

Data integration and Learning

With thanks to: Peter Flach, Raul Santos-Rodriguez, Alessandro Masullo
Miquel Perello-Nieto, Haixia Bi, Emma Tonkin, Taku Yamagata

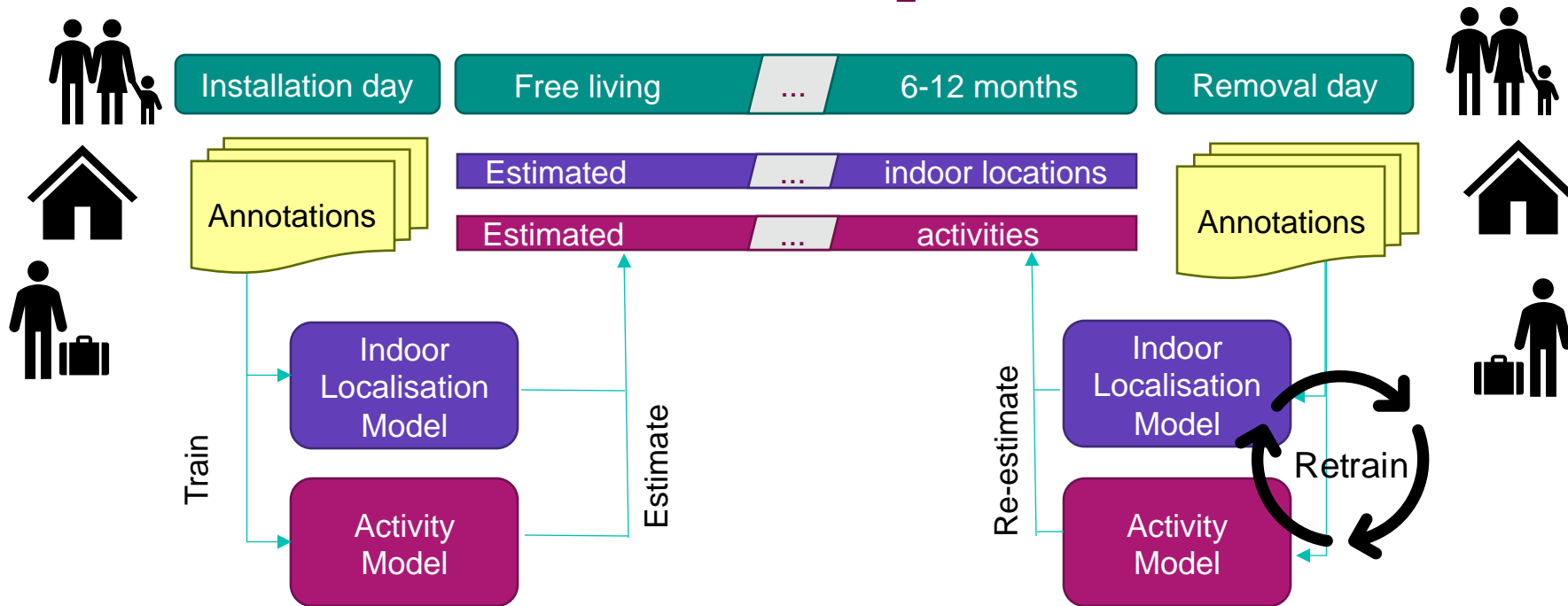
In This Section

1. Installation, annotation and training overview
2. Indoor localisation
3. Activity recognition
4. Case study
5. Other examples

