

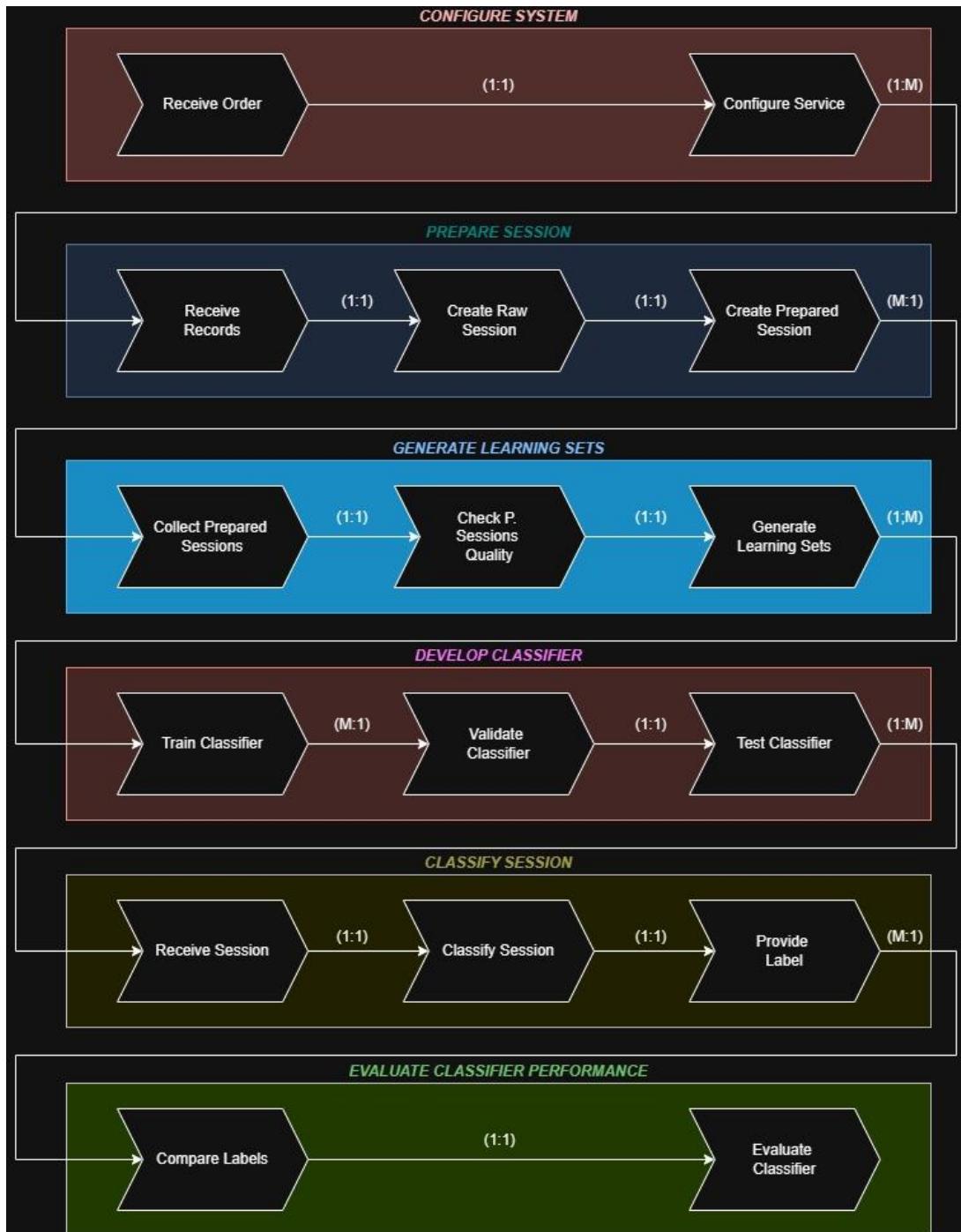
Project Documentation for Process Mining and Intelligence SPEECH EMOTION RECOGNITION

Group Members

Rojan Shrestha
Nimra Tahir
Francesco Panattoni
Ronald Omoding

**University of Pisa
Department of Information Engineering
Academic Year 2025/2026**

1. Process Landscape

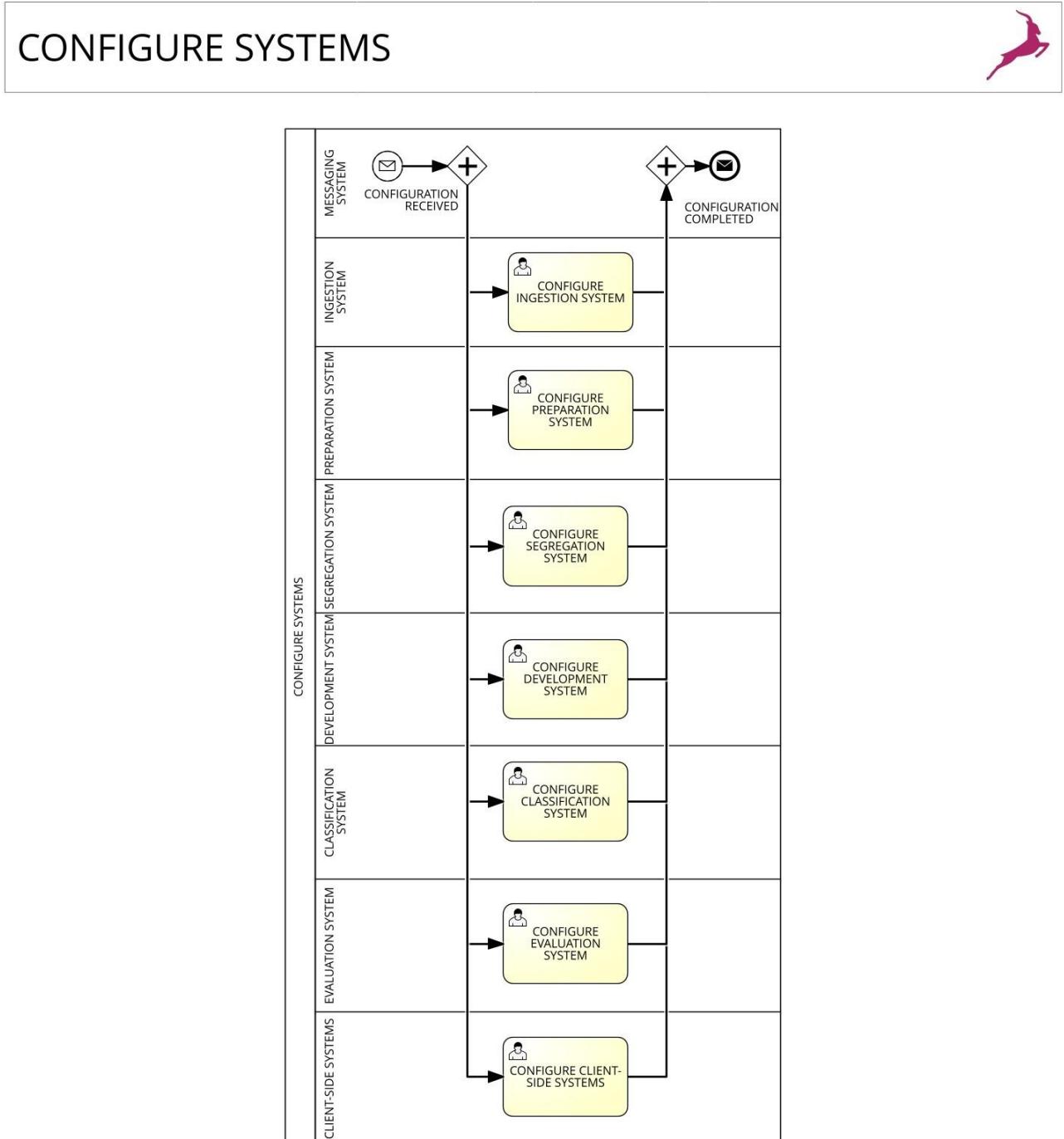


Cardinality Explanation:

1. **Receive Order → Configure Service (1:1)** - Each order received results in exactly one configuration.
2. **Configure Service → Receive Records (1:M)** - A single service configuration enables the system to collect many incoming records from (multiple) client-side sources.
3. **Receive Records → Create Raw Session (1:1)** - Each set of records received results in exactly one raw session. For SER, a raw session includes record sets (Audio Record, Calendar, Profile, and Annotator).
4. **Create Raw Session → Create Prepared Session (1:1)** - One raw session can be transformed into one prepared session (Feature Vector).
5. **Create Prepared Session → Collect Prepared Sessions (M:1)** - Multiple prepared sessions are collected to generate the learning sets.
6. **Collect Prepared Sessions → Check Sessions Quality (1:1)** - Each prepared session collected undergoes a quality check individually.
7. **Check Sessions Quality → Generate Learning Sets (1:1)** - A batch of quality-checked sessions generates a single learning set.
8. **Generate Learning Sets → Train Classifier (1:M)** - A single generated learning set may support several training iterations (epochs) during classifier development to explore many **combinations of hyperparameters** i.e. **Grid Search**.
9. **Train Classifier → Validate Classifier (N:1)** - Multiple training iterations feed into a single validation phase, where all trained variants are assessed through one validation process.
10. **Validate Classifier → Test Classifier (1:1)** - Each validated classifier is tested individually.
11. **Test Classifier → Receive Session (1:M)** - One tested classifier can classify multiple received sessions.
12. **Receive Session → Classify Session (1:1)** - Each session is classified individually.
13. **Classify Session → Provide Label (1:1)** - Each session classification results in exactly one output label.
14. **Provide Label → Compare Labels (M:1)** - Multiple provided labels from different classification sessions are used together in one evaluation step, which compares them to the corresponding human-provided labels.
15. **Compare Labels → Evaluate Classifier (1:1)** - Each evaluation cycle produces one overall classifier evaluation result. Even though multiple label comparisons occur inside the cycle, they collectively generate a single evaluation.

2. Skeleton Factory

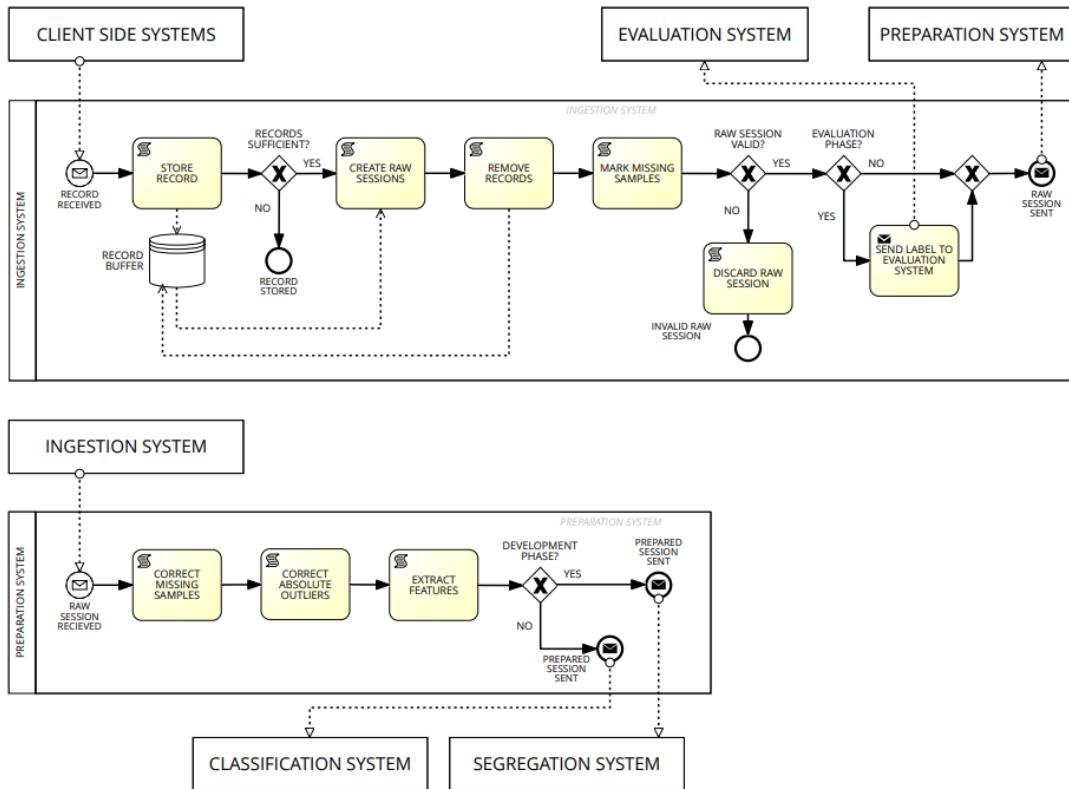
2.1 Configure Systems



This phase defines all global settings for the ML factory - configures data sources, feature extraction rules, hyperparameter ranges, and quality thresholds, ensuring that every downstream process operates with consistent and valid parameters.

