Engagement Detection with Mixed Reality
Intelligent Systems

MSc in Mechanical Engineering
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João Bento | 95940

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
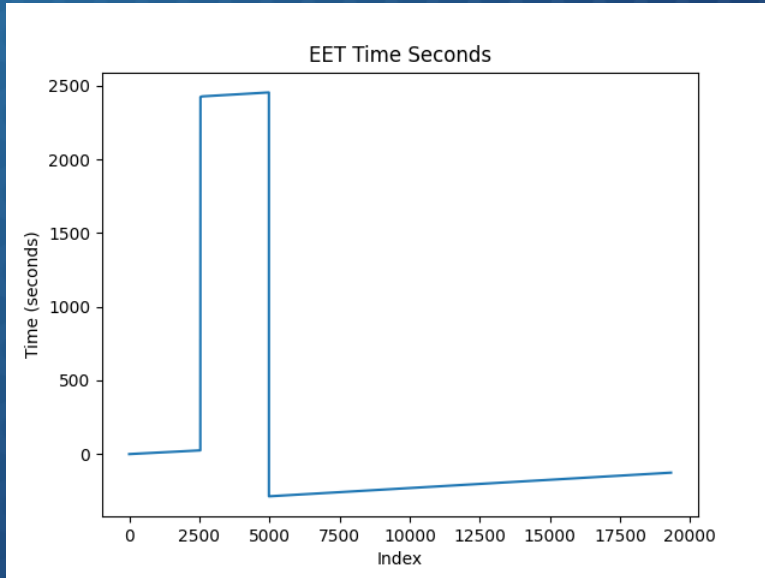Gyroscope
Magnetometer
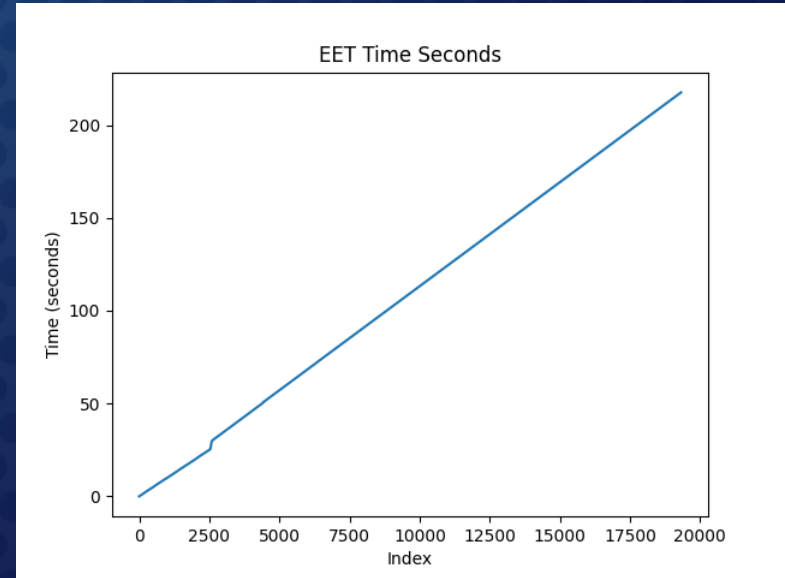
Preprocessing