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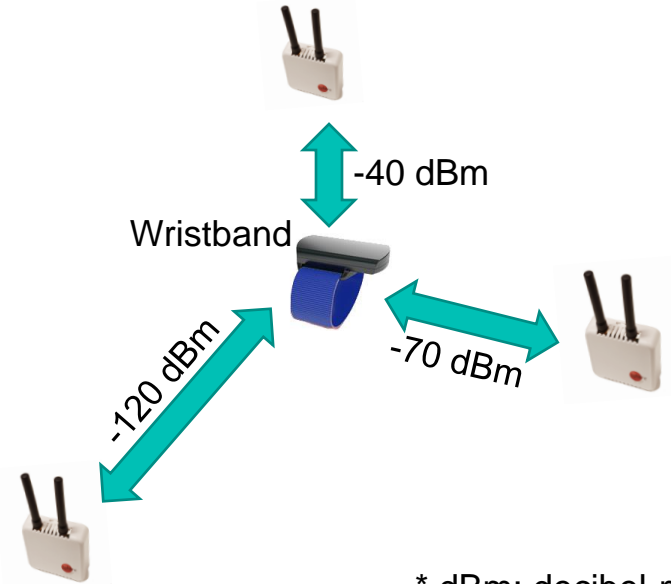
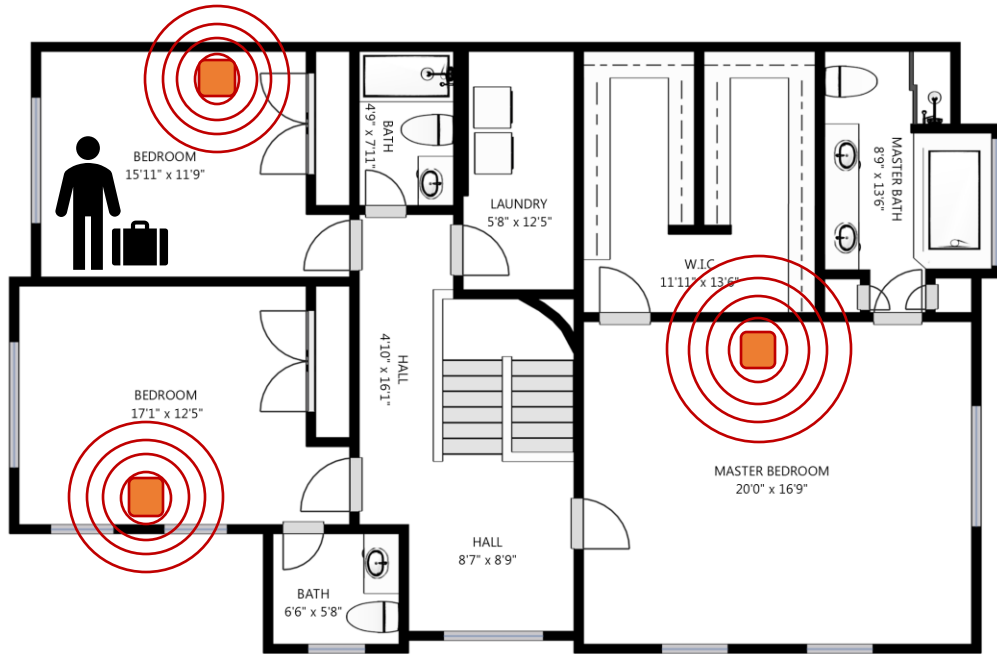
Installation and annotation process



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Localisation from Received Signal Strength



