s	7.7 s	8.1 s	8.4 s	0 s	0 s	0 s	0	0	0	0	0	0	0	0	0	
SET # ITERATION	658	0 s	0 s	0 s	9.5 s	10 s	10.5 s	0 s	0 s	0 s	0	0	0	0	0	0	0	0	0	

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AS-IS Simulation Results

Simulation Results

General information

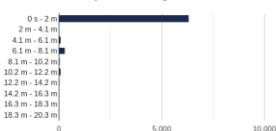
Completed process instances 6852

Total cost 0 EUR

Total simulation time 57 weeks

Charts

Process cycle times including off-timetable hours



Process cycle times excluding off-timetable hours



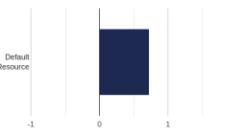
Process waiting times



Process costs (EUR)



Resource utilization %



AS-IS Scenario Statistics

Parameter	% of the Gate	Motivation
# Iteration not fine?	20%	According to the assumptions of the documentation, we set the % of iters that are not fine to 20%
Classifier Good?	86%	According to the assumptions of the docs, classifiers are good 86 of the time
Coverage satisfying?	33%	According to the assumptions of the docs, coverage is satisfying 33% of the time
Development Phase?	9%	In this gate, out of 5550, 500 are in the development phase
Is First Session?	1%	In this gate, out of 500, 1 is in the first session, but the gate won't load less than 1% on BIMP
Labels sufficient?	71%	Given 5550 good sessions, and assuming that those will yields 5 proper classifiers, given the probabilities at the preceding gates and the rounding because of necessity of the gate to have at least 1% as value, we need to have 71% of the sessions to have sufficient labels to respect the documentation's assumptions.
On going Validation?	90%	We set validation to 90% as most of the times this step involves the autonomous systems and not humans
Raw session Invalid?	10%	We assume that 90% of the times the raw session is valid.
Records Sufficient?	90%	We assume that 90% of the times the records are sufficient.
Session Sufficient?	99%	We set sessions sufficient to 99% as with the document's assumptions, we would need roughly 545 sessions to have 5 final good classifiers and here we already start with 500, which is

Parameter	% of the Gate	Motivation
		already lower than what we would need.
Unbalanced?	20%	The documentation assumes that
		20% of the classes are balanced
Valid Classifier?	95%	The documentation assumes that
		95% of the classifiers are valid

Modeling the TO-BE Process

[Francesco Boldrini, Zahra Omrani]

In the context of our application, "Emotion Based Music Selection", we suppose that during the initial configuration phase, we acquire the data through the ECG sensors and the user's schedules, currently playing music and other relevant informations.

This data could be processed by a research center through data-mining approaches: this would mean simplifying the process and making it more efficient, as we could use existing similar classifiers to initialize ours, rather than starting from scratch.

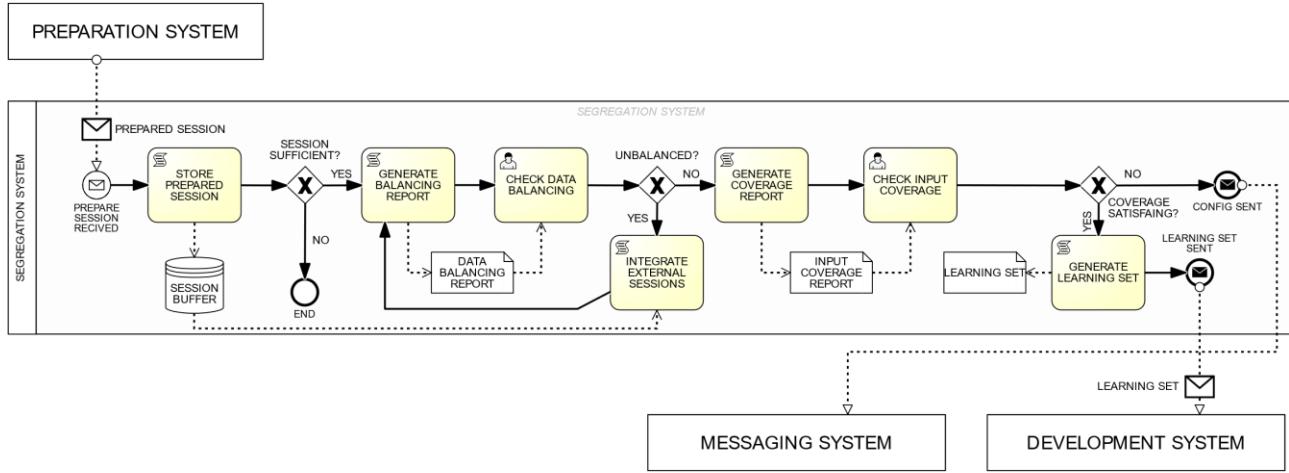
In fact similar networks and classifiers may work well with similar parameters over similar tasks.

For each category it is possible to define some improvement(s):

1. Hand-Off level Improvement(s): Re-use sessions from similar networks in the same category, rather than collecting other sessions.
2. Service Level Improvement(s): Use hyperparameters from similar networks in the same category, rather than starting from scratch.
3. Task Level Improvement(s): Reduce the cognitive effort necessary for the configurations, by starting from default parameters obtained from other similar networks, rather than starting from scratch.

Hand-Off level Improvement(s)

[Francesco Boldrini, Zahra Omrani]



