


Wearable Sensing Technologies for Monitoring Real-life Activities in Spinal Cord Injury Individuals

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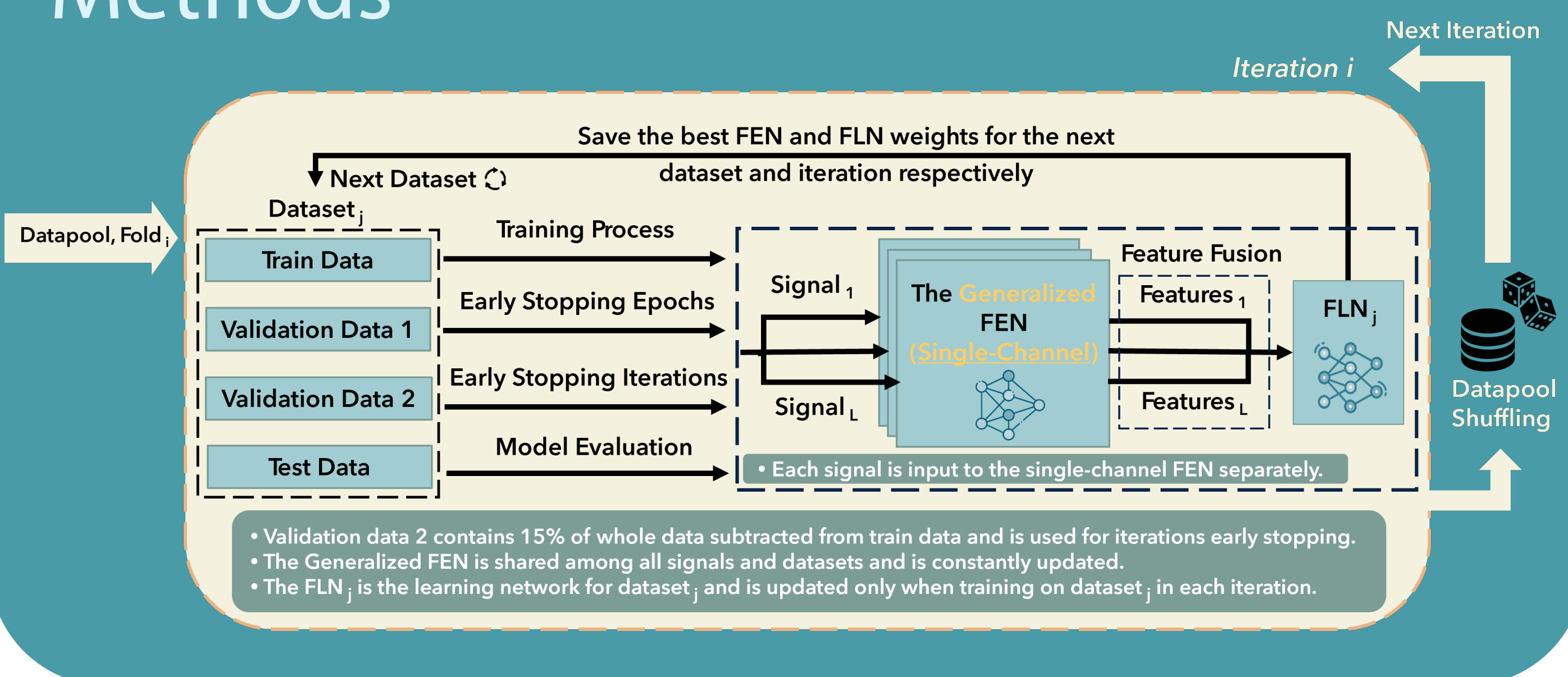
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Our wearable sensor system can unobtrusively track vitals, activities of daily living, and behaviors in community setting of individuals with spinal cord injuries.