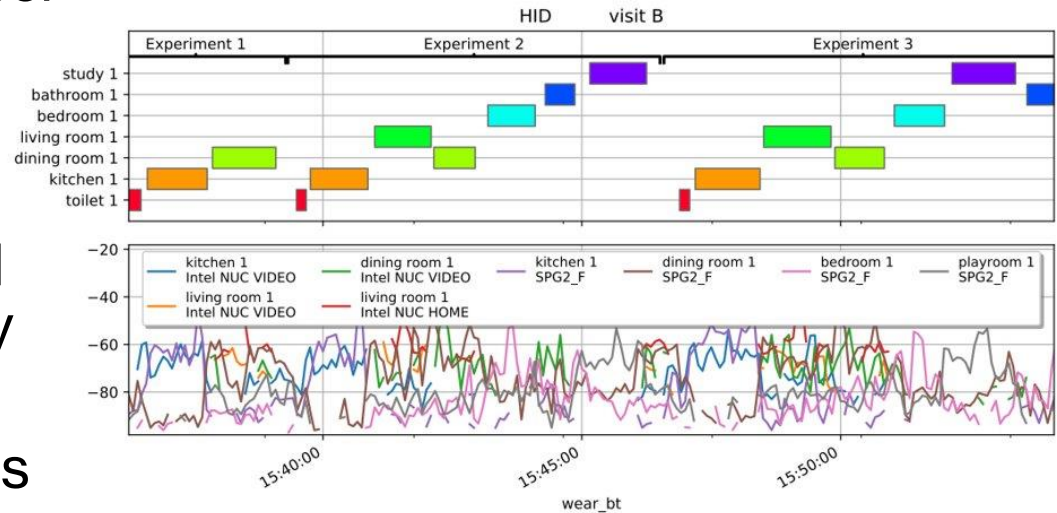
* dBm: decibel-milliwatts

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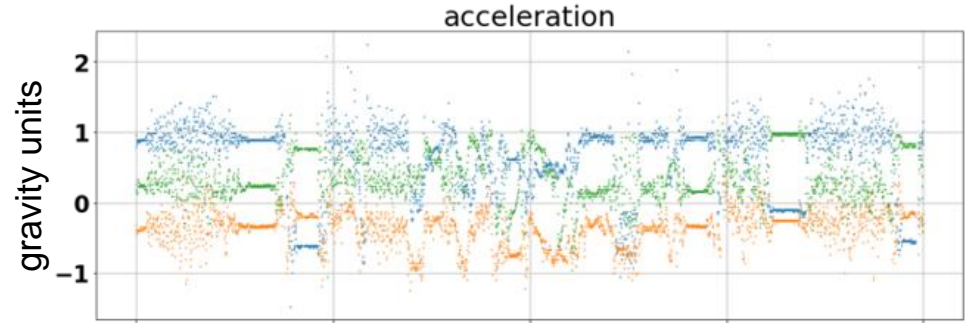
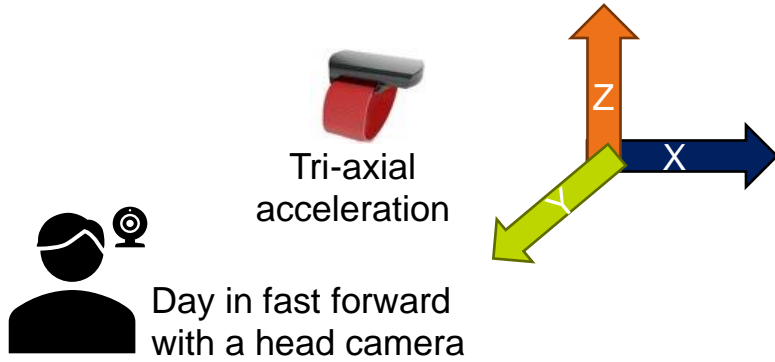
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Example of location annotations

- At least 2 experiments per visit
- Visit every room and annotate the location
- We record the Received Signal Strength Intensity
- Train model and use to estimate indoor locations