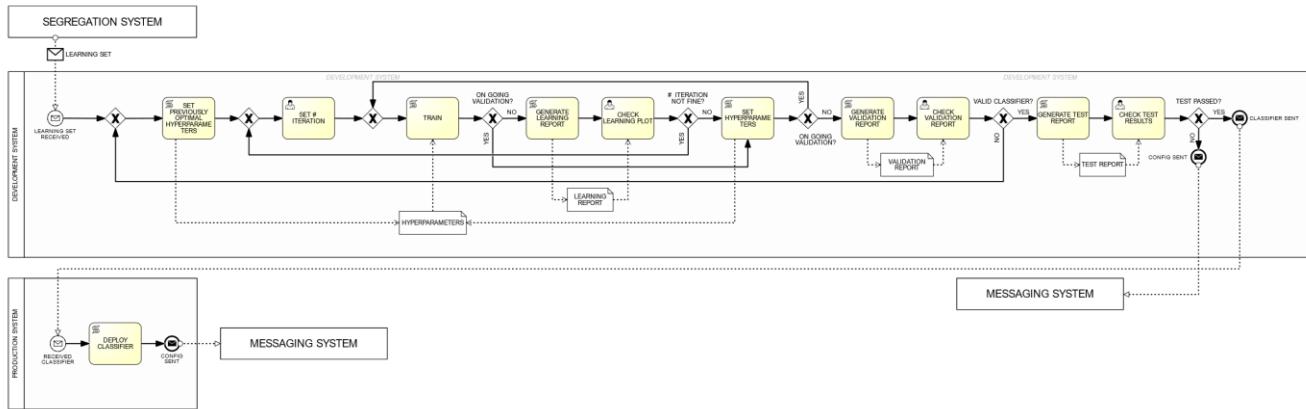
Change to the Generate Learning Sets

We modified the workflow in such a way that our saved previous sessions can be reused in the case of unbalanced data, rather than awaiting for the message system to respond to the issue arisen in the workflow.

This cuts on necessary times to respond to this erroneous situation, as the system can autonomously respond to the issue, rather than waiting for a human to intervene, thus improving the system's efficiency and re-use of data.

Service Level Improvement(s)

[Francesco Boldrini, Zahra Omrani]



Change to the Develop Classifier

The possibility of using hyperparameters from similar trained networks in the same category, rather than starting from scratch, is a great improvement in the service level.

The search for optimized parameters in the network no longer involves brute-forcing the optimization through a grid-search approach but rather re-uses a functioning network's parameters, saving time and computational resources.

Task Level Improvement(s)

[Francesco Boldrini, Zahra Omrani]

The task level improvement(s) involve reducing the cognitive effort necessary for the configurations, by starting from default parameters obtained from other similar networks, rather than starting from scratch.

In particular, we removed the need for a grid search in the check validation report, as we no longer need to check amongst the 5 best networks, but simply have to verify that the network respects the overfitting tolerance, as we are using parameters from a similar network in the same category.

Detailed use case for "Check validation report" task

O - Occurrence, CL - Cognitive Level, S - Normalized Salary, SC - Step Cost

Step	O	CL	S	SC
1 ACTOR opens "Check validation report" form.	1	1	2.5	2.5
2 SYSTEM shows the model trained on optimal parameters				
3 ACTOR calculates model Validation Loss minus the Training Loss	1	3	2.5	7.5
4.1 IF 3 is less than the Overfitting Tolerance	0.95	3	2.5	7.125
3.1.1 THEN ACTOR Confirms the selected model.	0.95	3	2.5	7.125
3.2 ELSE ACTOR Rejects the selected model.	0.05	3	2.5	0.375
6 SYSTEM shows a confirmation dialog.				
7 ACTOR closes the form.	1	1	2.5	2.5
Human task cost				27.125 < 175.5

Furthermore, by using suggested parameters in the configuration phase, we can reduce the cognitive level necessary for the configurations to 2 from 4, by starting from default parameters obtained from other similar networks, rather than needing extensive evaluations (level 4 cognitive level) to find the optimal parameters.

TO-BE detailed use case for "Configure Preparation System" task

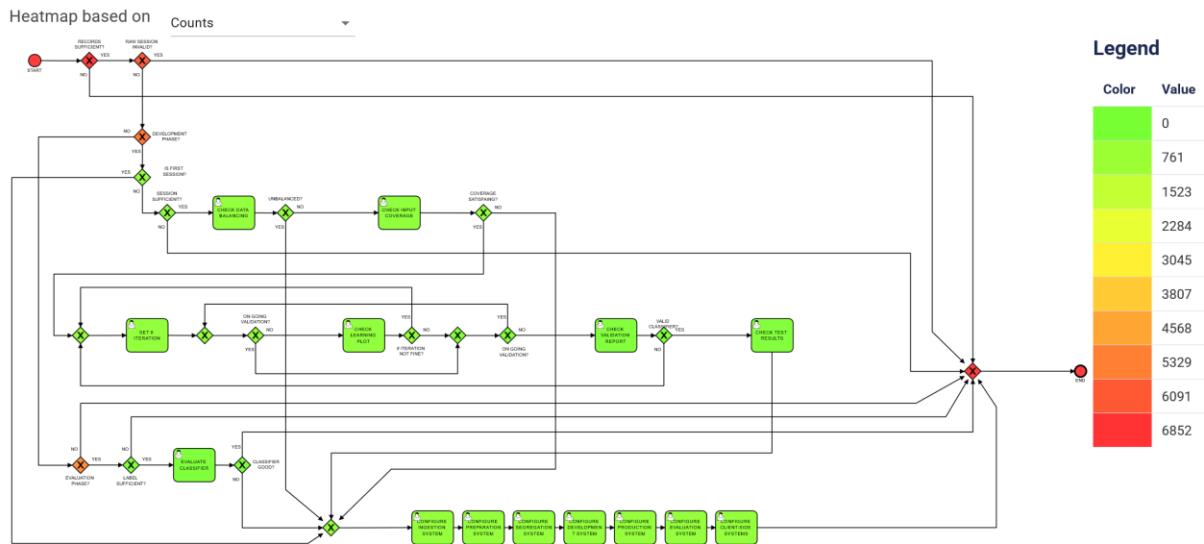
Step	O	CL	S	SC
1 ACTOR opens the "Configure Preparation System" form.	1	1	2.50	2.50
2 SYSTEM displays the current configuration.				
3 ACTOR sets the <i>alpha_max</i> .	1	2	2.50	5
4 ACTOR sets the <i>alpha_min</i> .	1	2	2.50	5
5 ACTOR sets the <i>beta_max</i> .	1	2	2.50	5