Motivation

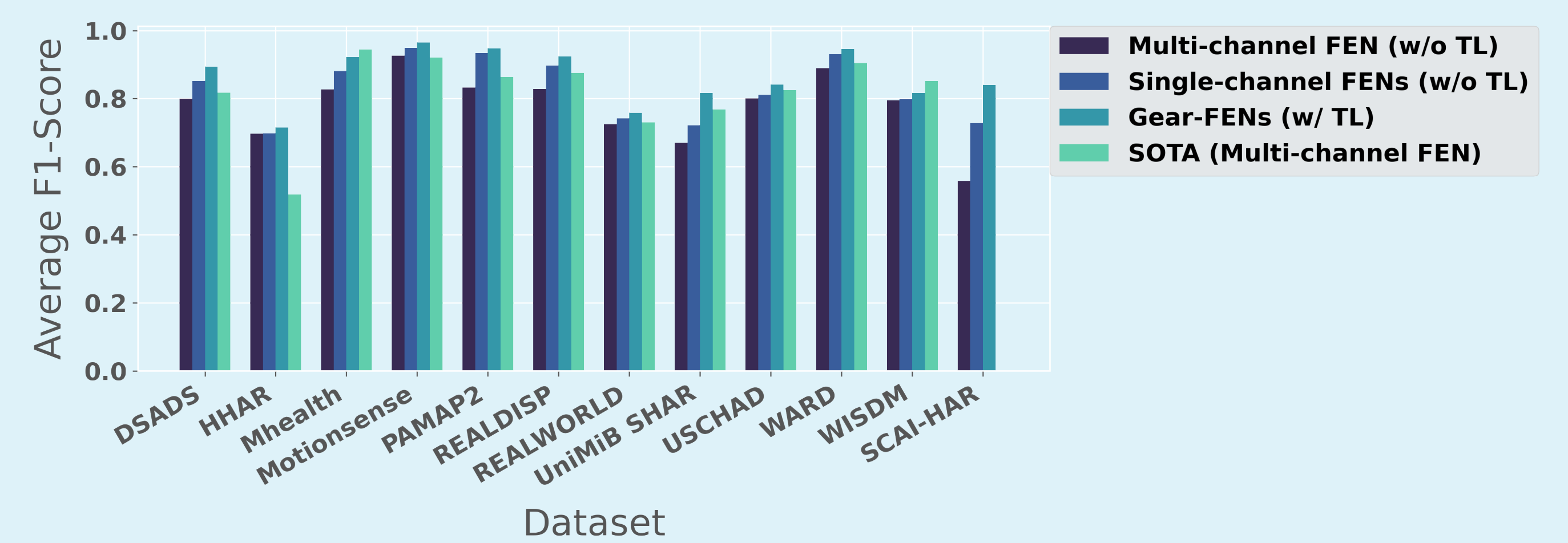
- Secondary health conditions impose premature mortality for SCI individuals.
- Monitoring vitals and ADLs can predict the onset of secondary health conditions.