2.2 Prepare Session

PREPARE SESSION

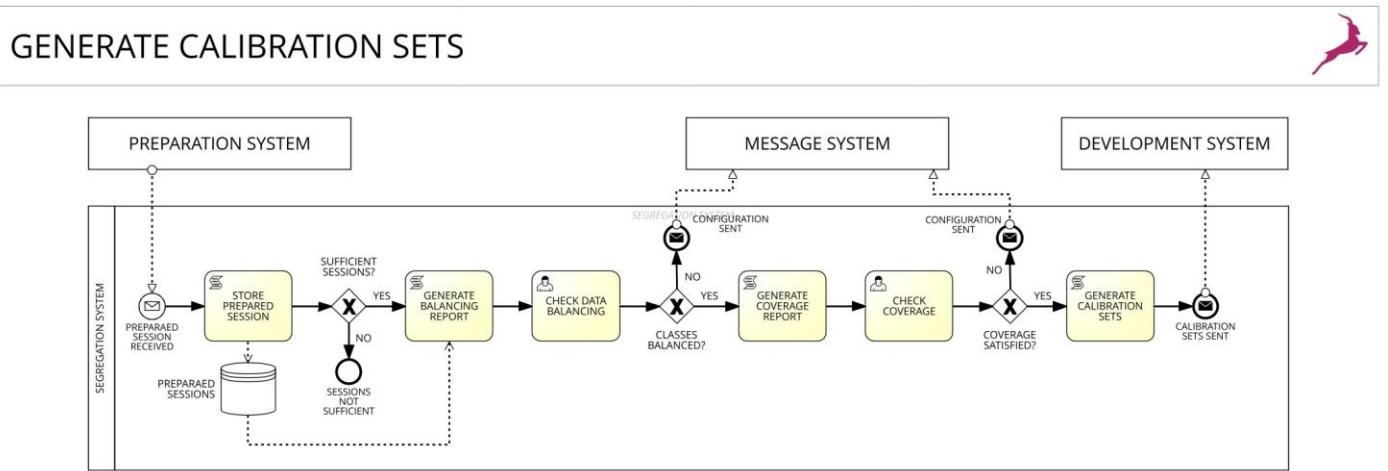


Ingest raw audio data, clean it, and extract features to produce structured sessions ready for model training.

The workflow checks “Evaluation Phase” inside the Ingestion System before sending the raw session onward, while the Preparation System contains a separate “Development Phase” gateway. Why is the phase decision split across two systems instead of being evaluated once and propagated as a single control flow condition?

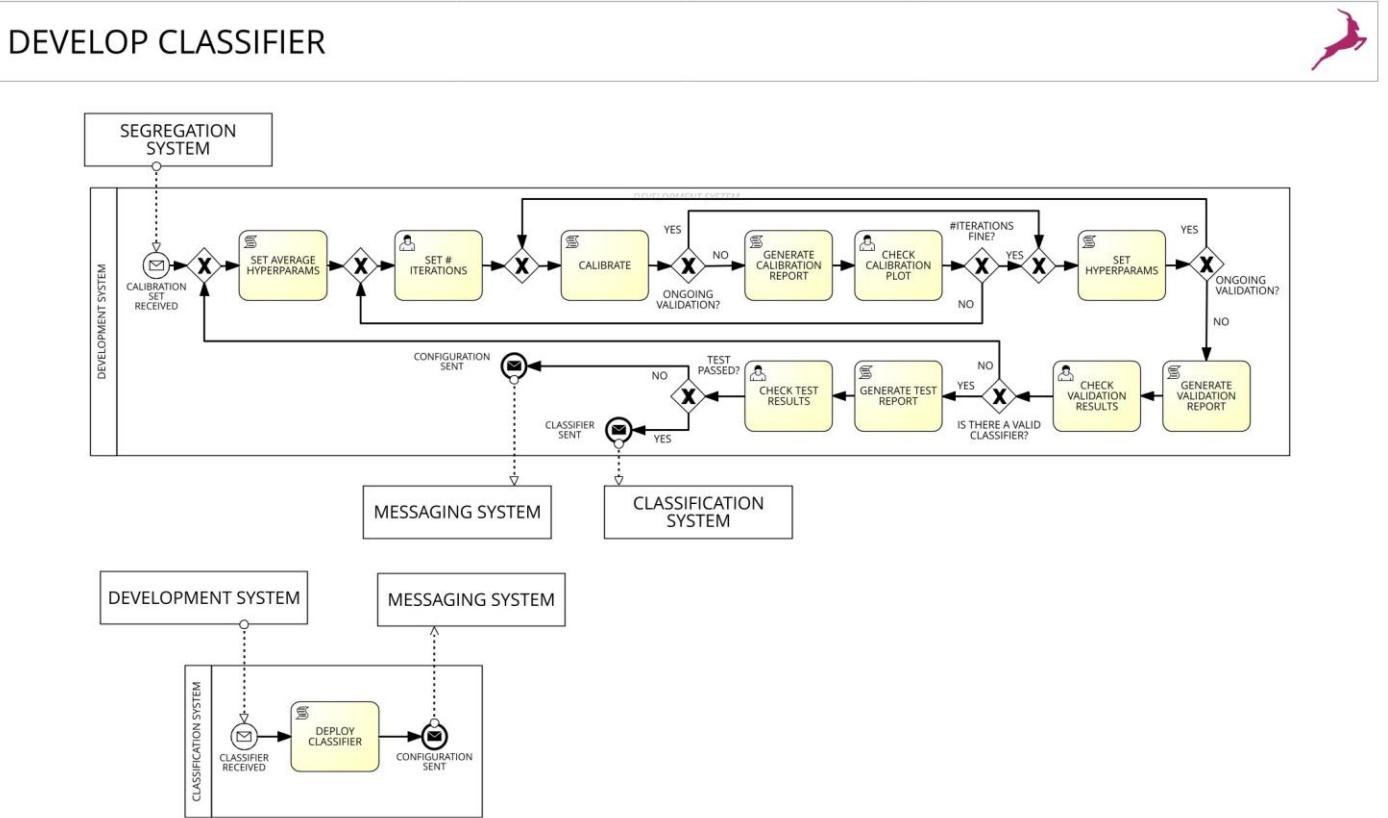
The phase decision is split across the two systems because each system controls a different stage of the lifecycle: Ingestion routes raw sessions for evaluation, while Preparation applies development-specific processing.

2.3 Generate Labeled Sets



Splits prepared sessions into training, validation, and test sets after verifying class balance and input-space coverage. Ensures that the resulting learning sets are sufficient, valid, and representative for reliable classifier training and evaluation.

2.4 Develop Classifier



Trains multiple hyperparameter configurations, validates them to select the best-performing classifier under overfitting constraints, and tests the selected model to confirm generalization. Produces a validated,

production-ready

classifier.

We have three tasks: “Generate training validation report”, “Generate test report” and “Generate validation reports”. What is the exact difference between them and why do we need all three?

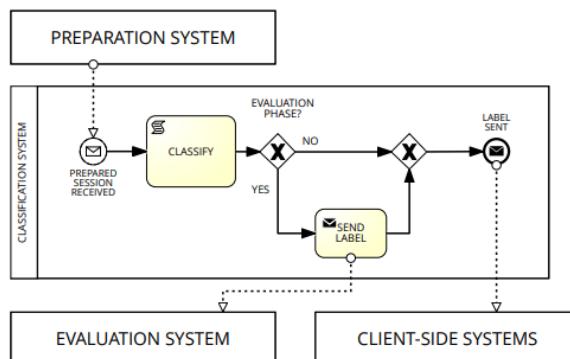
Training report = training-time diagnostics (loss curves, over/underfitting) for data scientists.
Test report = technical metrics on a hold-out set for ML/engineering go/no-go.
Validation report = business-level evaluation (KPIs, A/B, fairness) for product/PO; we should rename them to make this split explicit.

In Develop Classifier, the arrows labelled “ONGOING VALIDATIONS” loop back into the flow around “Train model” and “Set final hyperparameters”. Do these loops mean post-deployment monitoring, or are they just part of the normal development cycle?

During the **Develop Classifier** process, ongoing validations allow the data scientist to iterate retraining and reassessing the model multiple times - without restarting the full workflow. Once performance is confirmed and hyperparameters are finalized, we generate the deployment configuration. From that point onward, continuous monitoring and performance checks are managed within the **Evaluate Classifier Performance** phase.

2.5 Classify Session

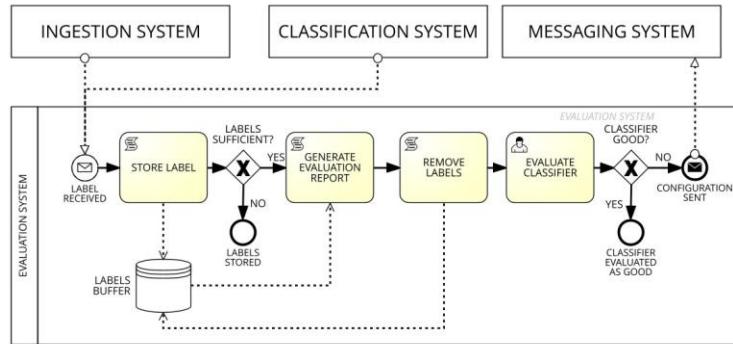
CLASSIFY SESSION



Use the deployed classifier to infer labels for new, unseen sessions in the production environment, providing real-time predictions on operational data.

2.6 Evaluate Classifier Performance

EVALUATE CLASSIFIER PERFORMANCE



Regularly checks how well the classifier is performing by comparing its predictions to the correct labels, making sure its accuracy stays acceptable over time.

What is the actual value of storing the labels? Are we doing it for a specific reason, or is it just a legacy step?

We should check the logs to see if these stored labels are ever actually reused. If they're just being archived and never touched again, it means this step is probably creating overhead without providing any real benefit, and we should reconsider its purpose.