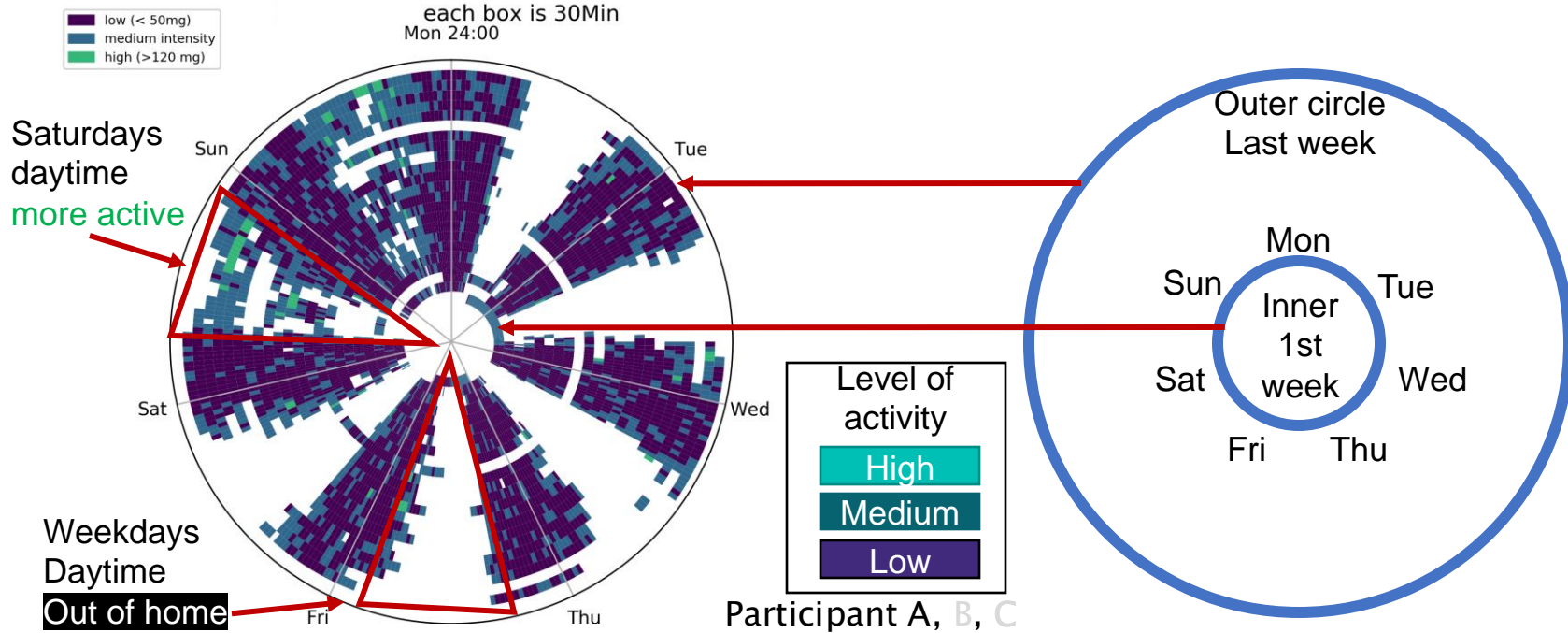
Activity levels and recognition



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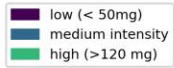
Activity levels for the full period