Step	O	CL	S	SC
6 ACTOR sets the <i>beta_min</i> .	1	2	2.50	5
7 ACTOR sets the <i>theta_max</i> .	1	2	2.50	5
8 ACTOR sets the <i>theta_min</i> .	1	2	2.50	5
9 ACTOR sets the <i>delta_max</i> .	1	2	2.50	5
10 ACTOR sets the <i>delta_min</i> .	1	2	2.50	5
11 ACTOR sets the <i>production_sys_addr</i> .	1	1	2.50	2.50
12 ACTOR sets the <i>segregation_sys_addr</i> .	1	1	2.50	2.50
13 ACTOR selects the <i>phase</i> from the dropdown (<i>inference, develop, evaluation</i>).	1	1	2.50	2.50
14 SYSTEM IF the configuration is correct and properly formatted:				
14.1 SYSTEM displays a confirmation message.				
14.2 ELSE (if the configuration is incorrect):				
14.2.1 SYSTEM displays an error message and aborts the process.				
15 ACTOR saves the form.	1	1	2.50	2.50
Human task cost				52.5 < 92.5

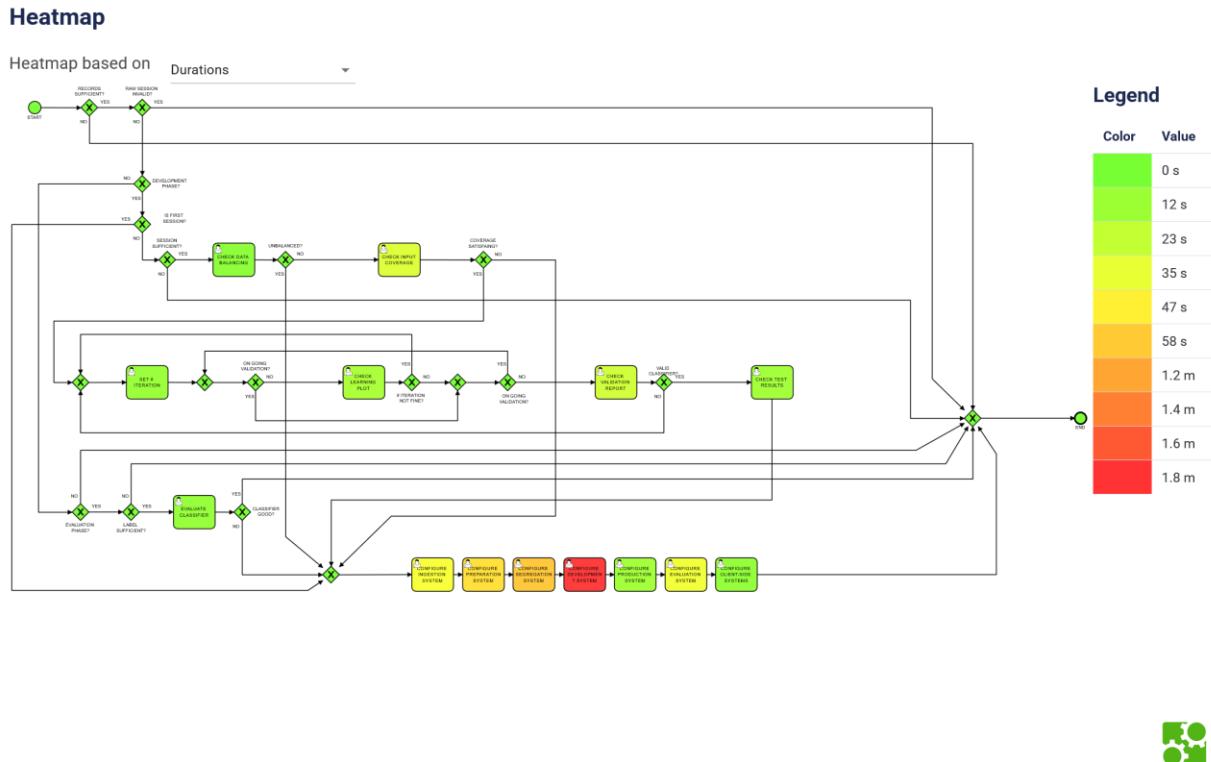
TO-BE Simulation

[Francesco Boldrini, Zahra Omrani]

Heatmap



TO-BE Heatmap of the counts of the parameters



TO-BE Heatmap of the time spent in each passage [Durations]

Scenario Statistics

	Minimum	Maximum	Average
Process instance cycle times including off-timetable hours	0 seconds	13.5 minutes	26.6 seconds
Process instance cycle times excluding off-timetable hours	0 seconds	13.5 minutes	26.6 seconds
Process instance costs	0 EUR	0 EUR	0 EUR

Activity Durations, Costs, Waiting times, Deviations from Thresholds

Name	Waiting time				Duration				Duration over threshold				Cost				Cost over threshold			
	Count	Min	Avg	Max	Min	Avg	Max	Min	Avg	Max	Min	Avg	Max	Min	Avg	Max	Min	Avg	Max	
CHECK DATA BALANCING	474	0 s	0 s	0 s	5.5 s	5.7 s	6 s	0 s	0 s	0 s	0	0	0	0	0	0	0	0	0	
CHECK INPUT COVERAGE	385	0 s	0 s	0 s	29.5 s	31.1 s	32.6 s	0 s	0 s	0 s	0	0	0	0	0	0	0	0	0	
CHECK LEARNING PLOT	409	0 s	0 s	0 s	14.3 s	15 s	15.7 s	0 s	0 s	0 s	0	0	0	0	0	0	0	0	0	
CHECK TEST RESULTS	119	0 s	0 s	0 s	11.9 s	12.5 s	13.1 s	0 s	0 s	0 s	0	0	0	0	0	0	0	0	0	
CHECK VALIDATION REPORT	121	0 s	0 s	0 s	25.8 s	27.1 s	28.4 s	0 s	0 s	0 s	0	0	0	0	0	0	0	0	0	
CONFIGURE CLIENT-SIDE SYSTEMS	482	0 s	0 s	0 s	7.1 s	7.5 s	7.9 s	0 s	0 s	0 s	0	0	0	0	0	0	0	0	0	
CONFIGURE DEVELOPMENT SYSTEM	482	0 s	0 s	0 s	1.7 m	1.8 m	1.8 m	0 s	0 s	0 s	0	0	0	0	0	0	0	0	0	
CONFIGURE EVALUATION SYSTEM	482	0 s	0 s	0 s	35.6 s	37.4 s	39.4 s	0 s	0 s	0 s	0	0	0	0	0	0	0	0	0	
CONFIGURE INGESTION SYSTEM	482	0 s	0 s	0 s	38 s	40 s	42 s	0 s	0 s	0 s	0	0	0	0	0	0	0	0	0	
CONFIGURE PREPARATION SYSTEM	482	0 s	0 s	0 s	49.9 s	52.4 s	55.1 s	0 s	0 s	0 s	0	0	0	0	0	0	0	0	0	
CONFIGURE PRODUCTION SYSTEM	482	0 s	0 s	0 s	11.9 s	12.5 s	13.1 s	0 s	0 s	0 s	0	0	0	0	0	0	0	0	0	
CONFIGURE SEGREGATION SYSTEM	482	0 s	0 s	0 s	57 s	1 m	1 m	0 s	0 s	0 s	0	0	0	0	0	0	0	0	0	
EVALUATE CLASSIFIER	45	0 s	0 s	0 s	7.7 s	8.1 s	8.4 s	0 s	0 s	0 s	0	0	0	0	0	0	0	0	0	
SET # ITERATION	447	0 s	0 s	0 s	9.5 s	10 s	10.5 s	0 s	0 s	0 s	0	0	0	0	0	0	0	0	0	