3. Data Modeling

Raw Inputs (client-side systems → records):

1. Audio Sensor → UUID, Audio Samples (numeric amplitude values of the waveform), Sampling Rate and Duration
2. Calendar → UUID, activity: {sport, meditation, work, home, relax}, period of day: {morning, afternoon, evening, night}
3. Profile → UUID, speaker ID, age, gender and language: {Mandarin, English}
4. Annotation → UUID, labeler ID, emotion: {Angry, Happy, Neutral, Sad, Surprise}, confidence (optional) and timestamp (optional)

One session = {Audio Record + Calendar+ Profile+ Annotator}

Output:

5. emotion: {Angry, Happy, Neutral, Sad, Surprise}

Features (net input, derived in the Preparation System - BPMN):

Prosodic features:

- (a) rmse (short-term energy)
- (b) zeroCrossingRate
- (c) chroma_stft}

Spectral features:

- (d) spectralCentroid
- (e) spectralBandwidth
- (f) SpectralRolloff

MFCC features (20-dimensional):

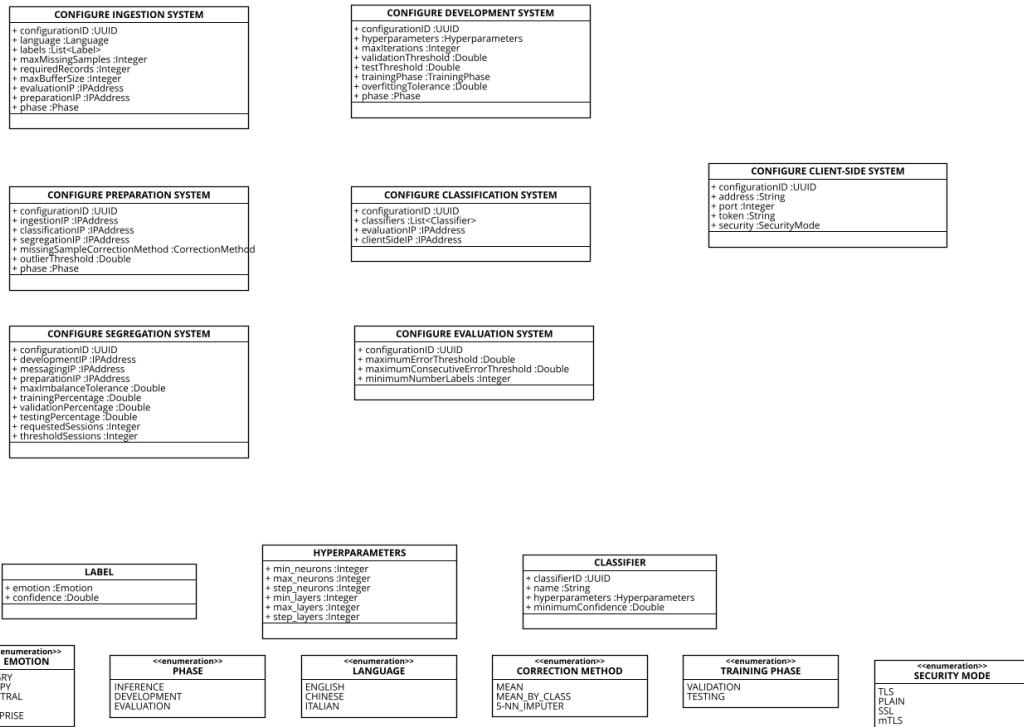
- (g) mfcc1 – mfcc20

NOTE:

1. Audio Sample = 1 numeric amplitude value at a time step
2. Sample is the smallest atomic unit of input data

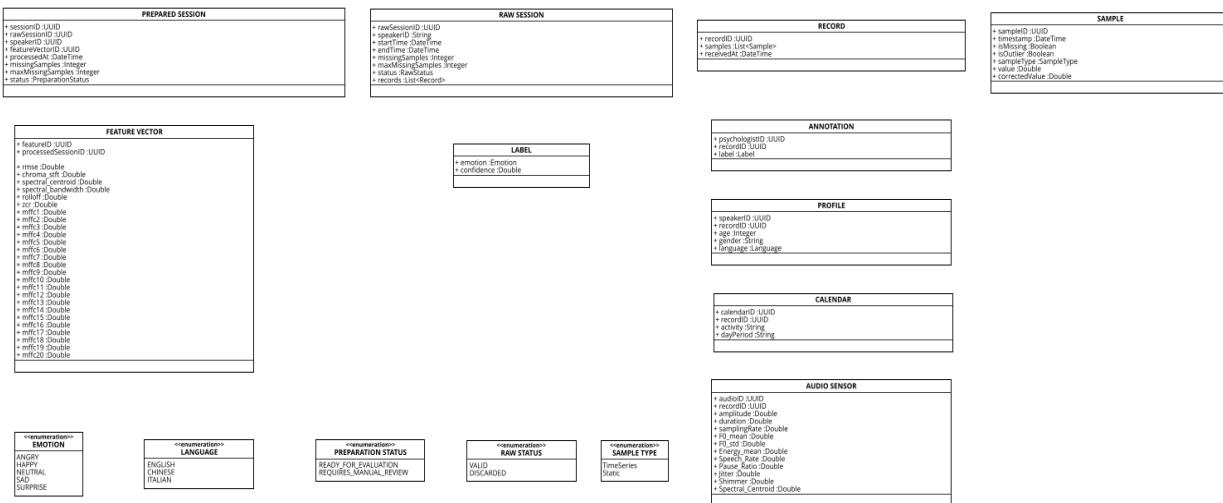
3.1 DATA MODEL of CONFIGURE SYSTEM (Francesco)

Data Model of Configure Systems



3.2 DATA MODEL of PREPARE SESSION (Francesco)

Data Model of Prepare Session

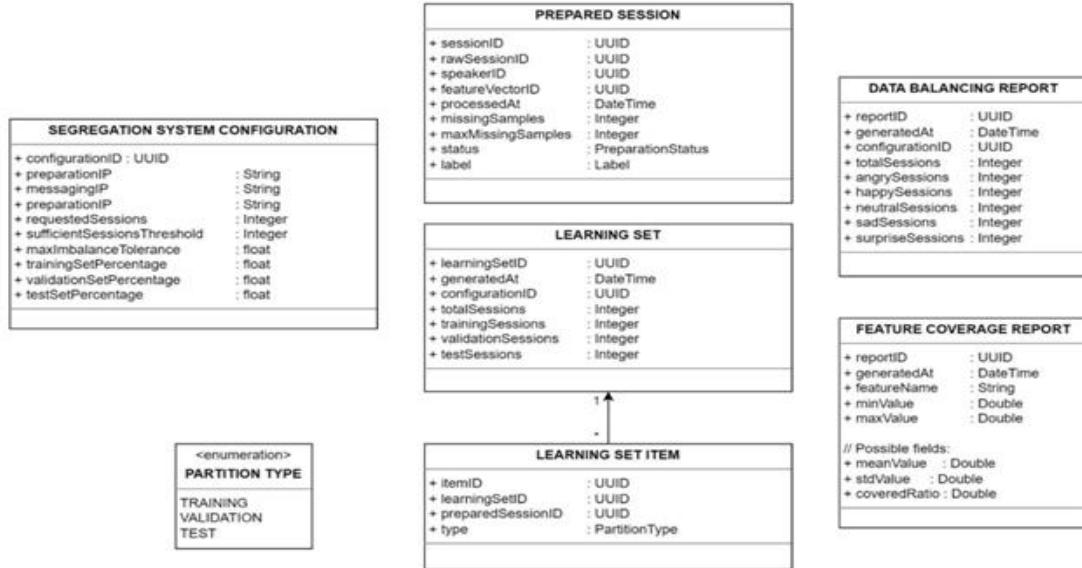


Feedback:

1. Annotation, Profile, Callender and Audio Sensor, all are Record but only Audio Sensor can have deviceId so deviceId as a field in Record is unclear
2. Samples usually refer to timeseries. If the values of Annotation,Profile, Calender are considered static samples and that they have a missing value, it means they affect the validity of the raw session. Since the only purpose of the raw session is to create a prepared session with feature vectors. Referring to **TASK Check Input Coverage**, the features are only dependent upon the audio samples, so it is unclear why the values of Annotation, Profile, Callender are considered samples.
3. Mffc1-mffc20 can be grouped together as a single field - list (Mffc)
4. Why Record and Annotation both have Label fields? "label: Label" only makes sense if Label is an enum but Label is a class per the model and should be referenced with UUID
5. Maybe label a separate field in Prepared Session?
6. PreparedSession.status can be "Evaluation: Boolean" same with Raw Status

3.3 DATA MODEL of GENERATE LEARNING SETS (Rojan)

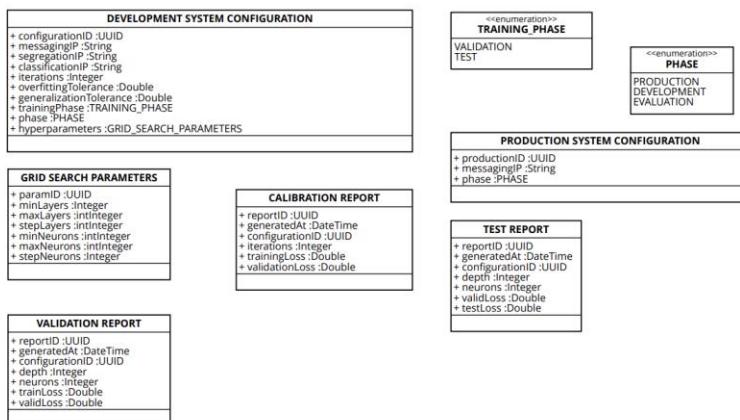
Data Model of Generate Learning Sets



1. Why configurationID ? To say “This report was generated using configuration X”
2. // Possible fields: FEATURE COVERAGE REPORT : + meanValue: Double, + stdValue: Double, + coveredRatio: Double
3. Do we need trainingSessions, validation Sessions, and test Sessions count?
4. Should LABEL be a separate entity? (OR)

3.4 DATA MODEL of DEVELOP CLASSIFIER (Nimra)

DataModel Of DEVELOP CLASSIFIER



- **DEVELOPMENT SYSTEM CONFIGURATION** is correctly modeled as the core object of this task: it holds the IPs of the connected systems, the iteration budget, and the tolerance thresholds. All reports (CALIBRATION REPORT, VALIDATION REPORT, TEST REPORT) are linked back to it via configurationID, so every result can be traced to the exact configuration used during that run.
- Splitting results into three report classes matches the factory: calibration, validation and test are distinct phases with different purposes.
 - Calibration focuses on iterations and the pair (trainingLoss, validationLoss) over time
 - Validation report stores grid-search results (depth, neurons, losses per configuration)
 - Test report captures the final chosen configuration with a single pair of validLoss and testLoss.
- In **CALIBRATION REPORT** and **VALIDATION REPORT**, losses and hyperparameter values are modeled as arrays [0..*]. This reflects the fact that one *calibration/validation run produces a full learning curve over many iterations*, not just a **single scalar**, and is aligned with how the “Check Calibration Plot” task uses these curves.
- **GRID SEARCH PARAMETERS** collects the search space (minLayers, maxLayers, stepLayers, minNeurons, maxNeurons, stepNeurons) instead of mixing these with live configuration fields. This keeps the development configuration focused on “what was actually run” while allowing the same parameter grid to be reused for multiple experiments if needed.
- Enumerations (TRAINING_PHASE, PHASE) ensure that both internal training status (validation/test) and global lifecycle status (development/production) remain explicit and valid.
- **PRODUCTION SYSTEM CONFIGURATION** only stores what is strictly needed for deployment (productionID, messagingIP, phase), which is coherent with the Deploy Classifier part of the factory. It reuses the same PHASE enum, so the production configuration can be interpreted together with the development configuration in dashboards or logs.
- All report entities use reportID : UUID and generatedAt : DateTime, while configurations use configurationID / productionID. This ensures that every training/validation/test run can be audited later (e.g., “which configuration generated this validation report on this date?”).
- Every metric can be traced back to the exact configuration and hyperparameters, supporting reproducibility, debugging, and regulatory transparency.