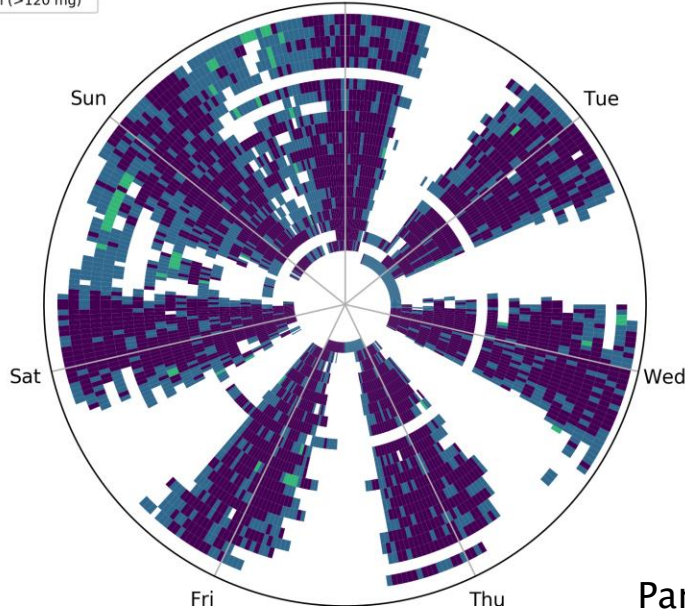
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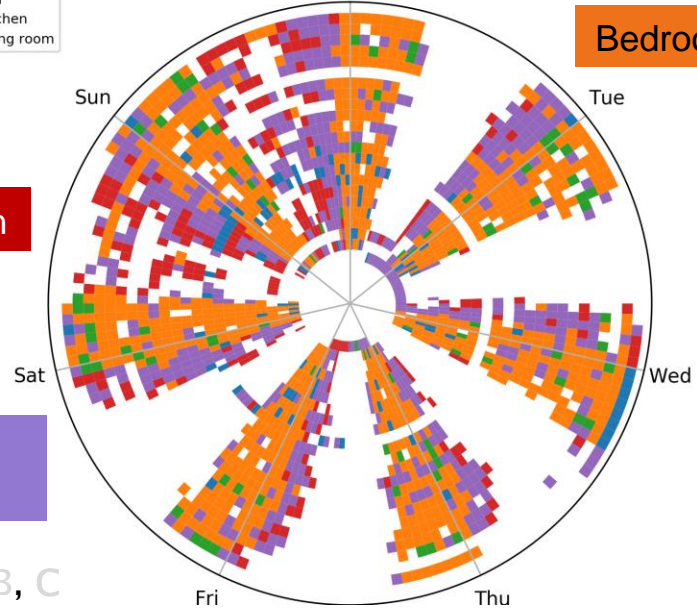
Activity levels and localisation predictions:



