



TÉCNICO LISBOA

# Engagement Detection with Mixed Reality Intelligent Systems

MSc in Mechanical Engineering  
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# My MSc thesis



Figure 1 – HoloLens 2.



Figure 2 – UR5e robotic arm.

# Data Acquisition



Video 1 – Spheres app.

# Feature Extraction



## Eye Tracking

Blink Rate  
Blink Duration



## Head Pose

Orientation  
Linear Velocity  
Angular Velocity



## IMU

Accelerometer  
Gyroscope  
Magnetometer

# Preprocessing

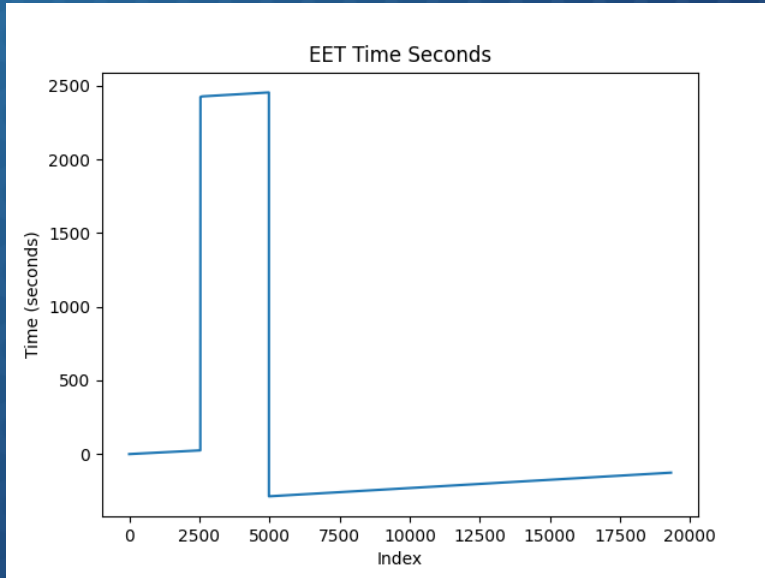


Figure 3 – Inter-synchronization.

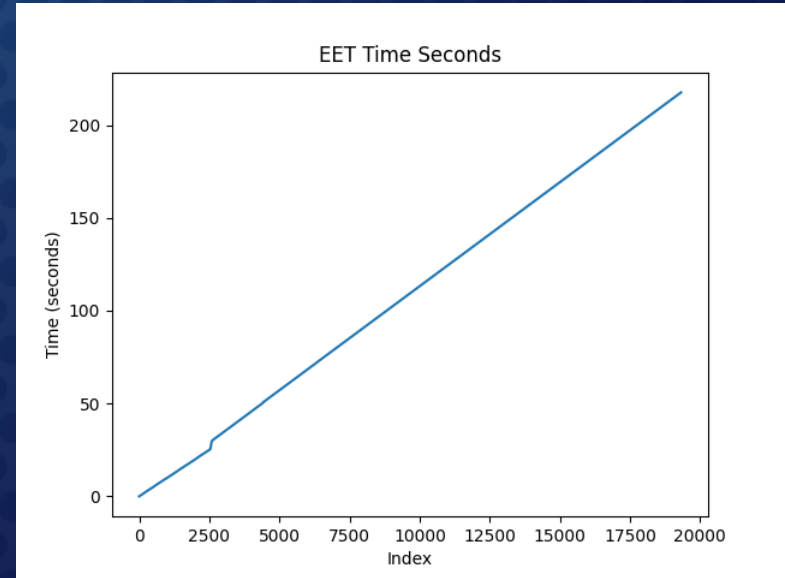


Figure 4 – Intra-synchronization.