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TO-BE Simulation Results

Simulation Results

General information

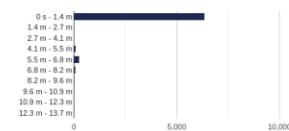
Completed process instances 6852

Total cost 0 EUR

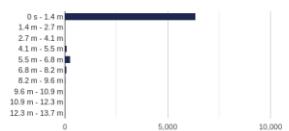
Total simulation time 57 weeks

Charts

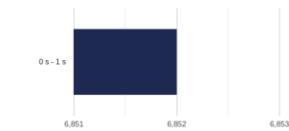
Process cycle times including off-timetable hours



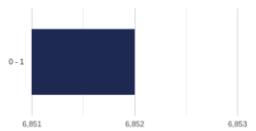
Process cycle times excluding off-timetable hours



Process waiting times



Process costs (EUR)



Resource utilization %



TO-BE Scenario Statistics

Process mining

[*Ettore Ricci, Paolo Palumbo*]

We mined the logs generated by the simulation of the collapsed workflow.

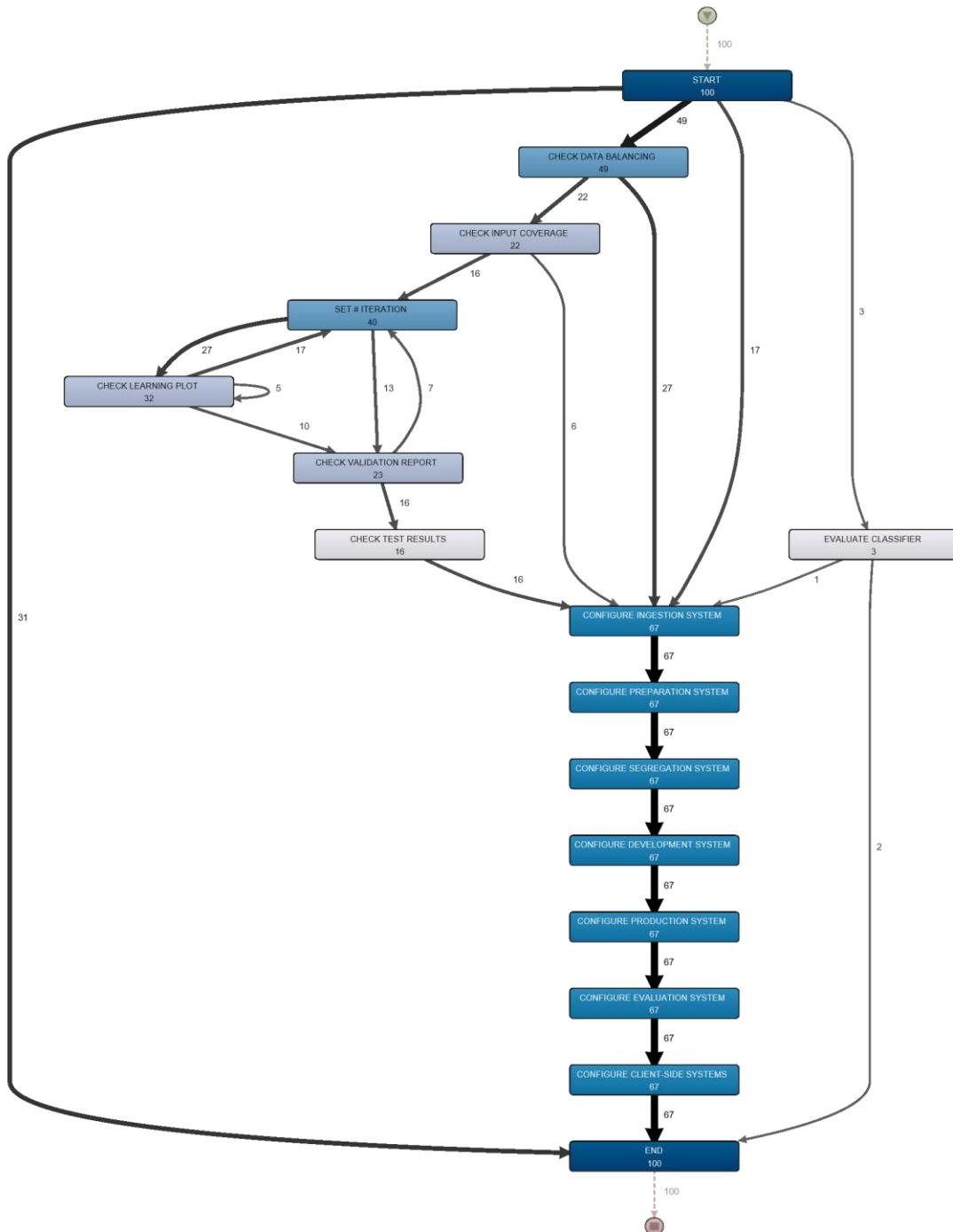
We modified the simulation configuration to make the 100 tokens flow through every path of the workflow. The most important gateways that we changed are listed in the following table.