Methods

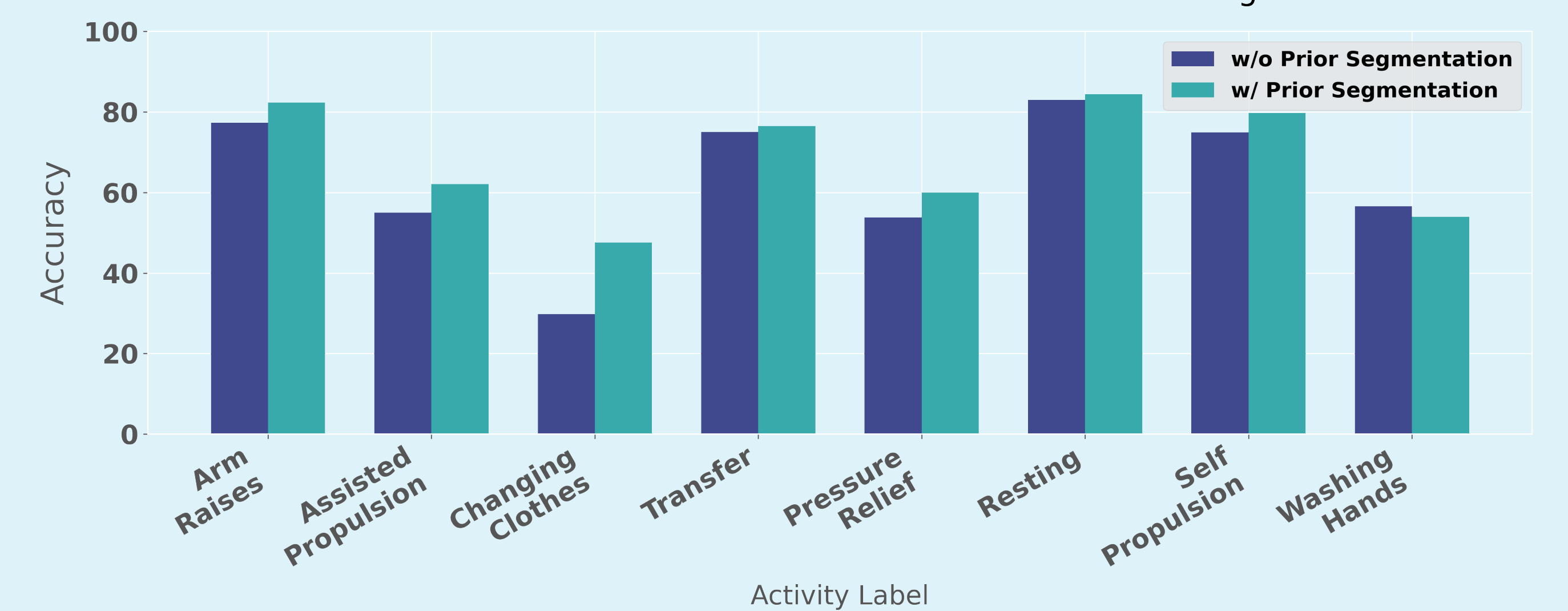


Results

Average F1-Scores Across Different Feature Extraction Methods (Attention-Based Residual BiLSTM Classification Network)

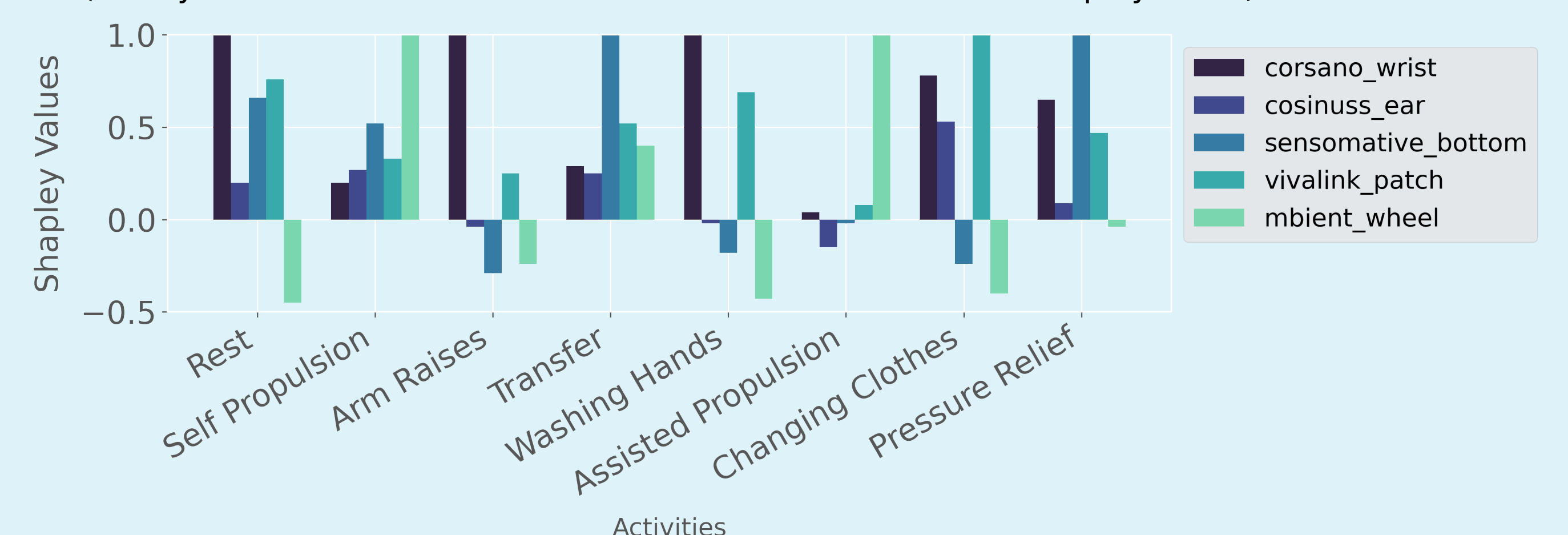


Performance of ADL Classification in Real-life Setting



Shapley Values per Device and Activity (Devices are the players)

(Binary one-vs-rest F1-score from XGBoost Classifier as the Shapley score)



$$\varphi_i(v) = \sum_{S \subseteq N \setminus \{i\}} \frac{|S|!(n-|S|-1)!}{n!} (v(S \cup \{i\}) - v(S))$$

$\varphi_i(v)$: Represents the Shapley value for device or player i .

