

SHL Transportation/Locomotion Recognition Challenge 2020

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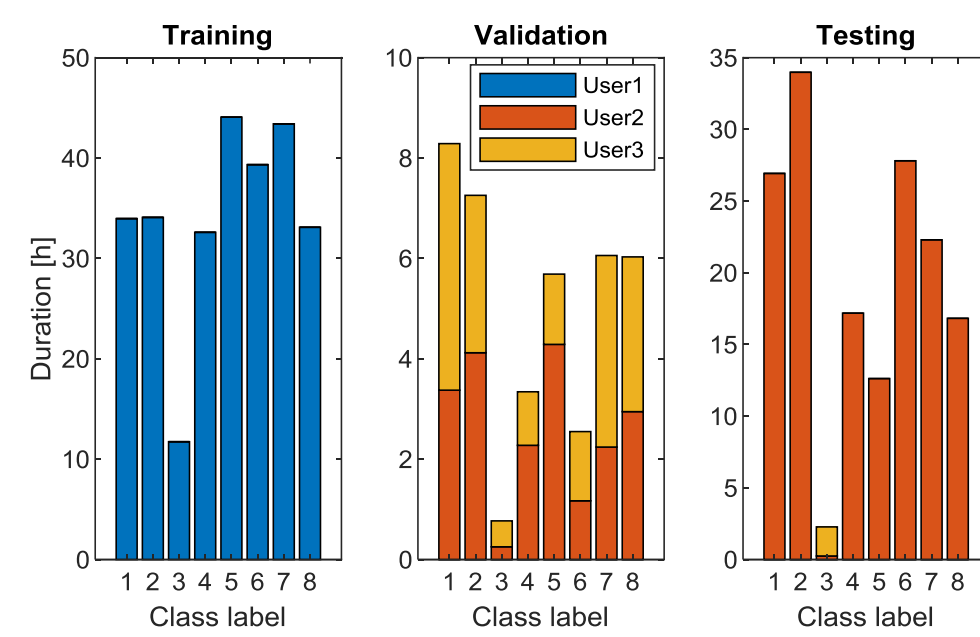
1. SHL recognition challenge

- The third version following successful 2018 [3, 4] and 2019 [5]
- Objective
 - Smartphone-based locomotion/transportation recognition
 - Algorithm designed to be smartphone user independent
 - Trained using data from User1 (Hand, Hips, Torso, Bag smartphones)
 - Evaluated on the data from User2 and User3 (Hips – unknown to participants)

- Eight classes
1 - Still; 2 - Walk; 3 - Run; 4 - Bike; 5 - Car; 6 - Bus; 7 - Train; 8 - Subway

Dataset

- Three users from SHL dataset [1, 2]
- Smartphones at 4 body positions
 - Ha – hand; Hi – Hips; B- Bag; T – Torso
- 7 sensor modalities, 21 data channels (motion and pressure)
- Data size
 - 272x4h training data (B/T/Hi/Ha);
 - 40x4h validation data (B/T/Hi/Ha);
 - 160x1h testing data (Hips)



Modality	File	Train (B/T/Hi)	Validation (B/T/Hi/Ha)	Test (Ha)
Accelerator	Acc x.txt	✓	✓	✓
	Acc y.txt	✓	✓	✓
	Acc z.txt	✓	✓	✓
Gyroscope	Gyr x.txt	✓	✓	✓
	Gyr y.txt	✓	✓	✓
	Gyr z.txt	✓	✓	✓
Magnetometer	Mag x.txt	✓	✓	✓
	Mag y.txt	✓	✓	✓
	Mag z.txt	✓	✓	✓
Linear accelerometer	LAcc x.txt	✓	✓	✓
	LAcc y.txt	✓	✓	✓
	LAcc z.txt	✓	✓	✓
Gravity	Gra x.txt	✓	✓	✓
	Gra y.txt	✓	✓	✓
	Gra z.txt	✓	✓	✓
Orientation	Ori w.txt	✓	✓	✓
	Ori x.txt	✓	✓	✓
	Ori y.txt	✓	✓	✓
Pressure	Pressure.txt	✓	✓	✓
	Label.txt	✓	✓	✗

