

What Are You Doing? Real-Time Activity Recognition using Mobile Phone Sensors

Bernhard Hiesl, <u>Marc Kurz</u>, Erik Sonnleitner Adaptive 2020 | October 2020 marc.kurz@fh-hagenberg.at

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Introduction & Goals

- Follow-up of work-in-progress paper from last year [1]
- Major Challenge/Goal: detection of people's activities regardless of the mobile phone's position & orientation
 - > 5 modes of locomotion considered: (i) standing, (ii) walking, (iii) running, (iv) walking upstairs, (v) walking downstairs

How can highly accurate real-time activity recognition be realized utilizing a dynamically (i.e. rotation, position and orientation independent) on-body-placed commercial smartphone?



Previous Work

- Data collection
 - > 15 participants, ~4,5hrs data
- Feature Extraction
 - > Sliding window, Savitzky-Golay-Filter
- Classification





Approach

- General Methodology [1]

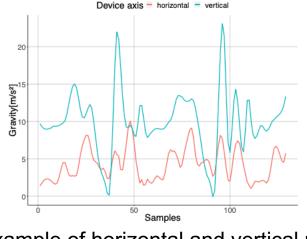




Approach :: Real-Time

- Sliding Window [2]





Example of horizontal and vertical movement within one single window

[2] L. Sun, D. Zhang, B. Li, B. Guo, and S. Li, "Activity recognition on an accelerometer embedded mobile phone with varying positions and orientations," in International conference on ubiquitous intelligence and computing. Springer, 2010, pp. 548–562.

Approach :: Dynamic Placement

- i. Sensor recordings of activities within several body positions
 - > Gathering of position independent data
 - > 5 pocket positions considered [1]
- ii. Sensor data transformation (horizontal & vertical)
 - > Becoming orientation independent

iii. Training of ML models

> kNN, Naive Bayes, Decision Tree, Random Forrest



Results :: Overview

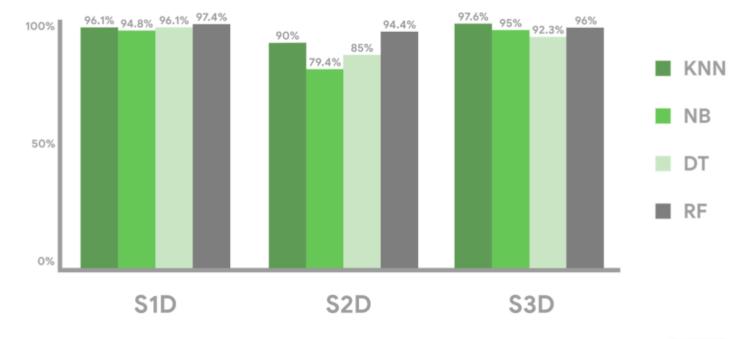
- Three different testing scenarios were applied
 - Environment changes (e.g. different stairs, different floor surfaces, etc.) S1
 - > Varying performing speed
 > Changing positions
 S3

- Each testing scenario is evaluated using sensor data:
 - Solution > Gallery dependent (subject within the training set)
 - Sallery independent (subject not inside the training set)

e.g. S1D means environment with gallery dependent subject...

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Results :: Offline* Evaluation

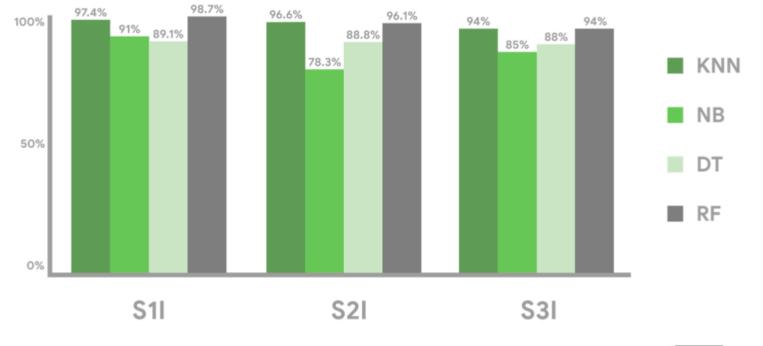






* using newly recorded data and performing recognition offline

Results :: Offline* Evaluation

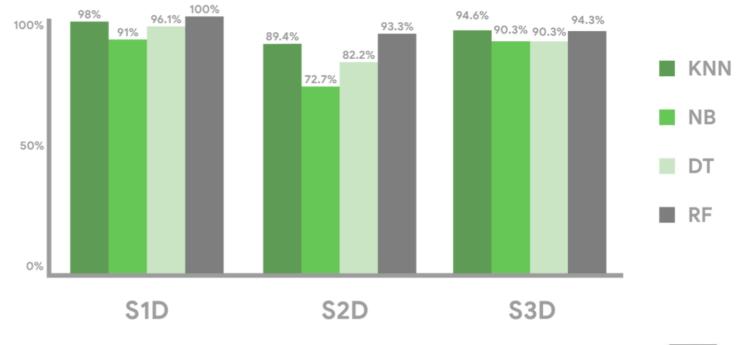




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* using newly recorded data and performing recognition offline

Results :: Online* Evaluation







* using sensor data processed in real-time on the smart-phone

Results :: Online* Evaluation



* using sensor data processed in real-time on the smart-phone

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Summary & Conclusion

- Developing a gait recognition system which is position and orientation independent
 - > Could be beneficial for different commercial use-cases, e.g.
 - Physiotherapy, elderly care, industrial manufacturing, etc.
- Performance up to 96.1% accuracy
- Real-Time processing of sensor data
- Detection of significant features from the feature vector (currently 177 features) could be improved
 - > E.g. by using a grid search algorithm
 - > Would most likely **further improve** the accuracy of the system





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