$\sum_{S \subseteq N \setminus \{i\}}$: Summation over all subsets S that do not include i .

$\frac{|S|!(n-|S|-1)!}{n!}$: The weight for each subset S , representing the probability of i being added to S .

$v(S \cup \{i\}) - v(S)$: The marginal contribution of device i to the subset S .

Discussion

- Our sensory system can be used for monitoring vitals and ADLs in SCI community setting.
- Our feature extraction model surpassed SOTA on several public datasets.
- Wristbands, Pressure mat (bottom), Chest ECG patch, and the Wheel IMU are the most important sensors for ADL classification.

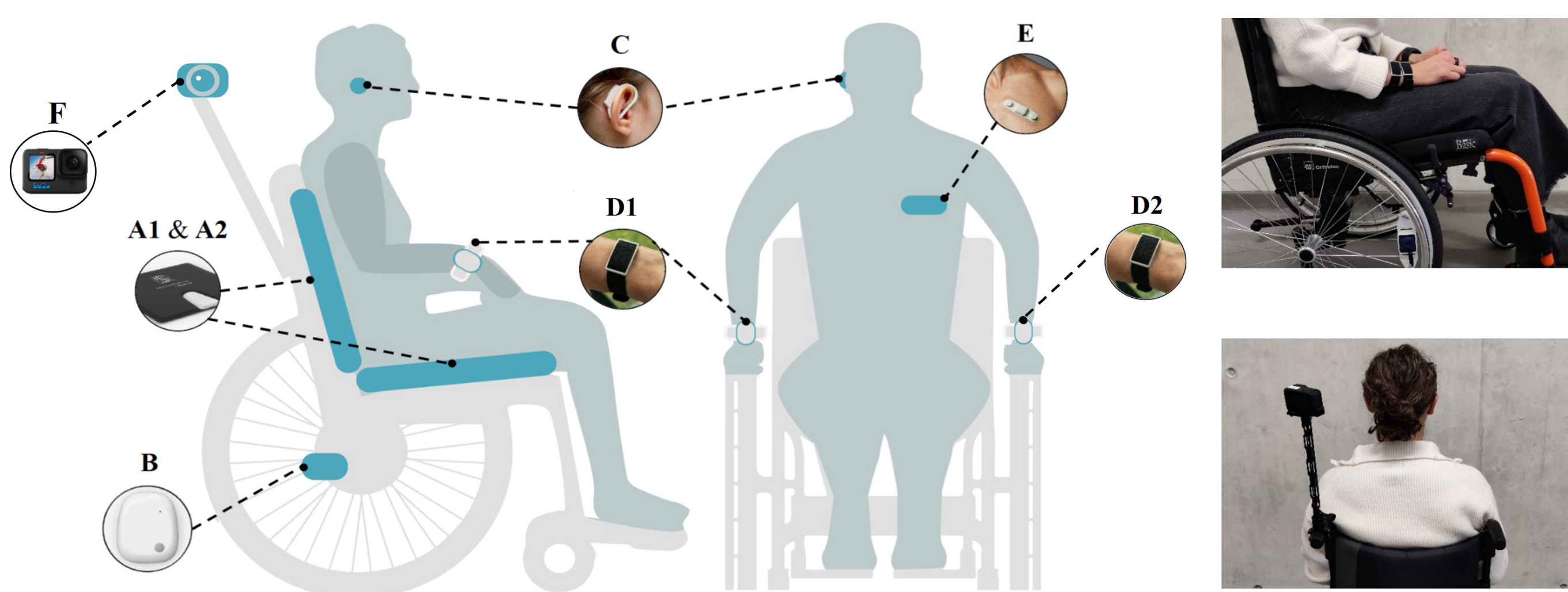
References

[1] Ejtehadi, M., Amrein, S., Hoogland, I. E., Riener, R., and Paez-Granados, D. Learning activities of daily living from unobtrusive multimodal wearables: Towards monitoring outpatient rehabilitation. In 2023 International Conference on Rehabilitation Robotics (ICORR), pp. 1–6, 2023. doi: 10.1109/ICORR58425.2023.10304743.