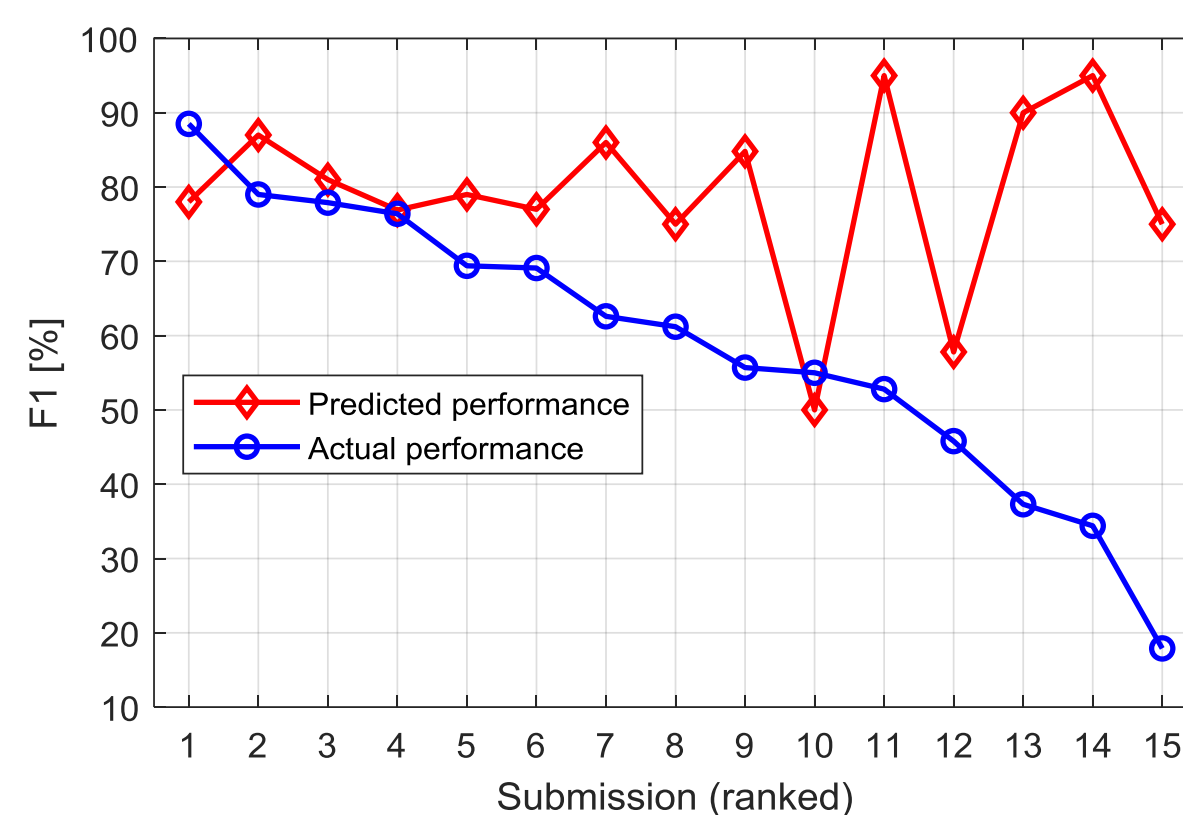


Evaluation Measure

- F1 score averaged among all the classes
- Decision window: 5 seconds

2. Submission results

- Strong interests from research community
 - 33 teams expressed interest
 - 15 submissions
- Ranking (F1 for testing data)
 - Highest F1: 88.5%
 - 80%-90%: 1 submission
 - 70%-80%: 3 submissions
 - 60%-70%: 4 submissions
 - < 60%: 7 submission