# Feature Selection

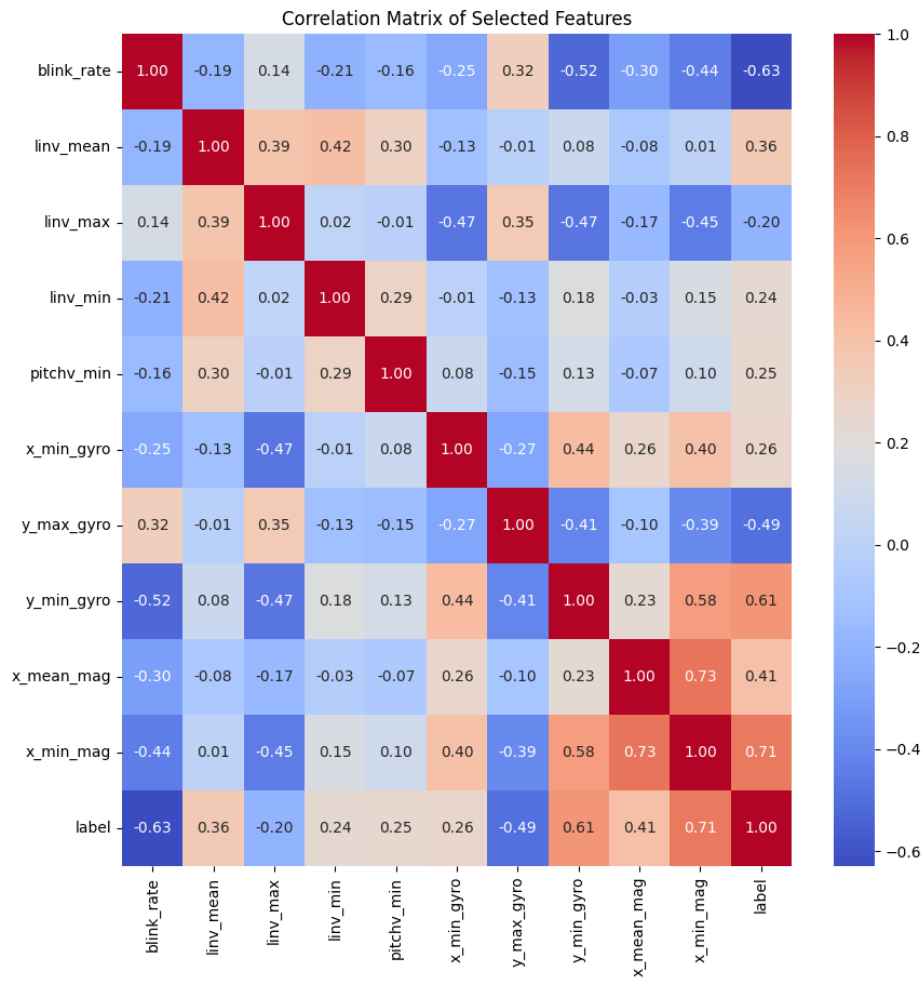


Figure 5 –  
Correlation matrix  
of selected  
features.

# Model Architecture

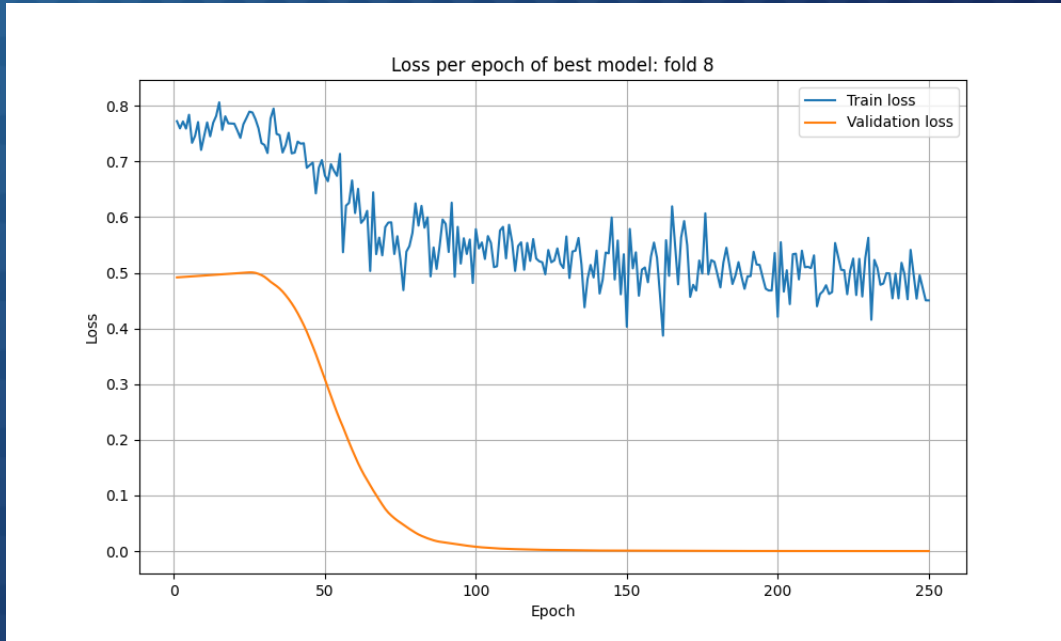


Figure 6 – Loss function per epoch.