Gateways configuration

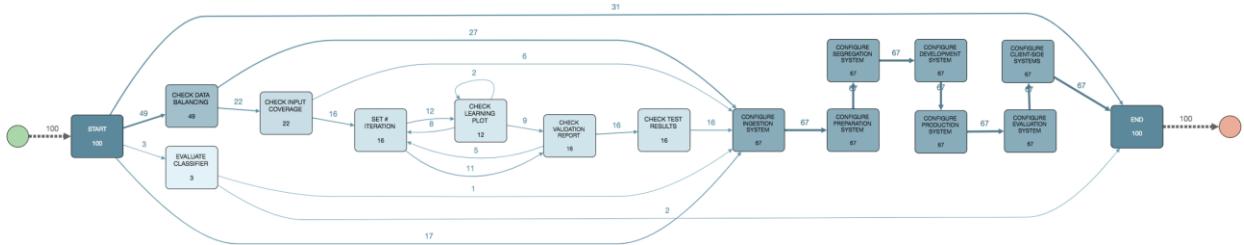
Gateway	Yes	No
RAW SESSION INVALID	5%	95%
RECORD SUFFICIENT	95%	5%
SESSION SUFFICIENT	95%	5%
IS FIRST SESSION	20%	80%
COVERAGE SATISFYING	70%	30%
DEVELOPMENT PHASE	70%	30%

Transaction mining

[Ettore Ricci, Paolo Palumbo]



Disco analysis

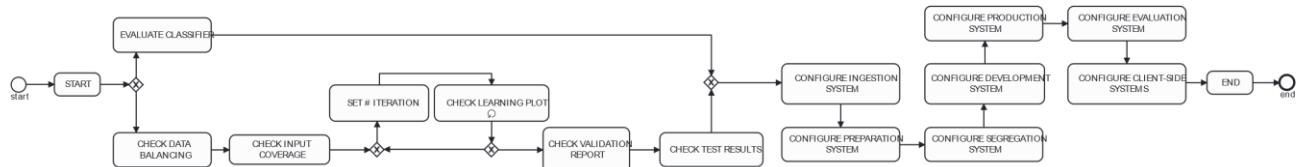


A promore analysis

As we can see, the two transition maps mined from Disco and from Apromore are identical. The only difference stays in the frequencies because in Disco the frequencies are calculated as the total number of times a transition is executed, even on the same token; while in Apromore the frequencies are calculated as the number of individual tokens that execute a transition. This behavior can be changed with a setting in both tools.

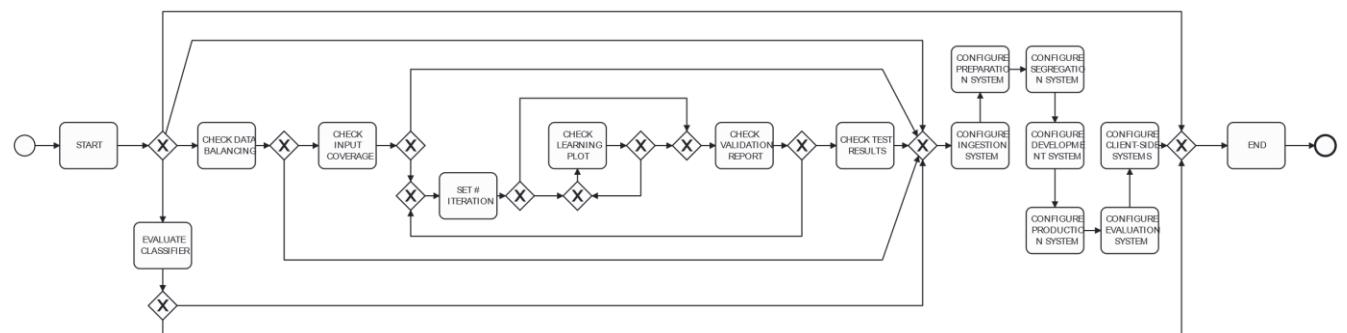
BPMN mining

[*Ettore Ricci, Paolo Palumbo*]



ProM mined BPMN model

We mined the logs using the "Heuristics Miner ProM6" mining algorithm.



Apronmore mined BPMN model

The BPMN model mined from Apromore is more detailed and covers more cases than the one mined from ProM. The key differences between the ProM model and the Apromore one are that the ProM model is missing the paths that skip the training and the configuration as well as one of the two paths

that skip only the training. Furthermore, the training loop is much simpler in the ProM model, as it is missing every path that restarts the training after "CHECK VALIDATION REPORT".

Conformance checking

[Ettore Ricci, Paolo Palumbo]

Comparison of the process mining tools

Tool	Fitness	Generalization	Precision	Simplicity
Apromore	0.9928	0.9837	0.8199	62
ProM	0.7313	0.9902	0.8653	39

Violations

[Ettore Ricci, Paolo Palumbo]

We modified the logs to introduce 3 violations in the workflow. The violations are the following:

1. Skipping the dataset creation ("CHECK DATA BALANCING" and "CHECK INPUT COVERAGE") using data from another user.
2. Skipping "SET # ITERATIONS" and "CHECK LEARNING PLOT" by using early stopping.
3. Skipping "CHECK DATA BALANCING" by using a resampling technique.

Each violation is introduced 3 times in the logs.

These 3 violations can be beneficial in terms of time and resources:

- The first violation can make the costs of the training significantly lower for the client, because using an old dataset allows us to skip the labeling of the new data and it usually is very expensive. Also the manual check of the dataset is skipped saving additional time and resources. It must be noted that this violation can be a problem for the privacy of the clients and also result in worse models if the data of the new user has different characteristics from the old one.
- The second violation can make the training faster, because we do not need anymore to check the learning plot manually and we can train each model only once instead of trying multiple times with different number of iterations. Also, the method previously used to determine the number of iterations was based on an heuristic and it can be prone to errors.
- The third violation can reduce the time and costs of the dataset creation, also making the training possible with unbalanced datasets.