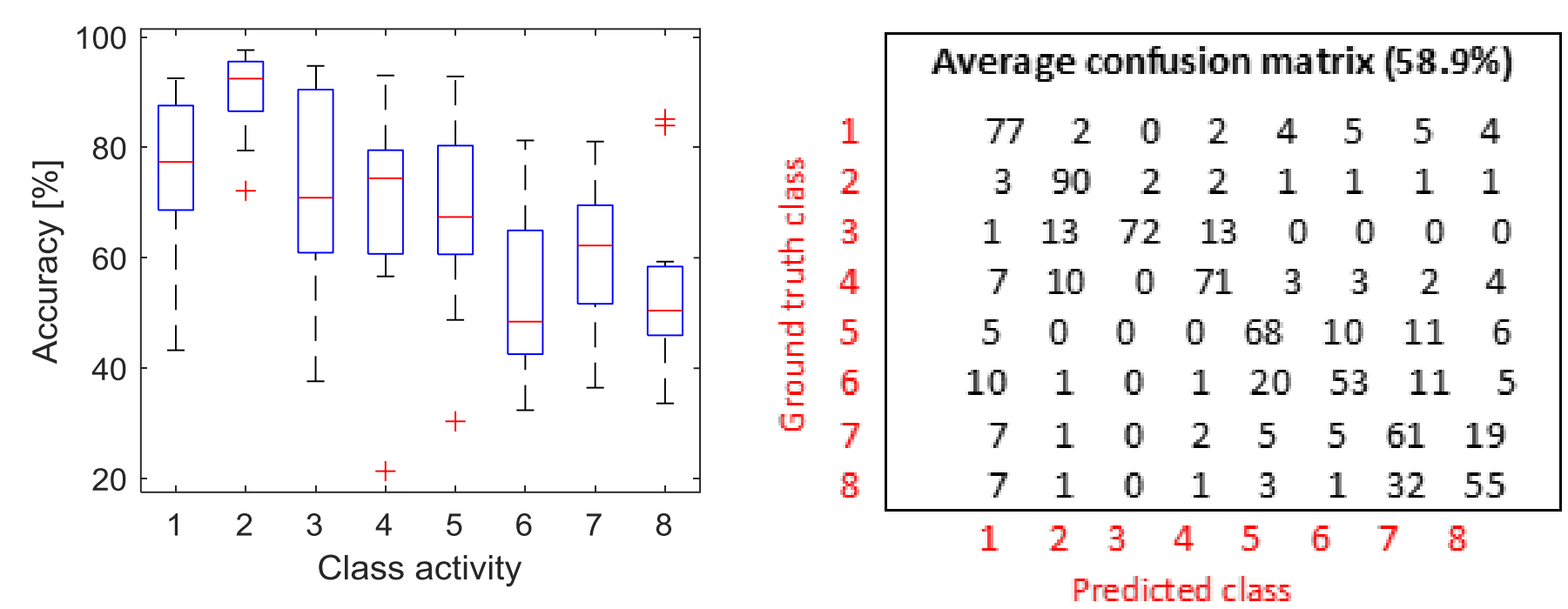
The detailed confusion matrices will be published at the summary paper.

5. Reference

- H. Gjoreski, M. Ciliberto, L. Wang, F. J. O. Morales, S. Mekki, S. Valentin, D. Roggen, "The University of Sussex-Huawei locomotion and transportation dataset for multimodal analytics with mobile devices," *IEEE Access*, 2018.
- L. Wang, H. Gjoreski, M. Ciliberto, S. Mekki, S. Valentin, D. Roggen, "Enabling reproducible research in sensor-based transportation mode recognition with the Sussex-Huawei dataset," *IEEE Access*, 2019.
- L. Wang, H. Gjoreski, K. Murao, T. Okita, D. Roggen, "Summary of the Sussex-Huawei locomotion-transportation recognition challenge", *Proc. HASCA 2018*.
- L. Wang, H. Gjoreski, M. Ciliberto, S. Mekki, S. Valentin, D. Roggen, "Benchmarking the SHL recognition challenge with classical and deep-learning pipelines", *Proc. HASCA 2018*.
- L. Wang, H. Gjoreski, M. Ciliberto, P. Lago, K. Murao, T. Okita, D. Roggen, "Summary of the Sussex-Huawei locomotion-transportation recognition challenge 2019", *Proc. HASCA 2019*.