3.5 DATA MODEL of CLASSIFY SESSION (Ronald)

Data Model of Classify Session

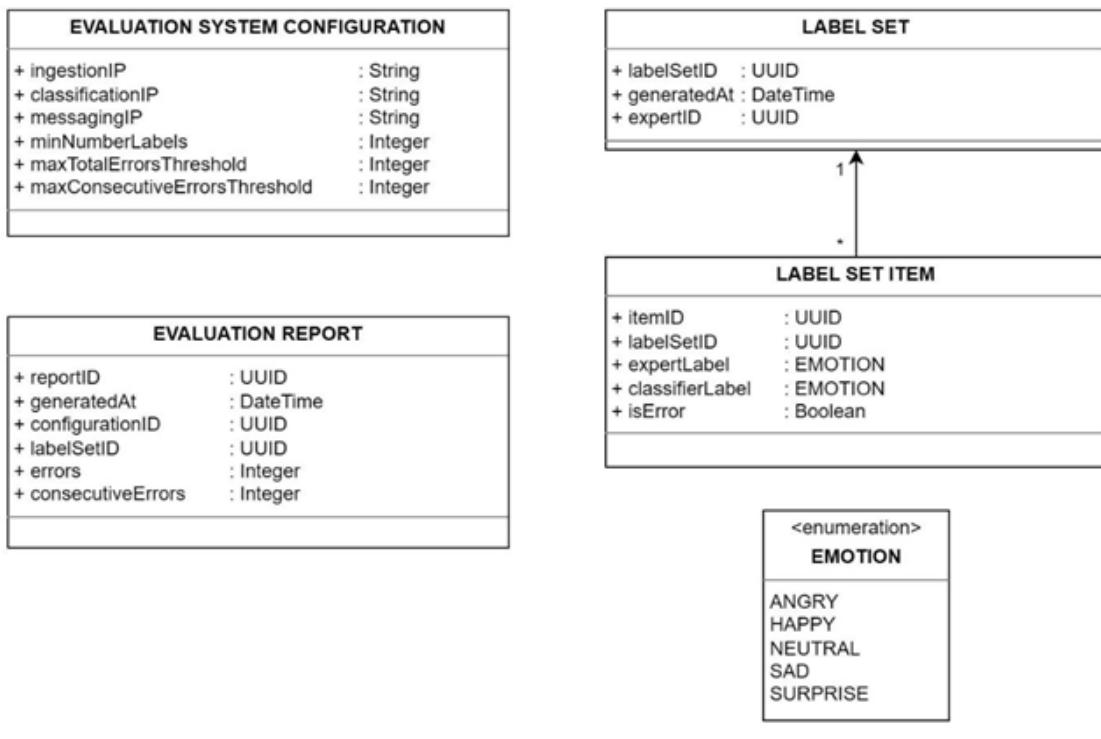


PREPARED SESSION	LABEL
+ sessionID :UUID + rawSessionID :UUID + speakerID :UUID + featureVectorID :UUID + processedAt :DateTime + missingSamples :Integer + maxMissingSamples :Integer + status :PreparationStatus	+ emotion :EMOTION + confidence :Double

EMOTION
<<enumeration>>
ANGRY
HAPPY
NEUTRAL
SAD
SURPRISE

3.6 DATA MODEL of EVALUATE CLASSIFIER PERFORMANCE (Rojan)

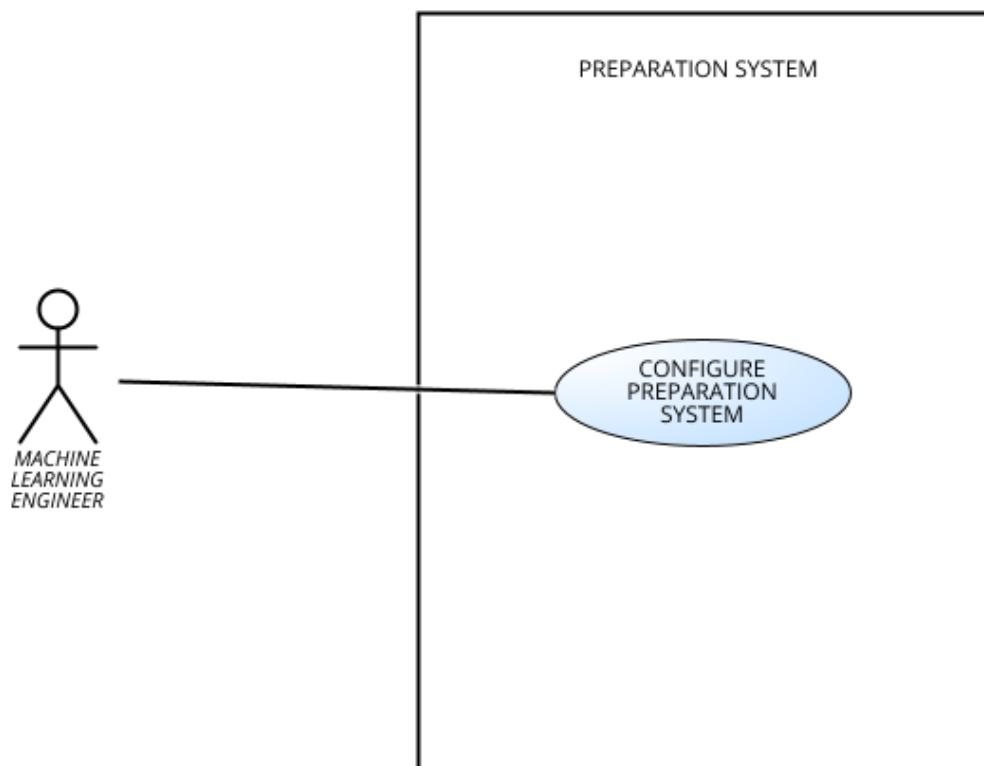
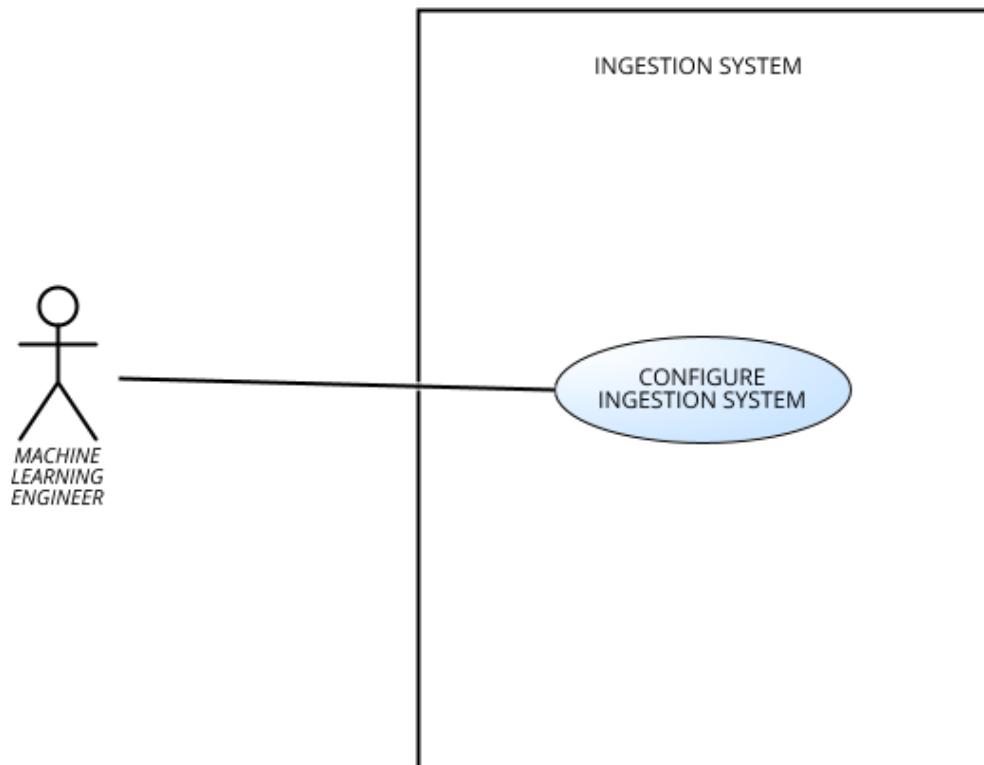
Data Model of Evaluate Classifier Performance

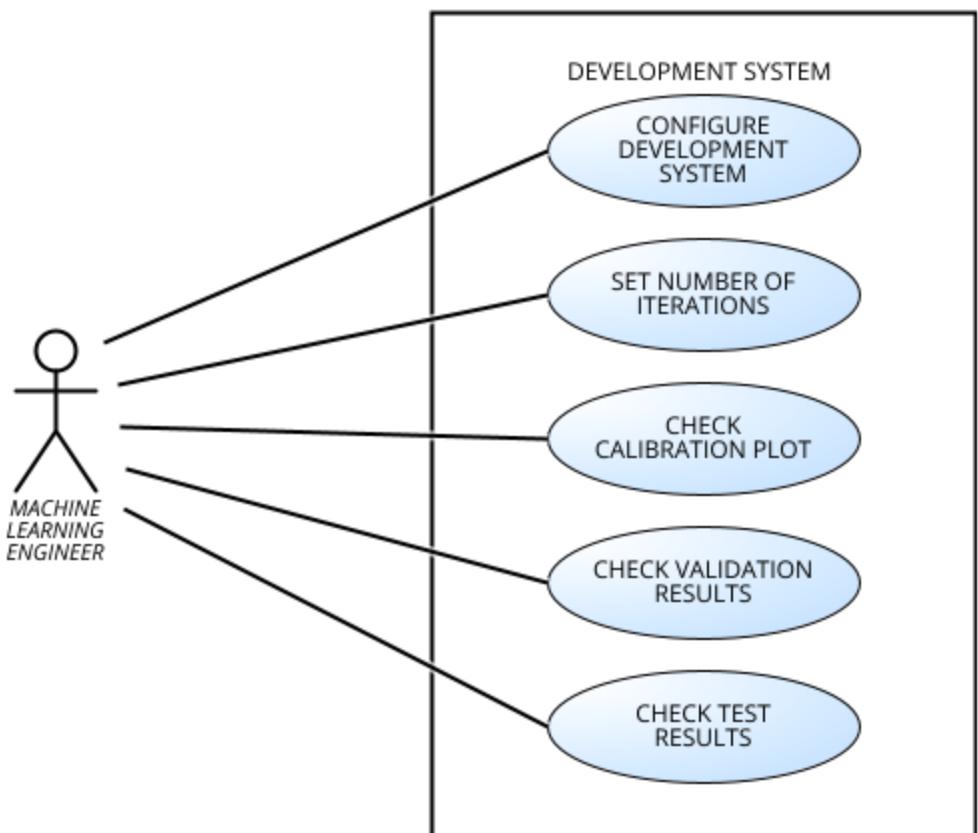
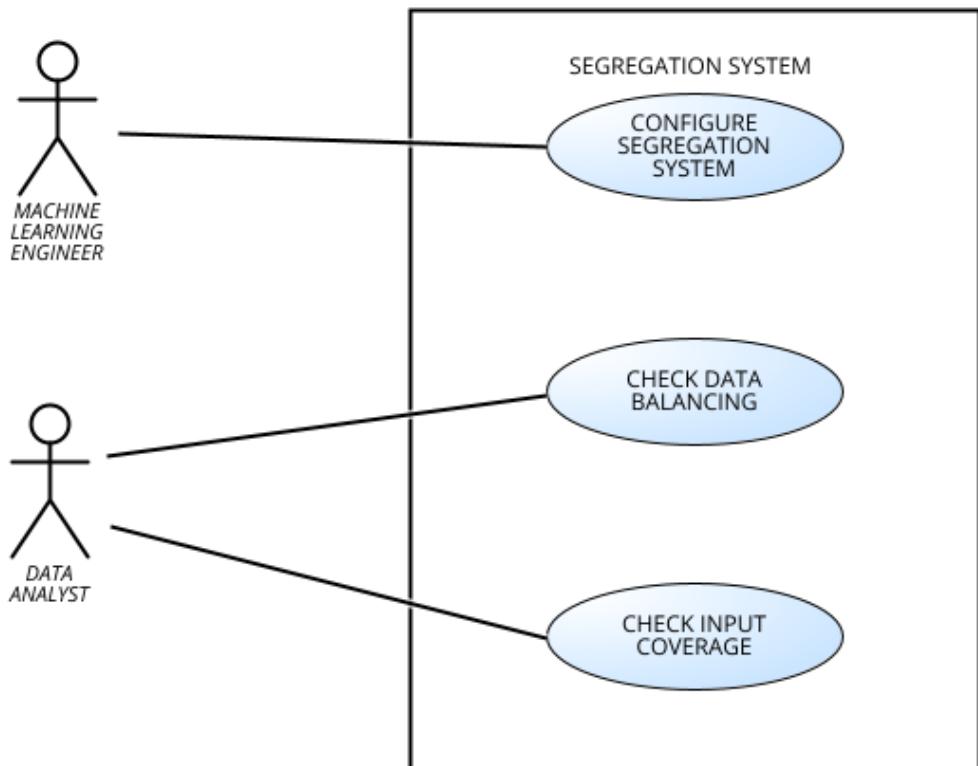