Figure 3 – Inter-synchronization.
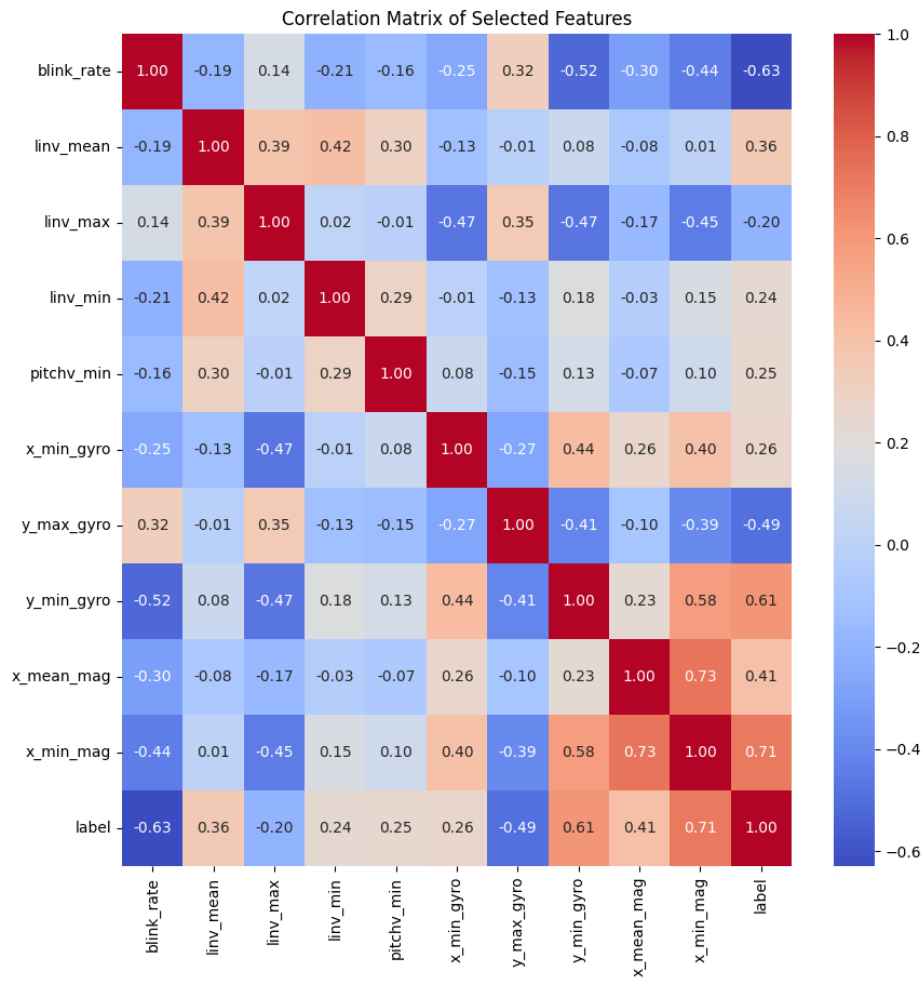

Figure 4 – Intra-synchronization.

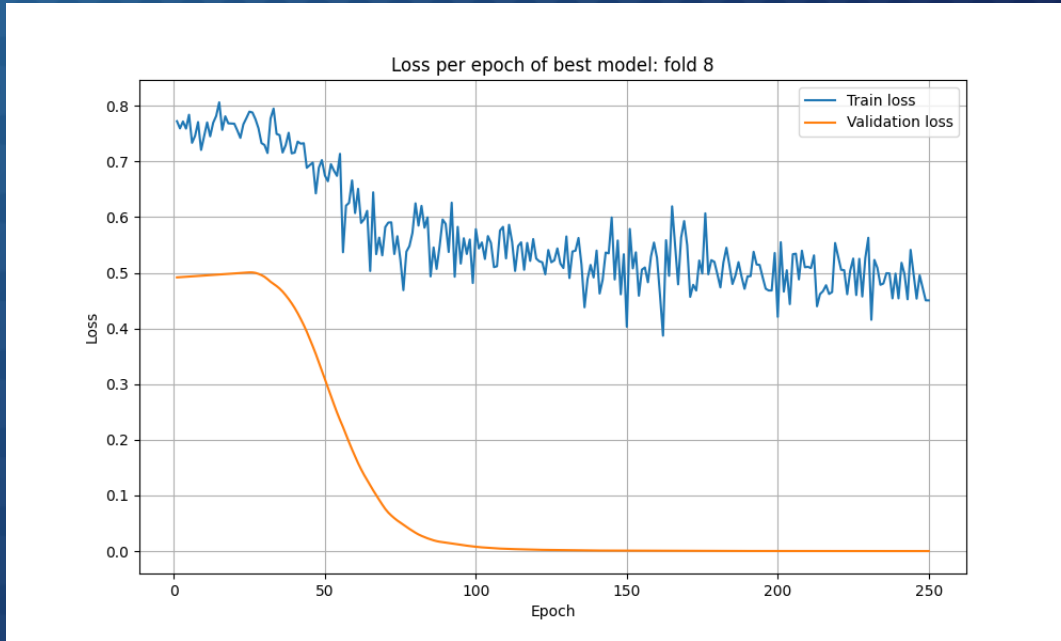Figure 5 – Correlation matrix of selected features.

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

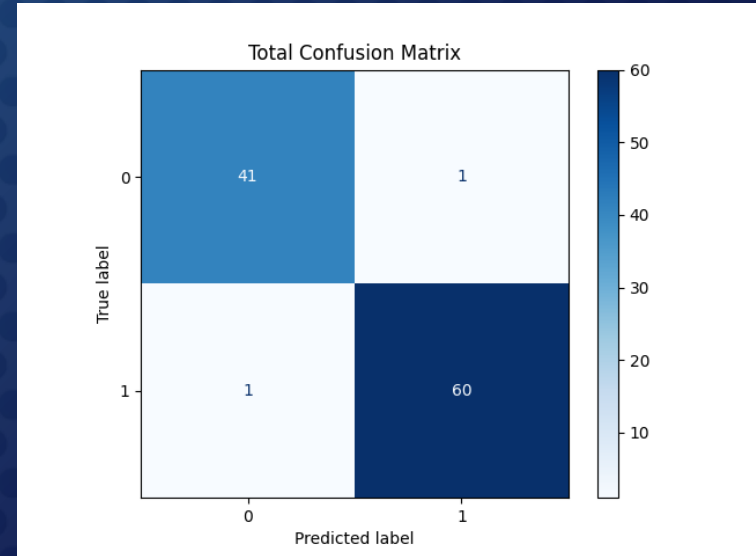Table 2 – Evaluation metrics for the TSK.

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.