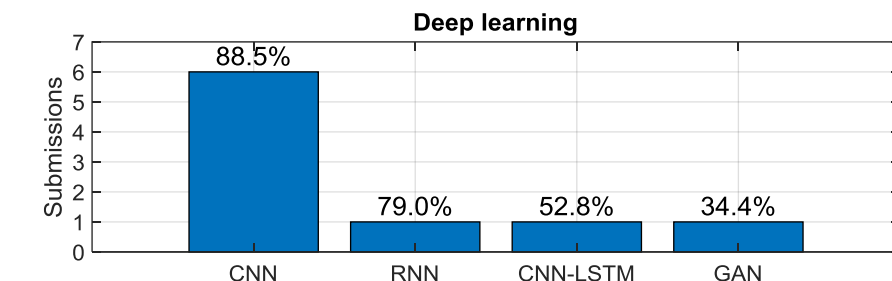
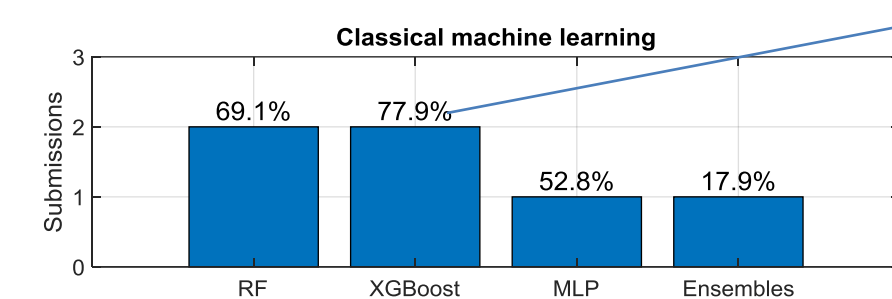
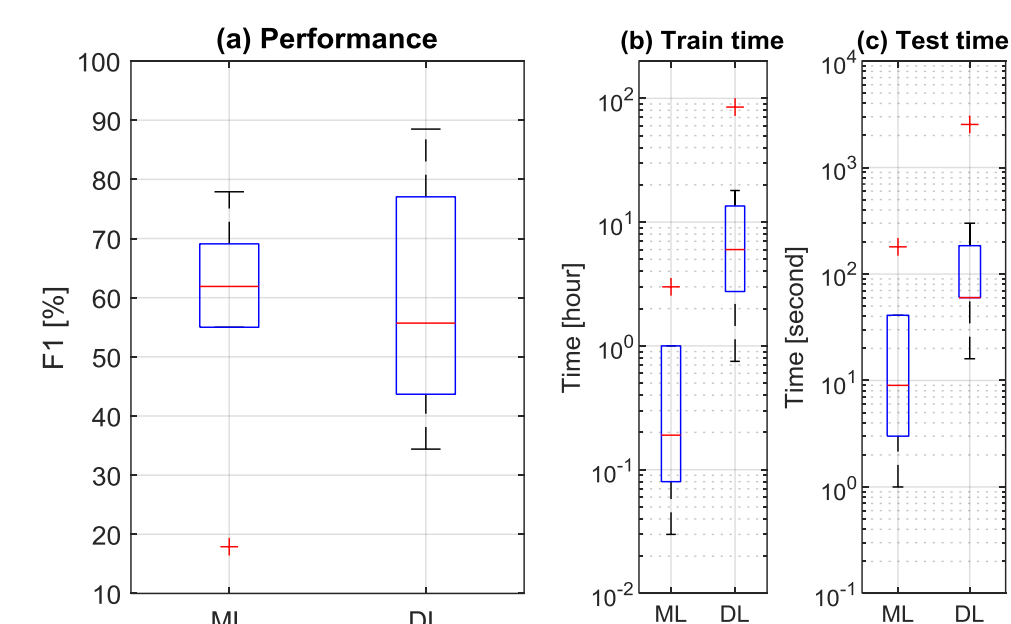
3. Analysis