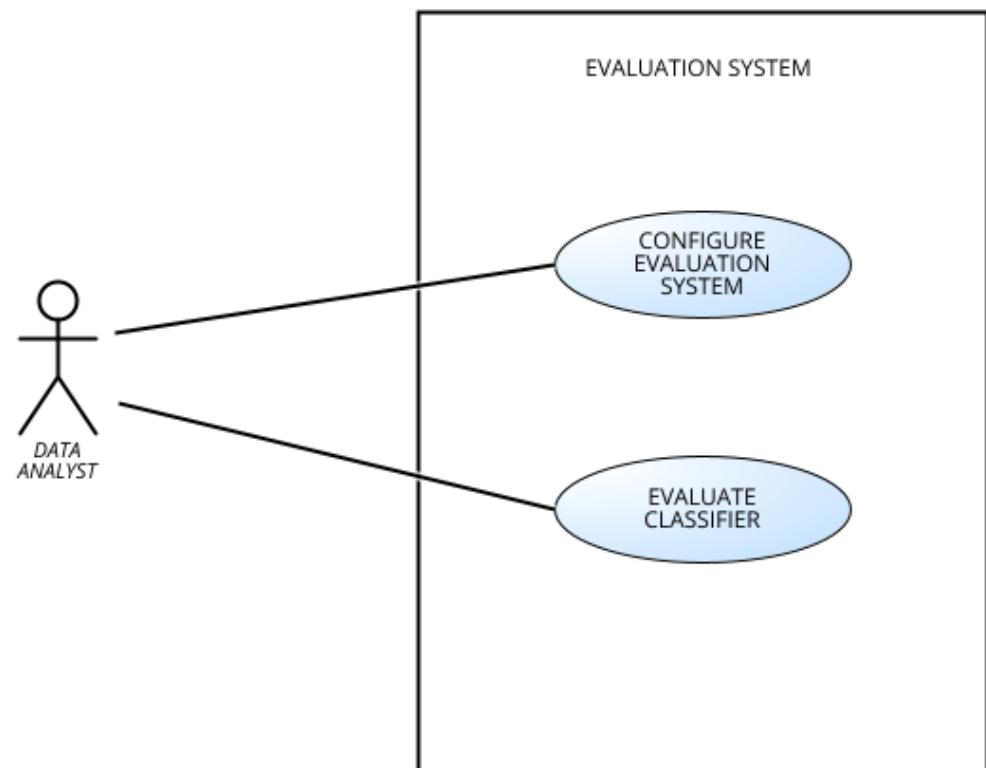
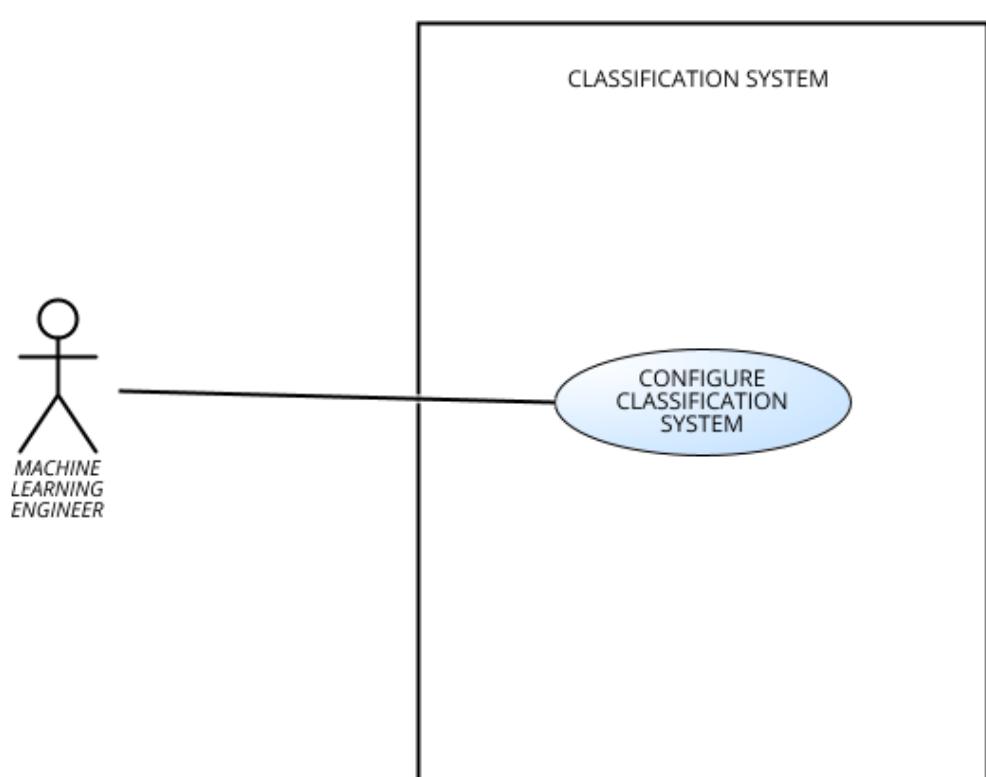
- Prof says “For each Process” so Configure Service would have all configuration models but since they are used in other process as well should we include each configuration model also in respective process? Ex: Segregation config model in both Configure Service and GENERATE LEARNING SETS?

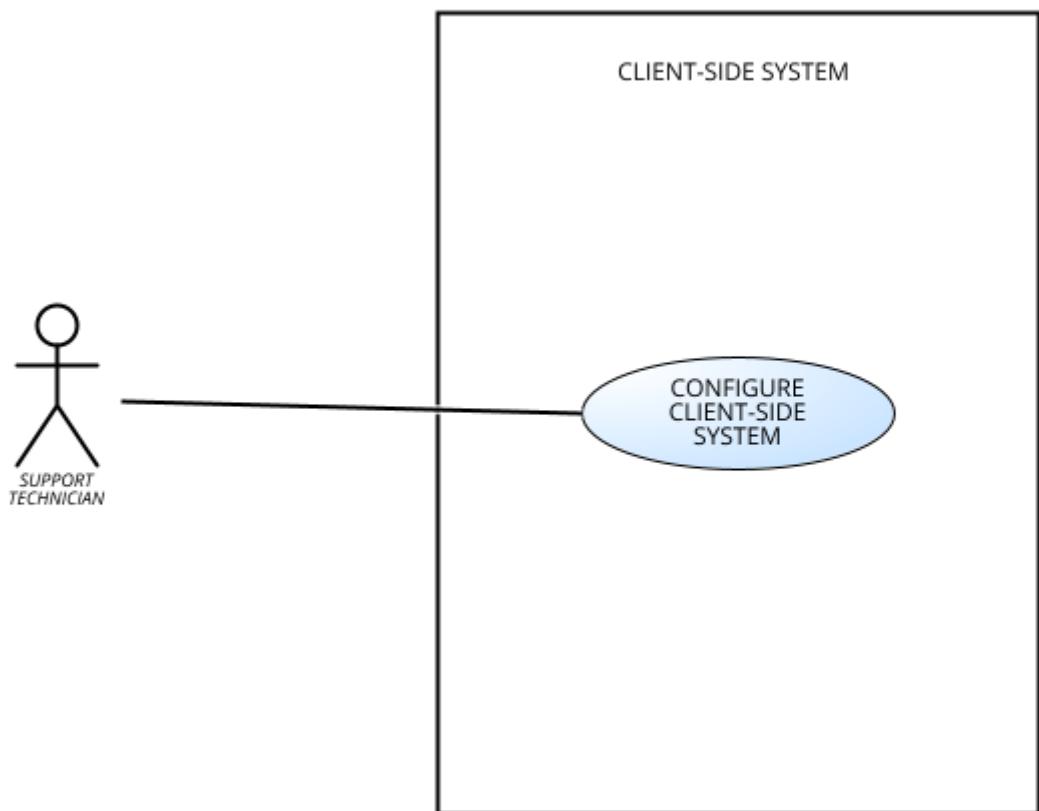
4. Task Modeling

4.1 Use Cases









4.2 Normalized Salaries

ACTOR	Description	LINK
Data Analyst	Collects, organizes, and cleans data; analyzes data for patterns/trends; builds databases; creates visualizations; and communicates insights.	https://www.payscale.com/research/IT/Job=Data_Analyst/Salary
Machine Learning Engineer	Builds ML systems, implements ML algorithms, analyzes data statistically, runs experiments/tests, maintains and improves ML systems, and bridges data scientists with engineering teams.	https://www.payscale.com/research/IT/Job=Machine_Learning_Engineer/Salary
Domain Expert (Psychologist)	Examines audio recordings, segmenting them into distinct utterances when needed, and assigns the appropriate emotional label to each segment. Specializes in interpreting vocal cues and emotional nuances to ensure accurate ground-truth annotations.	https://www.payscale.com/research/IT/Job=Psychologist/Salary
Support Technician	Install, configure and maintain hardware/software and network for clients/devices. Fix equipment issues, perform upgrades, set up new devices, assist	https://www.payscale.com/research/IT/Job=Support_Technician%2C_Information_Technology_(IT)_Support

	users, and manage system inventory.	
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Normalized salaries are calculated using the following method:

1. Determine the yearly salary;
2. Find the lowest salary (System Engineer);
3. Divide this amount by the lowest salary (used as the baseline).

4.3 Task

$$\text{STEP COST} = \text{OCCURENCE} \times \text{COGNITIVE} \times \text{SALARY}$$

OCCURENCE is the number of executions of the step.

COGNITIVE is the cognitive effort. It can have these values:

- **Remember (1)**: Perform the step by recalling a previous instance of the same action.
- **Understand (2)**: Perform the step by identifying a value within a set of predefined categories.
- **Apply (3)**: Perform the step by following a company-defined procedure.
- **Analyze (4)**: Perform the step by determining unknown categories.

SALARY is the normalized job salary (calculated in previous page).

4.3.1. Configure the Ingestion System (Nimra)

	STEP	O	CE	O × CE
1	USER opens the “Ingestion System Configuration form”	1	1	1
2	SYSTEM loads the configuration interface with default / current parameters			
3	USER reviews the raw-input sources configuration (Audio Sensor, Calendar, Profile, Annotation)	1	2	2
3.1	USER edits/confirms the Audio Sensor fields (UUID, sampling rate, max duration, waveform format, etc...)	1	3	3
3.2	USER edits/confirms the Calendar fields (activity set, period of day, mandatory flag)	1	3	3
3.3	USER edits/confirms the Profile fields (speaker ID, age, gender, language set)	1	3	3
3.4	USER edits/confirms the Annotation fields (labeler ID, emotion set, confidence/timestamp options)	1	3	3
4	USER reviews the minimum number of Records	1	2	2
4.1	USER edits/confirms the minimum number of Records	1	3	3
5	USER reviews the max missing Samples	1	2	2
5.1	USER edits/confirms the max missing Samples	1	3	3
6	USER reviews the IP Addresses	1	2	2
6.1	USER set Evaluation IP Address	1	3	3
6.2	USER set Preparation IP Address	1	3	3
7	USER selects <i>Save to store the configuration</i>	1	1	1
8	SYSTEM validates the configuration and shows a confirmation message			
9	USER closes the configuration interface	1	1	1
TOTAL (O × CE) COST		35		
USER is a Machine Learning Engineer: Normalized Salary		× 1.67		
TOTAL HUMAN COST		58.45		

INGESTION SYSTEM CONFIGURATION

MLOps Dashboard – Speech Emotion Recognition · User: System Engineer EXIT

RAW INPUT SOURCES CONFIGURATION

Audio Sensor fields	Sampling rate (Hz)
UUID format	16000
Standard UUID v4	
Maximum duration (seconds)	Waveform format
10.0	PCM 16-bit mono
Calendar fields	Period of day set
Activity set	{morning, afternoon, evening, night}
{sport, meditation, work, home, relax}	
Calendar record mandatory?	
Yes – reject sessions without Calendar	
Profile fields	Language set
Required attributes	{Mandarin, English}
Speaker ID, Age, Gender	
Annotation fields	Confidence field
Emotion label set	Optional
{Angry, Happy, Neutral, Sad, Surprise}	
Timestamp field	
Optional	

SESSION DEFINITION & VALIDATION

Session composition	Missing components policy
One session = Audio + Calendar + Profile + Annotation	Discard session and log warning
Logical rule used to aggregate records into one session.	
Invalid-session threshold (%)	Metadata inconsistency rule
10	Discard & log
Max allowed share of missing / inconsistent fields.	

BUFFER & STORAGE PARAMETERS

Buffer location	Max buffer size (GB)
s3://speech-emo-prod/raw-buffer	512
Raw session retention (days)	Archiving policy
7	Move to cold storage after retention

Cancel Save configuration

FEEDBACK: These are the parameters we think should be in the mockup considering they are used in Ingestion system (refer to gateways in skeleton factory)

1. required_records* - Minimum number of records needed before creating a raw session.
2. max_missing_samples*- Upper bound on allowable missing samples in a raw session. If exceeded, the raw session is discarded.
3. evaluation_sys_addr* -Address/endpoint of the Evaluation System for sending labels during evaluation phase.
4. preparation_sys_addr* + client sys addr - Address/endpoint of the Preparation System to which raw sessions are forwarded.
5. phase - Indicates the active lifecycle phase: development, production, evaluation

Remove Session Definition & Validation and Buffer & Storage Parameters.