[2] Ejtehadi, M., et al., GeAR-FEN: Generalized Feature Representation for Kinematic Human Activity Recognition (to be submitted)

[3] Zhang, J., Liu, Y., & Yuan, H. (2023). Attention-Based Residual BiLSTM Networks for Human Activity Recognition. In IEEE Access (Vol. 11, pp. 94173–94187). Institute of Electrical and Electronics Engineers (IEEE). <https://doi.org/10.1109/access.2023.3310269>.

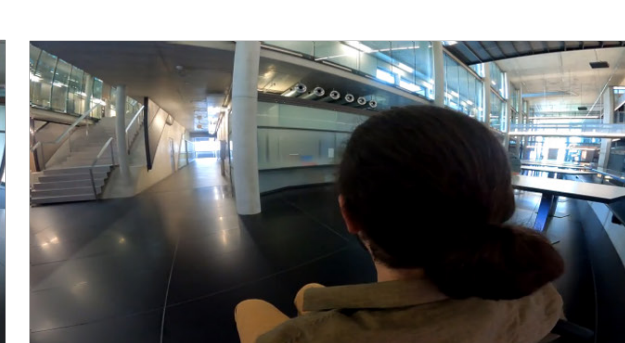
Data



Arm Raises

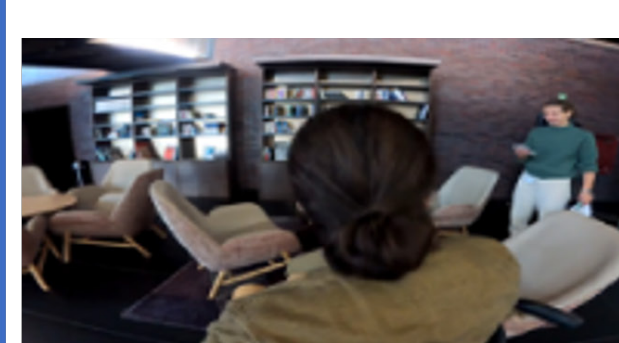


Pressure Relief