- Average performance from top11 submissions (F1>50%)



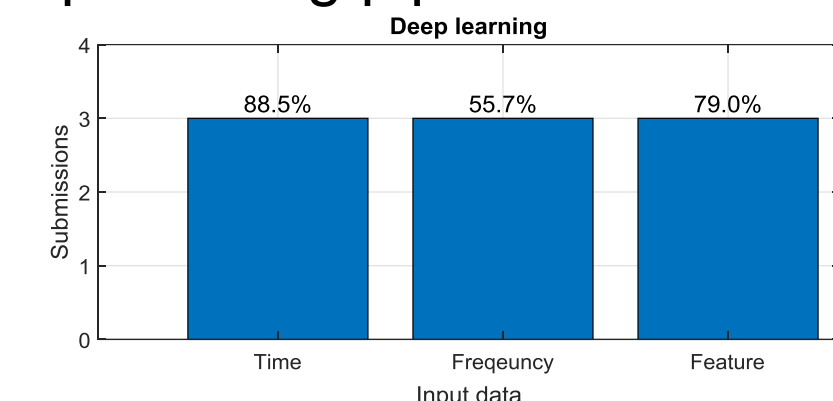
Machine learning pipelines

- Classical machine learning (ML): 6 submissions
- Deep learning (DL): 9 submissions

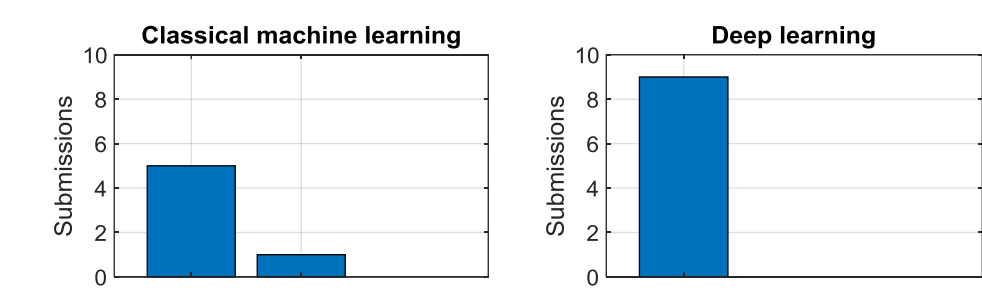


Highest performance in each group

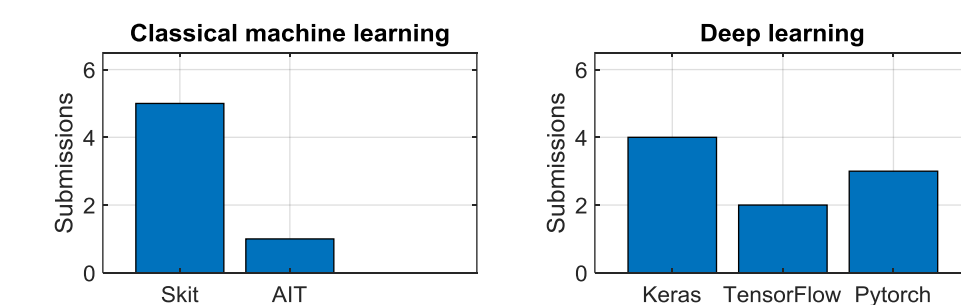
Input to deep learning pipeline



Implementation language



Implementation framework



4. Conclusions

- Highest F1 score: 88.5%
- DL outperforms ML
- Novel techniques to tackle position and user variation
 - Robust representation
 - Position-specific modelling
 - User-specific modelling
- DL outperforms ML
- Future work
 - Comprehensive evaluation with full dataset
 - Multimodal sensor