YOU DON'T HAVE TO REDO THE COST TABLE. WE ALREADY UPDATED.

4.3.2. Configure the Preparation System (Ronald)

	STEP	O	CE	$O \times CE$
1	USER opens the “Preparation System Configuration” form	1	1	1
2	SYSTEM loads the configuration interface with default parameters			
3	USER reviews the connected system IP addresses	1	2	2
3.1	USER edits/confirms Ingestion System IP address	1	3	3
3.2	USER edits/confirms Classification System IP address	1	3	3
3.3	USER edits/confirms Segregation System IP address	1	3	3
4	USER reviews the preprocessing parameters	1	2	2
4.1	USER opens Missing Samples Correction Method field	1	1	1
4.2	USER edits/confirms the Missing Samples Correction Method	1	3	3
4.3	USER edits/confirms the minimum outlier threshold	1	3	3
4.4	USER edits/confirms the maximum outlier threshold	1	3	3
4.5	USER opens Feature Extraction Method field	1	1	1
4.6	USER edits/confirms the Feature Extraction Method	1	3	3
5	USER opens Phase Selection field	1	1	1
5.1	USER selects Phase (development / production / evaluation)	1	3	3
6	USER selects <i>Save Configuration</i> to store the configuration	1	1	1
7	SYSTEM validates the configuration and shows a confirmation message			
8	USER closes the configuration interface	1	1	1
TOTAL (O × CE) COST		34		
USER is a Machine Learning Engineer: Normalized Salary		× 1.67		
TOTAL HUMAN COST		56.78		

Preparation System Configuration

System Addresses

Ingestion System IP Address

192.168.1.101

Classification System IP Address

192.168.1.102

Segregation System IP Address

192.168.1.103

Phase Selection

Development

Production

Evaluation

Preprocessing Parameters

Missing Samples Correction Method

Drop Missing

Absolute Outlier Detection Threshold

Minimum

-2.5

Maximum

2.5

Feature Extraction Method

Principal Component Analysis (PCA)

Close

Save Configuration

4.3.3. Configure the Segregation System (Rojan)

	STEP	O	CE	O × CE
1	USER opens the “ <i>Segregation System Configuration</i> ” form	1	1	1
2	SYSTEM loads the configuration interface with default parameters			
3	FOR EACH input fields of the System Addresses (IP address of Development, Messaging and Preparation System)	3		
3.1	USER edits/confirms the IP address		3	9
4	FOR EACH input fields of the Session Sufficiency (Number of Requested Sessions and Sufficient Sessions Threshold)	2		
4.1	USER edits/confirms the parameter		3	6
5	USER edits/confirms the Class Balancing Tolerance	1	3	3
6	FOR EACH input fields of the learning set split percentages (Training, Validation and Test)	3		
6.1	USER edits/confirms the percentages (%)		3	9
7	USER selects Save to store the configuration	1	1	1
8	SYSTEM validates the configuration and shows a confirmation message			
9	USER closes the configuration interface	1	1	1
TOTAL (O × CE) COST		30		
USER is a Machine Learning Engineer: Normalized Salary		× 1.67		
TOTAL HUMAN COST		50.1		

SEGREGATION SYSTEM CONFIGURATION

[EXIT](#)

SYSTEM ADDRESSES

Development System IP

Messaging System IP

Preparation System IP

SESSION SUFFICIENCY

Number of Requested Sessions

Sufficient Sessions Threshold

BALANCING PARAMETERS

Max Class Imbalance Tolerance (%)

LEARNING SET SPLIT PERCENTAGES

Training Set (%)

Validation Set (%)

Test Set (%)

[SAVE](#)

4.3.4. Configure the Development System (Ronald)

	STEP	O	CE	O × CE
1	USER opens the “Development System Configuration” form	1	1	1
2	SYSTEM loads the configuration interface with default parameters			
3	USER reviews the connected system IP addresses	1	2	2
3.1	USER edits/confirms Segregation System IP address	1	3	3
3.2	USER edits/confirms the Messaging System IP address	1	3	3
3.3	USER edits/confirms Classification System IP address	1	3	3
4	USER reviews the hyperparameter settings	1	2	2
4.1	USER opens Hyperparameter Grid field	1	1	1
4.2	USER edits/confirms <i>min_layers</i>	1	3	3
4.3	USER edits/confirms <i>max_layers</i>	1	3	3
4.4	USER edits/confirms <i>step_layers</i>	1	3	3
4.5	USER edits/confirms <i>min_neurons</i>	1	3	3
4.6	USER edits/confirms <i>max_neurons</i>	1	3	3
4.7	USER edits/confirms <i>step_neurons</i>	1	3	3
4.8	USER edits/confirms Maximum Number of Iterations	1	3	3
4.9	USER edits/confirms Overfitting Tolerance	1	3	3
5	USER reviews the validation parameters	1	2	2
5.1	USER edits/confirms the Validation Threshold	1	3	3
6	USER reviews the test parameters	1	2	2
6.1	USER edits/confirms the Test Pass Threshold	1	3	3
7	USER selects <i>Save Configuration</i> to store the configuration	1	1	1
8	SYSTEM validates the configuration and shows a confirmation message			
9	USER closes the configuration interface	1	1	1
TOTAL (O × CE) COST		51		
USER is a Machine Learning Engineer: Normalized Salary		× 1.67		
TOTAL HUMAN COST		85.17		

Development System Configuration

System IP Addresses

Segregation System IP Address

Messaging System IP Address

Classification System IP Address

Hyperparameter Grid

Min Layers

Max Layers

Step Layers

Min Neurons

Max Neurons

Step Neurons

Training Parameters

Maximum Number of Iterations

Overfitting Tolerance

Validation Parameters

Validation Threshold

Test Parameters

Test Pass Threshold

CloseSave Configuration

4.3.5. Configure the Classification System (Nimra)

	STEP	O	CE	$O \times CE$
1	USER opens the “Classification System Configuration form”.	1	1	1
2	SYSTEM loads the configuration interface with current/default parameters.			
3	FOR EACH input fields of the System Addresses (IP address of Development, Messaging, Preparation, Evaluation and Client-Side System)	5		
3.1	USER edits/confirms the IP address		3	15
4	USER edits/confirms the lifecycle phase	1	3	3
5	USER selects <i>Save to store the configuration.</i>	1	1	1
6	SYSTEM validates the configuration and shows a confirmation message.			
6.1	USER closes the configuration interface.	1	1	1
TOTAL (O × CE) COST		21		
USER is a Machine Learning Engineer: Normalized Salary		× 1.67		
TOTAL HUMAN COST		35.07		

CLASSIFICATION SYSTEM CONFIGURATION

EXIT

MLOps Dashboard – Speech Emotion Recognition · User: Machine Learning Engineer

SYSTEM ADDRESSES

Preparation System IP / endpoint

172.16.40.1

Endpoint where prepared sessions are received.

Segregation System IP / endpoint

172.16.40.2

Endpoint where classified sessions are sent.

Messaging System IP / endpoint

172.16.40.3

Used for configuration and monitoring messages.

CLASSIFICATION PARAMETERS

Active classifier ID / version

SER-CNN-v3

Minimum confidence threshold

0.75

Policy for low-confidence / invalid sessions

Send to manual review

Cancel

Save

FEEDBACK

Parameter Purpose

`evaluation_sys_addr` Where to send labels during evaluation phase

`client_side_sys_addr` Where to send labels during normal production

`phase` Controls routing (evaluation vs normal mode)

YOU DON'T HAVE TO REDO THE COST TABLE. WE ALREADY UPDATED.

4.3.6. Configure the Evaluation System (Rojan)

	STEP	O	CE	O × CE
1	USER opens the <i>Evaluation System Configuration</i> form	1	1	1
2	SYSTEM loads the configuration interface with default parameters			
3	FOR EACH input fields of the System Addresses (IP address of Ingestion, Classification and Messaging System)	3		
3.1	USER edits/confirms the IP address		3	9
4	FOR EACH input fields of the Evaluation parameters (Minimum Number of Labels, Maximum Total Errors Threshold and Maximum Consecutive Errors Threshold)	3		
4.1	USER edits/confirms the parameter		3	9
5	USER selects <i>Save to store the configuration</i>	1	1	1
8	SYSTEM validates the configuration and shows a confirmation message			
9	USER closes the configuration interface	1	1	1
TOTAL (O × CE) COST		21		
USER is a Data Analyst: Normalized Salary		× 1.45		
TOTAL HUMAN COST		30.45		

EVALUATION SYSTEM CONFIGURATION

EXIT

SYSTEM ADDRESSES

Ingestion System IP

172.16.20.1

Classification System IP

172.16.20.2

Messaging System IP

172.16.20.3

EVALUATION PARAMETERS

Minimum Number of Labels

50

Maximum Total Errors Threshold

5

Maximum Consecutive Errors Threshold

2

SAVE

4.3.7. Configure the Client-Side System (Francesco)

	STEP	O	CE	$O \times CE$
1	USER opens the “Configure Client-Side System form”	1	1	1
2	SYSTEM displays current configuration			
3	USER reviews the ingestion address	1	2	2
3.1	USER sets the ingestion address	1	3	3
4	USER reviews the port	1	2	2
4.1	USER sets the port	1	3	3
5	USER sets the access token	1	3	3
6	USER selects the security mode	1	1	1
6.1	USER sets the security mode	1	3	3
7	USER selects SUBMIT	1	1	1
8.1	IF the configuration is correct	0.95		
8.1.1	SYSTEM displays a confirmation notice			
8.1.2	SYSTEM saves the new configuration			
8.2	ELSE	0.05		
8.2.1	SYSTEM displays an error message and aborts			
9	USER closes the form	1	1	1
TOTAL (O × CE) COST		20		
USER is a Support Technician: Normalized Salary		× 1		
TOTAL HUMAN COST		20		

The screenshot shows a user interface for configuring a client-side system. At the top, there's a title bar with 'CFG' and 'Configure Client-Side System'. Below the title, there are several input fields and a dropdown menu:

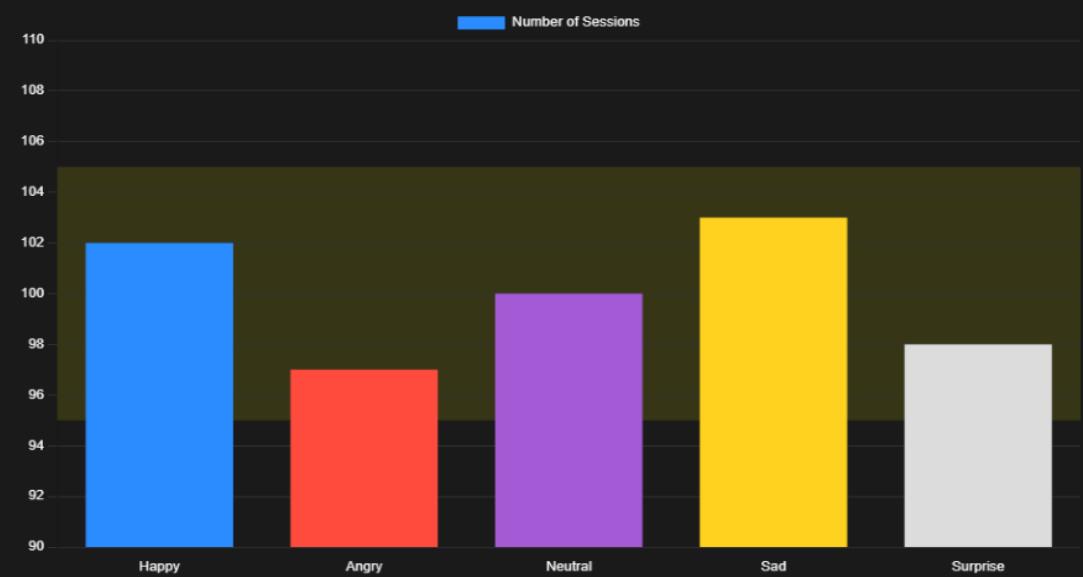
- Ingestion address:** A text input field containing the URL `https://factory.ser.ingest/api`.
- Port:** A text input field containing the value `443`.
- Access token:** A text input field containing redacted content.
- Security mode:** A dropdown menu with options: `TLS` (selected), `Plain`, `SSL`, and `mTLS`.