Cases, violations and fitness on models generated by ProM and Apromore

CaseID	Violation	Fitness ProM	Fitness Apromore
10	1	0.91	0.87
20	1	0.85	0.84
47	1	0.86	0.86
53	2	0.91	0.93
63	2	0.84	0.82
88	2	0.91	0.93
6	3	0.91	0.93
72	3	0.91	0.85
81	3	0.94	0.87

10 START	53 START	6 START
10 CHECK DATA BALANCING	53 CHECK DATA BALANCING	6 CHECK DATA BALANCING
10 CHECK INPUT COVERAGE	53 CHECK INPUT COVERAGE	6 CHECK INPUT COVERAGE
10 SET # ITERATION	53 SET # ITERATION	6 SET # ITERATION
10 CHECK LEARNING PLOT	53 CHECK LEARNING PLOT	6 CHECK VALIDATION REPORT
10 CHECK VALIDATION REPORT	53 SET # ITERATION	6 CHECK TEST RESULTS
10 CHECK TEST RESULTS	53 CHECK LEARNING PLOT	6 CONFIGURE INGESTION SYSTEM
10 CONFIGURE INGESTION SYSTEM	53 CHECK VALIDATION REPORT	6 CONFIGURE PREPARATION SYSTEM
10 CONFIGURE PREPARATION SYSTEM	53 CHECK TEST RESULTS	6 CONFIGURE SEGREGATION SYSTEM
10 CONFIGURE SEGREGATION SYSTEM	53 CONFIGURE INGESTION SYSTEM	6 CONFIGURE DEVELOPMENT SYSTEM
10 CONFIGURE DEVELOPMENT SYSTEM	53 CONFIGURE PREPARATION SYSTEM	6 CONFIGURE PRODUCTION SYSTEM
10 CONFIGURE PRODUCTION SYSTEM	53 CONFIGURE SEGREGATION SYSTEM	6 CONFIGURE EVALUATION SYSTEM
10 CONFIGURE EVALUATION SYSTEM	53 CONFIGURE DEVELOPMENT SYSTEM	6 CONFIGURE CLIENT-SIDE SYSTEM
10 CONFIGURE CLIENT-SIDE SYSTEM	53 CONFIGURE PRODUCTION SYSTEM	6 END
10 END	53 CONFIGURE EVALUATION SYSTEM	72 START
20 START	53 CONFIGURE CLIENT-SIDE SYSTEM	72 CHECK DATA BALANCING
20 CHECK DATA BALANCING	53 END	72 CHECK INPUT COVERAGE
20 CHECK INPUT COVERAGE	63 START	72 SET # ITERATION
20 SET # ITERATION	63 CHECK DATA BALANCING	72 CHECK LEARNING PLOT
20 CHECK LEARNING PLOT	63 CHECK INPUT COVERAGE	72 CHECK LEARNING PLOT
20 CHECK VALIDATION REPORT	63 SET # ITERATION	72 SET # ITERATION
20 SET # ITERATION	63 CHECK VALIDATION REPORT	72 CHECK LEARNING PLOT
20 CHECK LEARNING PLOT	63 CHECK LEARNING PLOT	72 SET # ITERATION
20 SET # ITERATION	63 CHECK LEARNING PLOT	72 CHECK LEARNING PLOT
20 CHECK VALIDATION REPORT	63 CHECK LEARNING PLOT	72 SET # ITERATION
20 CHECK TEST RESULTS	63 CHECK LEARNING PLOT	72 CHECK LEARNING PLOT
20 CONFIGURE INGESTION SYSTEM	63 CHECK LEARNING PLOT	72 SET # ITERATION
20 CONFIGURE PREPARATION SYSTEM	63 CHECK LEARNING PLOT	72 CHECK VALIDATION REPORT
20 CONFIGURE SEGREGATION SYSTEM	63 CHECK LEARNING PLOT	72 CHECK TEST RESULTS
20 CONFIGURE DEVELOPMENT SYSTEM	63 CHECK LEARNING PLOT	72 CONFIGURE INGESTION SYSTEM
20 CONFIGURE PRODUCTION SYSTEM	63 CHECK LEARNING PLOT	72 CONFIGURE PREPARATION SYSTEM
20 CONFIGURE EVALUATION SYSTEM	63 CHECK LEARNING PLOT	72 CONFIGURE SEGREGATION SYSTEM
20 CONFIGURE CLIENT-SIDE SYSTEM	63 CHECK LEARNING PLOT	72 CONFIGURE DEVELOPMENT SYSTEM
20 END	63 CHECK LEARNING PLOT	72 CONFIGURE PRODUCTION SYSTEM
47 START	63 CHECK LEARNING PLOT	72 CONFIGURE EVALUATION SYSTEM
47 CHECK DATA BALANCING	63 CHECK LEARNING PLOT	72 CONFIGURE CLIENT-SIDE SYSTEMS
47 CHECK INPUT COVERAGE	63 END	72 END
47 SET # ITERATION	88 START	81 START
47 CHECK VALIDATION REPORT	88 CHECK DATA BALANCING	81 CHECK DATA BALANCING
47 CHECK TEST RESULTS	88 CHECK INPUT COVERAGE	81 CHECK INPUT COVERAGE
47 CONFIGURE INGESTION SYSTEM	88 SET # ITERATION	81 SET # ITERATION
47 CONFIGURE PREPARATION SYSTEM	88 CHECK LEARNING PLOT	81 CHECK LEARNING PLOT
47 CONFIGURE SEGREGATION SYSTEM	88 SET # ITERATION	81 CHECK VALIDATION REPORT
47 CONFIGURE DEVELOPMENT SYSTEM	88 CHECK LEARNING PLOT	81 SET # ITERATION
47 CONFIGURE PRODUCTION SYSTEM	88 SET # ITERATION	81 CHECK LEARNING PLOT
47 CONFIGURE EVALUATION SYSTEM	88 CHECK LEARNING PLOT	81 SET # ITERATION
47 CONFIGURE CLIENT-SIDE SYSTEMS	88 SET # ITERATION	81 CHECK LEARNING PLOT
47 END	88 CHECK VALIDATION REPORT	81 CHECK TEST RESULTS
	88 CHECK TEST RESULTS	81 CONFIGURE INGESTION SYSTEM
	88 CONFIGURE INGESTION SYSTEM	81 CONFIGURE PREPARATION SYSTEM
	88 CONFIGURE PREPARATION SYSTEM	81 CONFIGURE SEGREGATION SYSTEM
	88 CONFIGURE SEGREGATION SYSTEM	81 CONFIGURE DEVELOPMENT SYSTEM
	88 CONFIGURE DEVELOPMENT SYSTEM	81 CONFIGURE PRODUCTION SYSTEM
	88 CONFIGURE PRODUCTION SYSTEM	81 CONFIGURE EVALUATION SYSTEM
	88 CONFIGURE EVALUATION SYSTEM	81 CONFIGURE CLIENT-SIDE SYSTEMS
	88 CONFIGURE CLIENT-SIDE SYSTEMS	81 END
	88 END	