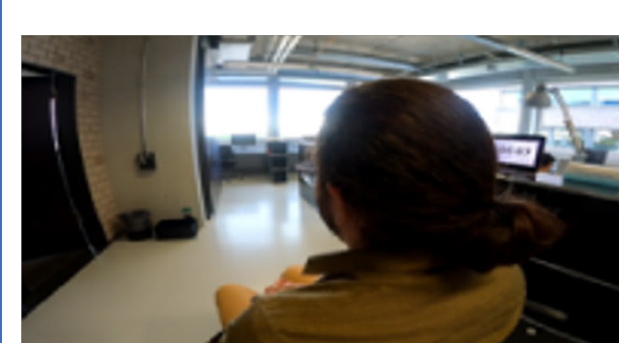
Exercise

Transfer



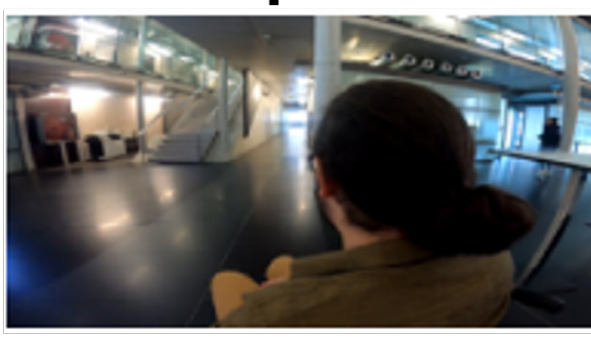
Transfer

Resting

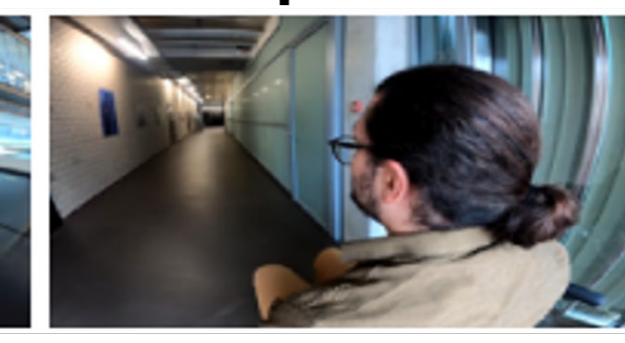


Sedentary

Assisted Propulsion



Self Propulsion



Mobility

Washing Hands

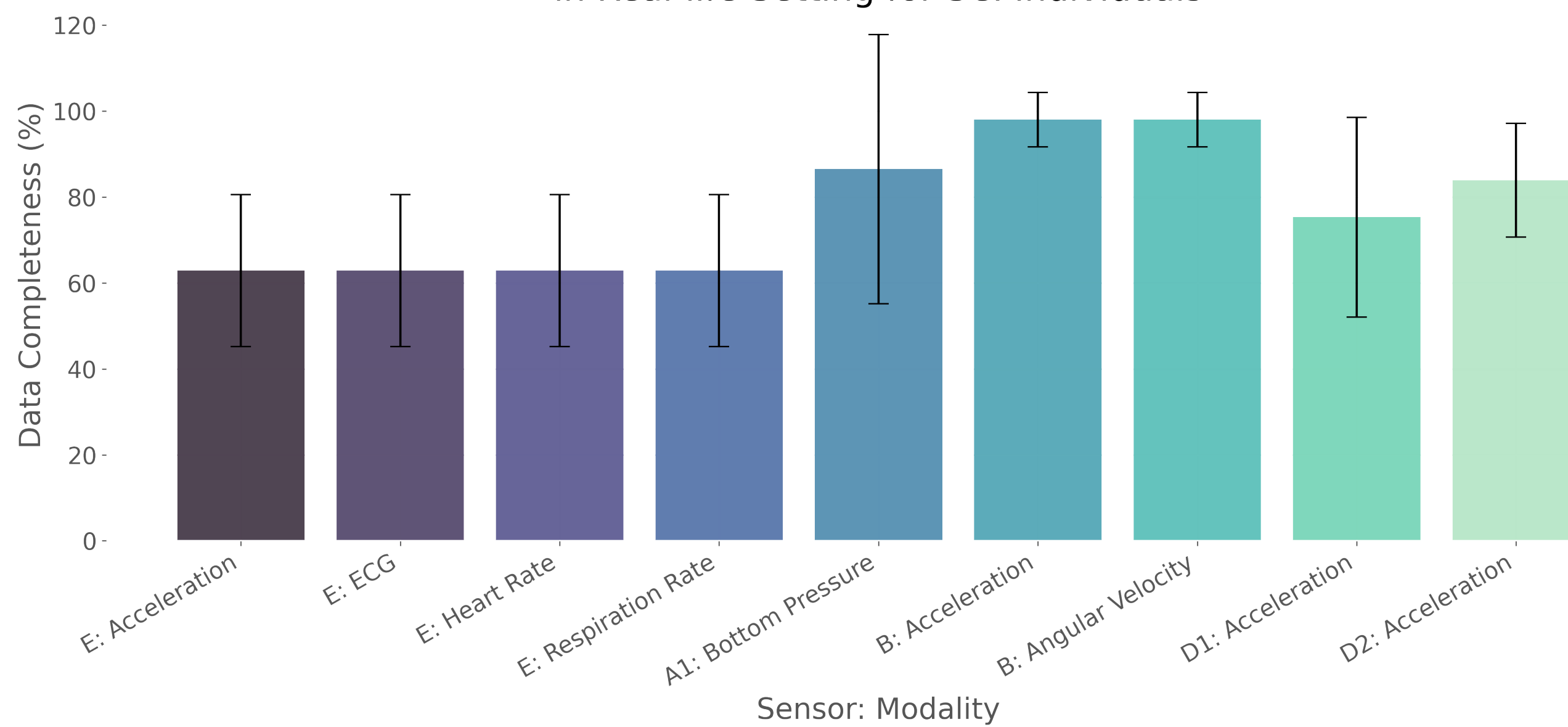


Changing Clothes



Self Care

Non-exhaustive Demonstration of Data Completeness of the Sensory System in Real-life Setting for SCI Individuals



- Analyses were done with 10/15 participants

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