At the bottom of the form are two buttons: `SUBMIT` (green) and `RESET` (orange).

4.3.8. Check Data Balancing (Rojan)

	STEP	O	CE	O × CE
1	USER opens the data balancing form	1	1	1
2	SYSTEM loads the report (a bar chart with five bars, one for each emotion class)			
3	FOR EACH <i>class</i>	5		2
3.1	USER reads the number of sessions of the <i>class</i>		3	15
3.2	USER reads the tolerance band ($\pm 5\%$)		3	15
3.3	USER compares the <i>class</i> sessions count with the tolerance band		3	15
3.4	USER determines if the <i>class</i> is BALANCED		3	15
4.1	IF data is balanced	0.2		
4.1.1	USER selects YES		1	0.2
4.2	ELSE	0.8		
4.2.1	USER selects NO		1	0.8
4.2.2	SYSTEM shows “Send Configuration” section			
4.2.3	USER fills required fields		3	2.4
5	USER selects SUBMIT	1	1	1
6	SYSTEM shows confirmation and stores the outcome			
7	USER closes the form	1	1	1
TOTAL (O × CE) COST		68.4		
USER is a Data Analyst: Normalized Salary		× 1.45		
TOTAL HUMAN COST		99.18		

DATA BALANCING



Parameters

Number of Sessions: **500**

Tolerance (%): **5%**

System Verdict
BALANCED

Approve Data Balance: YES NO

SUBMIT

4.3.9. TASK Check Input Coverage (Francesco)

	STEP	O	CE	O × CE
1	USER opens the “Coverage Report”	1	1	1
2	SYSTEM loads the report (a radar scatter plot)			
3.1	FOR EACH feature	8		
3.1.1	USER check if the feature on the radar plot has the correct distribution on the radius		4	32
4.1	IF the input coverage is satisfied	0.33		
4.1.1	USER selects ACCEPT		1	0.33
4.2	ELSE	0.67		
4.2.1	USER selects REJECT		1	0.67
5	SYSTEM shows confirmation and stores the outcome			
6	USER closes the form	1	1	1
TOTAL (O × CE) COST		35		
USER is a Data Analyst: Normalized Salary		× 1.45		
TOTAL HUMAN COST		50.75		

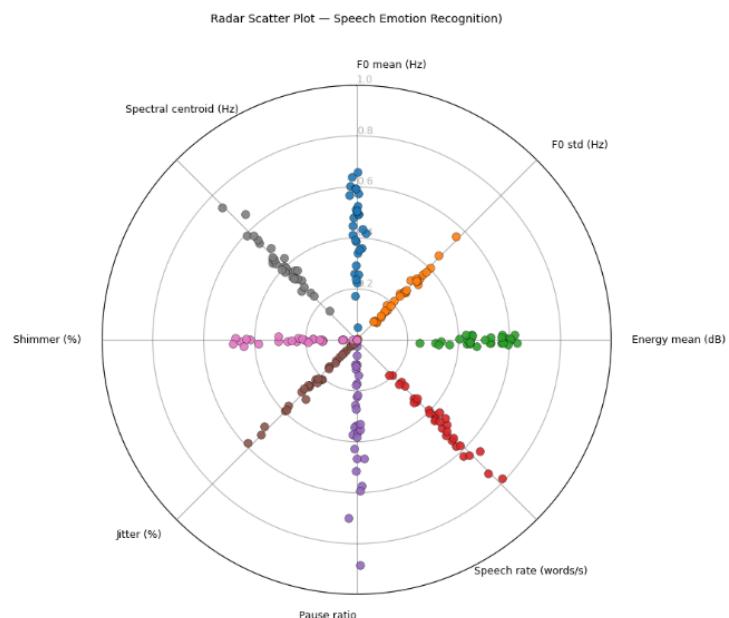
Choose features:

1. **F0 mean (Hz):** Average of the fundamental frequency of the utterance. Interpretable as the overall "tone" of the voice;
2. **F0 std (Hz):** Standard deviation of the F0. Measures tonal variability (flat vs. modulated intonation);
3. **Energy mean (dB):** Average of perceived energy. Indicates arousal (high = anger/enthusiasm, low = sadness);
4. **Speech rate (words/syllables per second):** Rate of speech calculated per word or syllable. Interpretable as agitation vs. calm;
5. **Pause ratio (duration of silence / total duration in seconds):** Percentage of silent time. Useful for distinguishing frequent pauses or fragmented speech;
6. **Jitter (relative, %):** Cycle-to-cycle variation in frequency. Indicates instability/tension in the voice;
7. **Shimmer (relative, %):** Cycle-to-cycle variation in amplitude. Reports tremor or vocal fatigue;
8. **Spectral centroid (Hz):** "Center of mass" of the spectrum. Measures the brightness/timbre of the voice (higher = "brighter").

DCR

Data Coverage Report

Radar Scatter Plot — Speech Emotion Recognition

[View](#)[Export](#)**Features**

Range

FEATURE	[MIN; MAX]
F0 mean (Hz)	[70; 330]
F0 std (Hz)	[2; 80]
Energy mean (dB)	[30; 90]
Speech rate (words/s)	[1; 6]
Pause ratio	[0; 0.6]
Jitter (%)	[0; 1.5]
Shimmer (%)	[0; 6]
Spectral centroid (Hz)	[500; 4500]

[Accept](#)[Reject](#)

In this example, the graph shows that the data covers the entire acoustic feature space well, with no gaps or marked outliers (except for something in Pause Ratio and Spectral Centroid), and reveals which features exhibit greater or lesser variability (little for standard F0, much for speech rate and spectral centroid). The plot serves to verify feature completeness, variability, and quality before modeling.

3.3.10. TASK Set Number of Iterations (Nimra)

	STEP	O	CE	O × CE
1	USER opens “Set Number of Iterations” form	1	1	1
2	SYSTEM loads a default/current number of iterations			
3	USER see the previous number of iterations	1	2	2
4	USER edits/confirms the number of iterations based on task complexity and personal experience	1	3	3
4	USER selects “Submit” button to confirm the number of iterations	1	1	1
5	SYSTEM shows confirmation and stores new value			
6	USER closes the form	1	1	1
TOTAL (O × CE) COST		8		
USER is a Machine Learning Engineer: Normalized Salary		× 1.67		
TOTAL HUMAN COST		13.36		

Set Number of Iterations

MLOps Dashboard · Speech Emotion Recognition · Machine Learning Engineer

EXIT

Learning Curve
Training and validation loss over iterations (read-only).

Training Iterations	Training Loss	Validation Loss
0	6.0	6.0
100	4.0	3.5
200	3.0	2.5
300	2.5	2.0
400	2.0	1.8
500	1.8	1.6
600	1.6	1.5
700	1.5	1.4
800	1.4	1.3
900	1.3	1.2
1000	1.2	1.1
1100	1.1	1.2
1200	1.0	1.8

Iterations configuration
Suggested number of iterations
1250

Are suggested iterations acceptable?
 Yes – keep suggested iterations
 No – set manually

Manual setting (used when suggested iterations are rejected)
New number of iterations Comment (optional)
e.g., 900 Reason for manual change

Cancel **Submit**

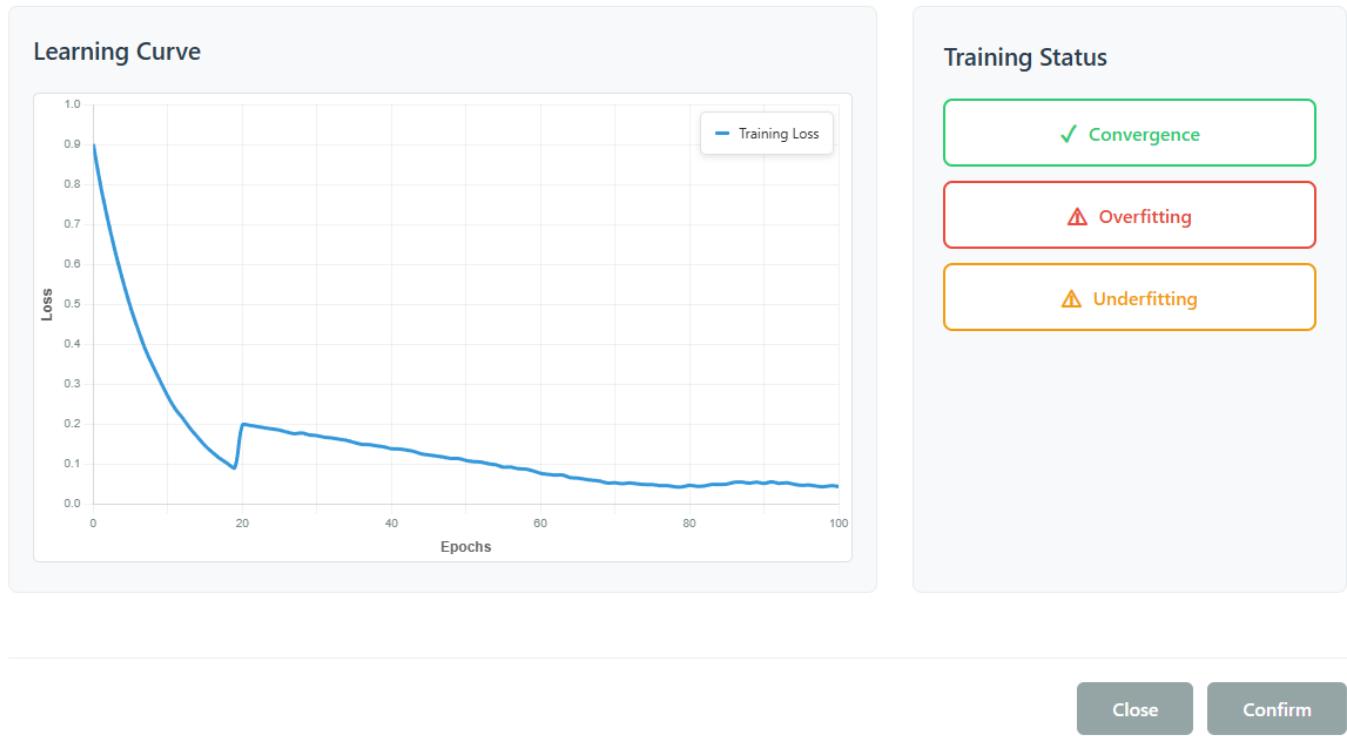
FEEDBACK:

No graph just a set num of iteration (Please refer to past proj)

3.3.11. TASK Check Calibration Plot (Ronald)

STEP	ACTION	O	CE	$O \times CE$
1	USER opens the “Check Calibration Plot” form	1	1	1
2	SYSTEM loads the Learning Curve plot (training loss curve)			
3	USER inspects Learning Curve plot for convergence, overfitting or underfitting	1	4	4
3.1	IF the Learning Curve plot converges (loss stabilizes)	0.4		
3.1.1	USER confirms by selecting “Convergence” button		1	0.4
3.1.2	SYSTEM generates and shows validation report			
3.2	ELSE	0.6		
3.2.1	USER selects “Overfitting” or “Underfitting” button		1	0.6
4	USER selects “submit”	1	1	1
5	SYSTEM shows confirmation message for submission			
6	USER closes form	1	1	1
TOTAL (O × CE) COST		8		
USER is a ML Engineer: Normalized Salary		× 1.67		
TOTAL HUMAN COST		13.36		