# Model Performance - NN

|           | Test  | Total |
|-----------|-------|-------|
| Accuracy  | 0.952 | 0.981 |
| F1 Score  | 0.963 | 0.984 |
| Precision | 1     | 0.984 |
| Recall    | 0.929 | 0.984 |

Table 1- Evaluation metrics for the NN.

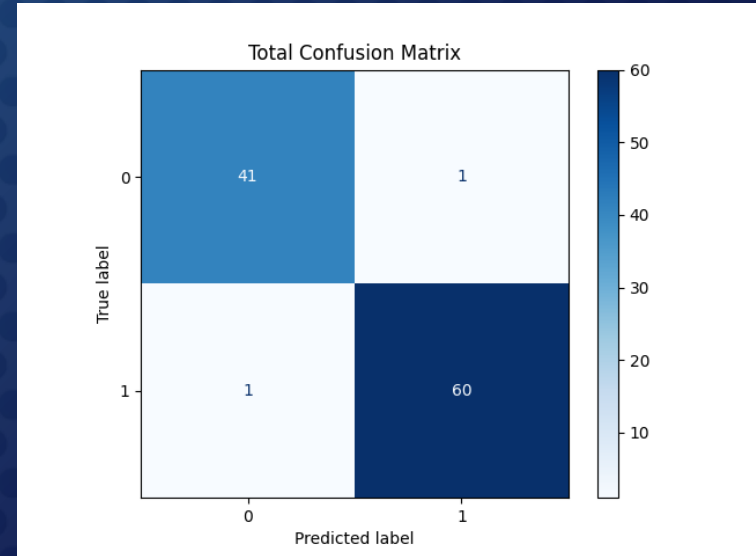


Figure 7 – Confusion matrix for the total dataset with NN.



# Model Performance - TSK

|           | Test  | Total |
|-----------|-------|-------|
| Accuracy  | 0.952 | 0.981 |
| F1 Score  | 0.963 | 0.984 |
| Precision | 1     | 0.984 |
| Recall    | 0.929 | 0.984 |

Table 2 – Evaluation metrics for the TSK.

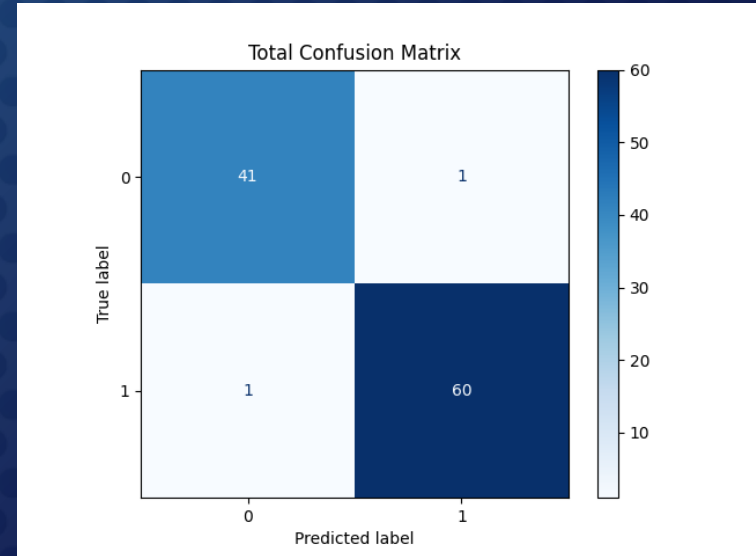


Figure 8 – Confusion matrix for the total dataset with TSK.

# Conclusion

- User engagement in MR can be assessed using eye tracking, body pose and inertial data;
- Blink patterns, head pose and motion features are meaningful indicators of engagement.
- Intelligent algorithms can detect engagement vs. disengagement even with limited data, showing great potential;

# Future Work – Second Project

- Dataset expansion with multiple users;
- Improve synchronization;
- Enhance feature extraction (fixations and saccadic movement);
- Optimize models with parameter tuning and complex architectures;
- Implement real-time adaptation.