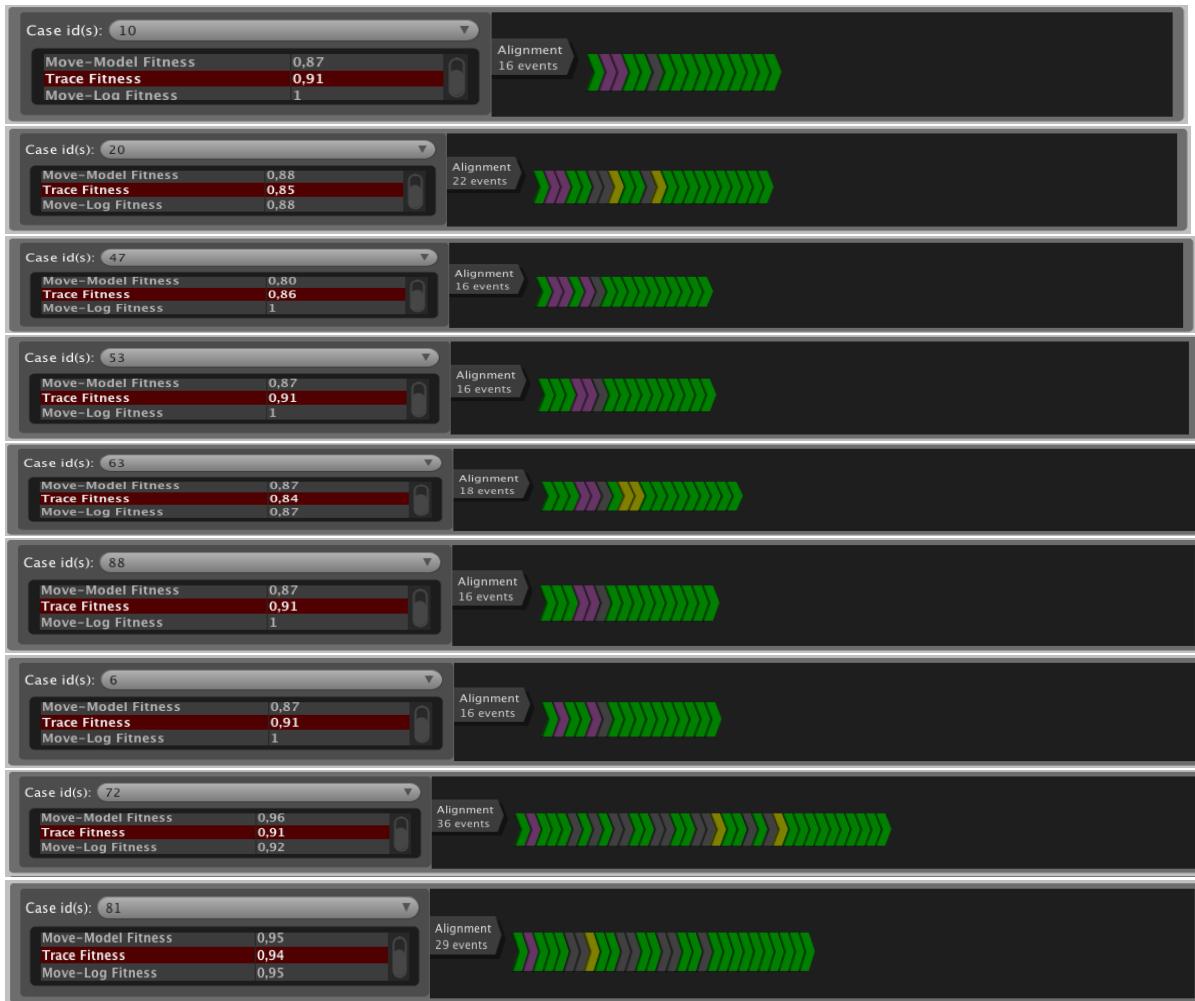
Screenshots of the logs while introducing the violations. The highlighted activities are the ones removed from the logs.