Check Calibration Plot



3.3.12. TASK Check Validation Results (Rojan)

	STEP	O	CE	O × CE
1	USER opens validation results form	1	1	1
2	SYSTEM loads validation report			
3	FOR EACH model	5		
3.1	USER checks whether delta MSE (the difference between Validation and Training MSE) for given model is under the overfitting tolerance		3	15
3.2	USER marks the model OK or Overfit based on the delta MSE		3	15
4.1	IF validation results acceptable	0.95		
4.1.1	USER identifies the best network (the lowest validation error marked OK)		3	2.85
4.1.2	USER identifies the second-best network (the second lowest validation error marked OK)		3	2.85
4.1.3	USER compares the validation error between the two networks		3	2.85
4.1.4	If the error is very similar (the difference is one order of magnitude w.r.t. their error), USER selects the network with the lowest complexity otherwise selects the best network		4	3.8
4.2	ELSE	0.05		
4.2.1	USER rejects the result		1	0.05
4.2.2	SYSTEM shows “Reason for Rejection” section			
4.2.3	USER provides the feedback for the rejection		3	0.15
5	USER selects SUBMIT	1	1	1
6	SYSTEM shows confirmation and stores decision			
7	USER closes the form	1	1	1
TOTAL (O × CE) COST		45.55		
USER is a ML Engineer: Normalized Salary		× 1.67		
TOTAL HUMAN COST		76.0685		

Check Validation Results

Candidate Models

Top-5 by Validation MSE

Model	Layers	Neurons	Train MSE	Valid MSE	Δ MSE	Status	Best
M-014	3	128	0.083	0.091	0.008	OK	<input checked="" type="radio"/> Best
M-011	4	256	0.081	0.093	0.012	OK	<input type="radio"/> Best
M-008	5	512	0.076	0.110	0.034	OVERFIT	<input type="radio"/> Best
M-006	2	64	0.089	0.104	0.015	OK	<input type="radio"/> Best
M-002	3	64	0.097	0.122	0.025	OVERFIT	<input type="radio"/> Best

Overfitting tolerance
 < 0.02

Reviewer comment (optional)

 Reject

 Approve

3.3.13. TASK Check Test Results (Francesco)

	STEP	O	CE	O × CE
1	USER opens “ <i>Test Results form</i> ”	1	1	1
2	SYSTEM shows the test results			
3	USER checks Delta	1	3	3
4	USER checks overfitting tolerance	1	3	3
5	USER checks if the difference between the test results and the validation results (Delta) is below the overfitting tolerance	1	3	3
6.1	IF the test results are not satisfactory.	0.99		
6.1.1	USER selects ACCEPT		1	0.99
6.1.2	SYSTEM saves <i>Classifier</i>			
6.2	ELSE	0.01		
6.2.1	USER selects REJECT		1	0.01
6.2.2	SYSTEM sends configuration message to messaging system			
7	SYSTEM shows confirmation and stores decision			
8	USER closes the form		1	1
TOTAL (O × CE) COST		12		
USER is a ML Engineer: Normalized Salary		× 1.67		
TOTAL HUMAN COST		20.04		

TR TEST REPORT
Model Selection Summary

Winner Classifier	NET-L3N128
Validation MSE	0.031
Test MSE	0.044
Delta	0.013
Overfitting Threshold	0.17

Delta = | Validation MSE - Test MSE |

ACCEPT
REJECT

3.3.14. TASK Evaluate Classifier (Francesco)

	STEP	O	CE	$O \times CE$
1	USER opens the “Evaluate Classifier Performance form”	1	1	1
2	SYSTEM displays a session table containing the expert label (ground truth) and the classifier label (predicted). Any mismatch between the two indicates an error.			
3	USER reads total errors number	1	3	3
4	USER reads total errors number threshold	1	3	3
5	USER checks if the total errors number is below the threshold	1	3	3
6	USER reads total consecutive errors	1	3	3
7	USER reads total consecutive errors threshold	1	3	3
8	USER checks if the total consecutive errors number is below the threshold	1	3	3
9.1	IF at least one error number is above its threshold	0.14		
9.1.1	USER selects REJECT		1	0.14
9.1.2	SYSTEM sends configuration message to messaging system			
9.2	ELSE	0.86		
9.2.1	USER selects ACCEPT		1	0.86
9.2.2	SYSTEM marks the <i>Classifier</i> as Good			
10	SYSTEM shows confirmation and stores decision			
11	USER closes the form	1	1	1
TOTAL (O × CE) COST		21		
USER is a Data Analyst: Normalized Salary		× 1.45		
TOTAL HUMAN COST		30.45		

CER Classifier Evaluation Report

Predicted Labels vs Ground Truth

Results			
SESSION ID	Classifier Label	Expert Label	Correct
1	Happy	Happy	YES
2	Neutral	Sad	NO
3	Sad	Sad	YES
4	Surprise	Surprise	YES
5	Angry	Happy	NO
6	Happy	Happy	YES
7	Neutral	Neutral	YES
8	Surprise	Surprise	YES
9	Sad	Sad	YES
10	Happy	Happy	YES

Scroll down the table to see the individual rows.

Q

Summary

Misclassified Labels	Misclassified Threshold
7	10
Consecutive Misclassified Labels	Consecutive Misclassified Threshold
3	4
Misclassified OK (7 < 10)	Consecutive OK (3 < 4)

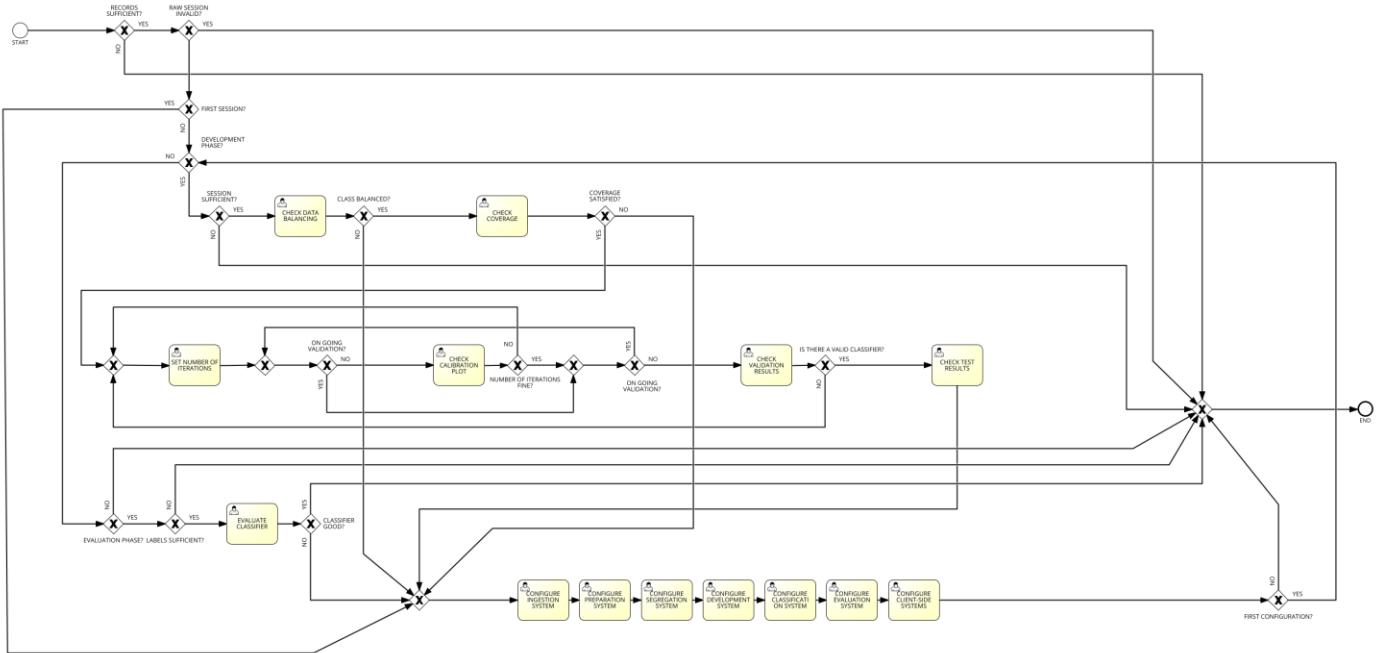
The boxes turn green when the value is below the corresponding threshold.

Documentation on features:

<https://drive.google.com/file/d/1TGj5K0cTLhr3NYvcmxfeUKfatDMmIdQ2/view?usp=sharing>

5 Simulation

5.1 Collapsed Workflow



5.2 Token Definition

We choose a Session as a Token because the simulation engine represents each process instance by a moving execution marker. That marker must carry the session identifier through the workflow so the simulator can account for branching logic, resource use, and frequencies. Modeling the session as a token ensures the simulator can track and route each instance correctly, especially when probabilities or conditional paths depend on whether it is the first occurrence or a later one.

5.3 AS-IS Simulation

This are the **Uniform Distribution ($\pm 5\%$)** for Human Tasks' Costs (in alphabetical order):

- **CHECK CALIBRATION PLOT (13.36)** → [12.692; 14.028]
- **CHECK COVERAGE (50.75)** → [48.2125; 53.2875]
- **CHECK DATA BALANCING (99.18)** → [94.221; 104.139]
- **CHECK TEST RESULTS (20.04)** → [19.038; 21.042]
- **CHECK VALIDATION RESULTS (76.0685)** → [72.2651; 79.8719]
- **CONFIGURE CLASSIFICATION SYSTEM (35.07)** → [33.3165; 36.8235]
- **CONFIGURE CLIENT-SIDE SYSTEMS (20)** → [19; 21]
- **CONFIGURE DEVELOPMENT SYSTEM (85.17)** → [80.9115; 89.4285]
- **CONFIGURE EVALUATION SYSTEM (30.45)** → [28.9275; 31.9725]
- **CONFIGURE INGESTION SYSTEM (58.45)** → [55.5275; 61.3725]
- **CONFIGURE PREPARATION SYSTEM (56.78)** → [53.941; 59.619]
- **CONFIGURE SEGREGATION SYSTEM (50.1)** → [47.595; 52.605]
- **EVALUATE CLASSIFIER (30.45)** → [28.9275; 31.9725]
- **SET NUMBER OF ITERATIONS (13.36)** → [12.692; 14.028]

This are the **percentages** of the gateways (in alphabetical order):

GATEWAYS	YES	NO	MOTIVATION
CLASS BALANCED?	20%	80%	Documentation: "Balanced classes in 5 iterations, then 4 occurrences ko and 1 ok, i.e. 1/5 = 20% chance ok".
CLASSIFIER GOOD?	14%	86%	Documentation: "Evaluation fine for 6 iterations, 6 occurrences ok and 1 ko, the 1/7 = 14% ok".
COVERAGE SATISFIED?	33%	67%	Documentation: "Input coverage in 3 iterations, 2 occurrences ko and 1 ok, then 1/3 = 33% ok".
DEVELOPMENT PHASE?	9%	91%	In this gateway, out of 5550, 500 are in the development phase
EVALUATION PHASE?	1%	99%	In this gateway, out of 5550, 50 are in the evaluation phase
FIRST CONFIGURATION?	0.05%	99.95%	Only 5 out of 10,000 input tokens represent a first session, corresponding to the five classifiers that each require exactly one initial session.
FIRST SESSION?	0.05%	99.95%	Only 5 out of 10,000 input tokens represent a first session, corresponding to the five classifiers that each require exactly one initial session.
IS THERE A VALID CLASSIFIER?	95%	5%	Documentation: "Validation result 95% fine".
LABELS SUFFICIENT?	71%	29%	Given 5550 valid sessions and the assumption that these produce five

			proper classifiers, the preceding gateway probabilities imply that 71% of all sessions must contain sufficient labels to satisfy the documentation's assumptions.
NUMBER OF ITERATIONS FINE?	80%	20%	Documentation: "Number of iterations in 5 iterations, then 4 occurrences ko and 1 ok, i.e. $1/5 = 20\%$ chance ok".
ON GOING VALIDATION?	97.5%	2.5%	For the selected hyperparameter grid, the model has 5 possible layer counts (1–5 in steps of 1) and 8 possible neuron settings (32–256 in steps of 32), yielding 40 total configurations. Since only one configuration is under ongoing validation, 39 out of 40 are already validated. This corresponds to 97.5% marked as "Yes" for ongoing validation and 2.5% marked as "No."
RAW SESSION INVALID?	10%	90%	We assume that 90% of the times the raw session is valid.
RECORDS SUFFICIENT?	10%	90%	We assume that 90% of the times the raw session is valid.
SESSION SUFFICIENT?	99%	1%	We set the proportion of sessions with sufficient labels to 99% because, under the document's assumptions, roughly 545 such sessions would be required to obtain five final good classifiers. In this scenario we begin with only 500, which is already below the required amount.