New fitness with violations included in the logs

Tool	Fitness
Apromore	0.9875
ProM	0.7256



Violations in the Apromore model visualized with ProM

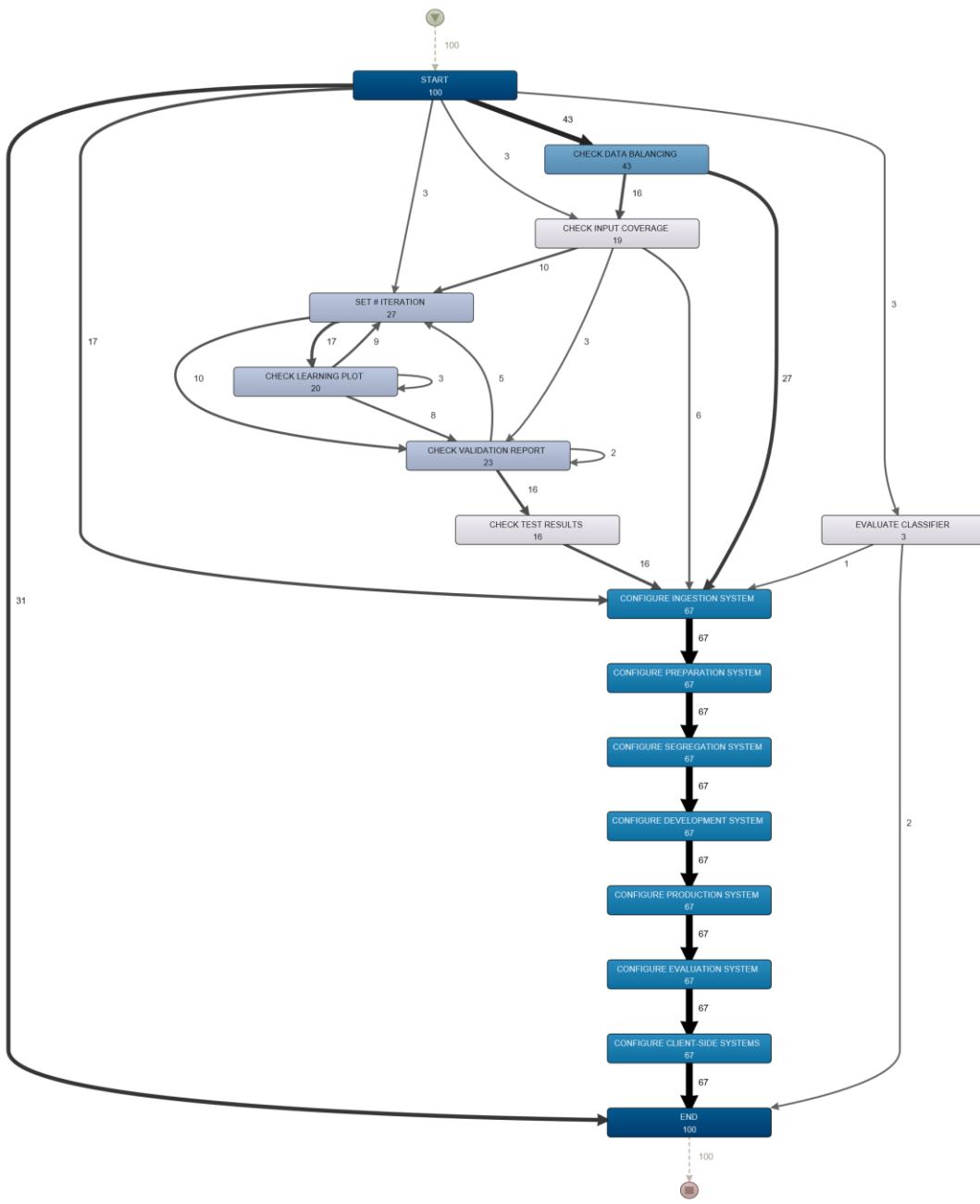


Violations in the ProM model visualized with ProM

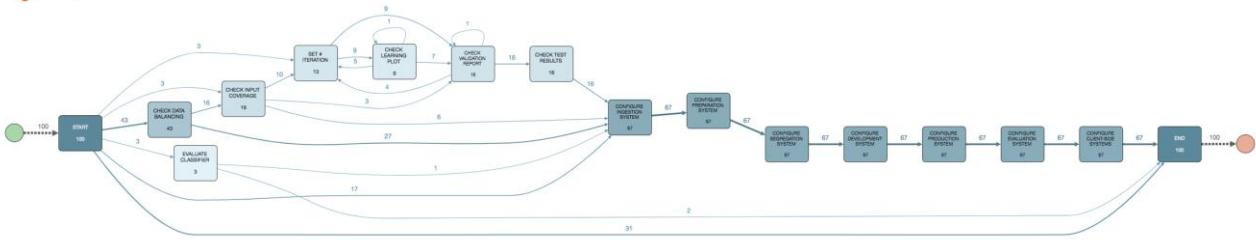
Transaction mining with violations

[*Ettore Ricci, Paolo Palumbo*]

Mining the logs with the violations included, we get these results:



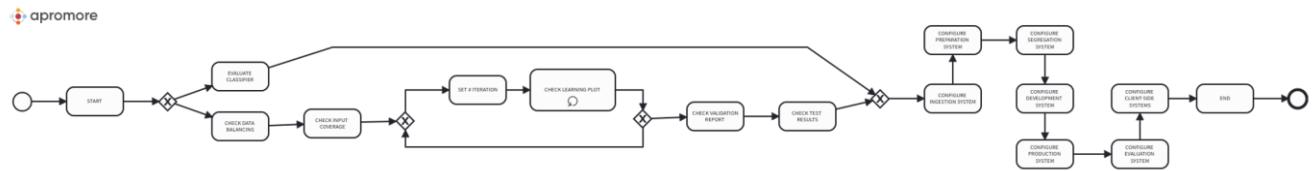
Disco transition map mined with violations



Apromore transition map mined with violations

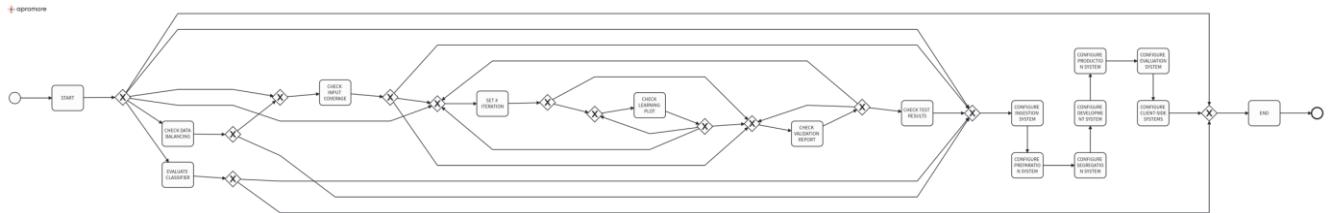
BPMN mining with violations

[*Ettore Ricci, Paolo Palumbo*]



ProM mined BPMN model with violations

As we can see, the BPMN model mined from ProM with the violations included does not change at all from the one without the violations.



Apromore mined BPMN model with violations

On the other hand, the BPMN model mined from Apromore with the violations included changes according to the violations, ultimately having a higher fitness.

Conformance checking with violations

[*Ettore Ricci, Paolo Palumbo*]

Comparison of the process mining tools with violations

Tool	Fitness	Generalization	Precision	Simplicity
Apromore	1	0.9780	0.6742	69
ProM	0.7256	0.9909	0.8941	39

As expected, the fitness of the ProM mined model is the same as the one calculated with the same log, on the old model. Because the ProM model is much simpler than the Apromore one, its Generalization and Precision are higher while the Simplicity is lower. The Apromore model got more complex because of the violations, making its Generalization and Precision lower than the old model, also the Simplicity is a bit higher. The Apromore model, however, has a perfect fitness, because it is able to capture all the possible paths of the workflow, even with the violations.

Because the ProM model did not change, we won't include its cases as they are the same as the ones from the old model.



Violations in the new Apromore model visualized with ProM