each box is 30Min
Mon 24:00



each box is 1H
Mon 24:00



Kitchen

Living room

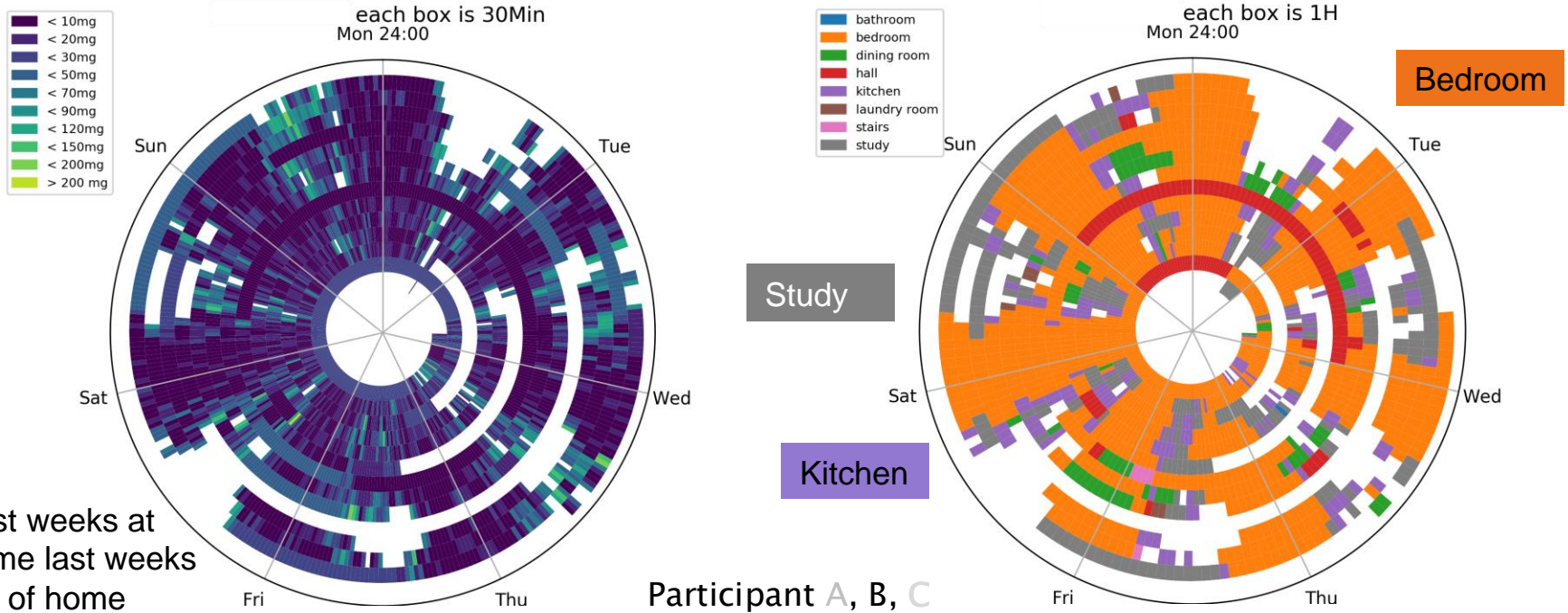
Bedroom

Participant A, B, C

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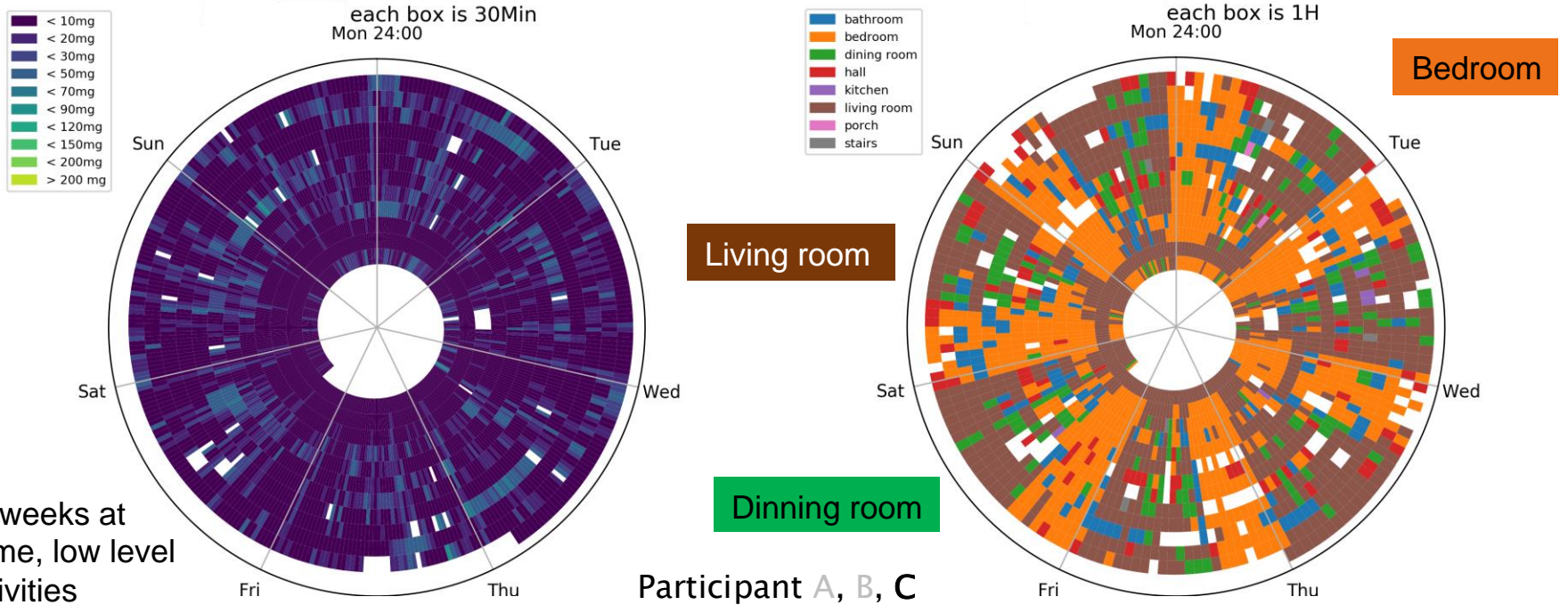
Activity levels and localisation predictions:



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Activity levels and localisation predictions:

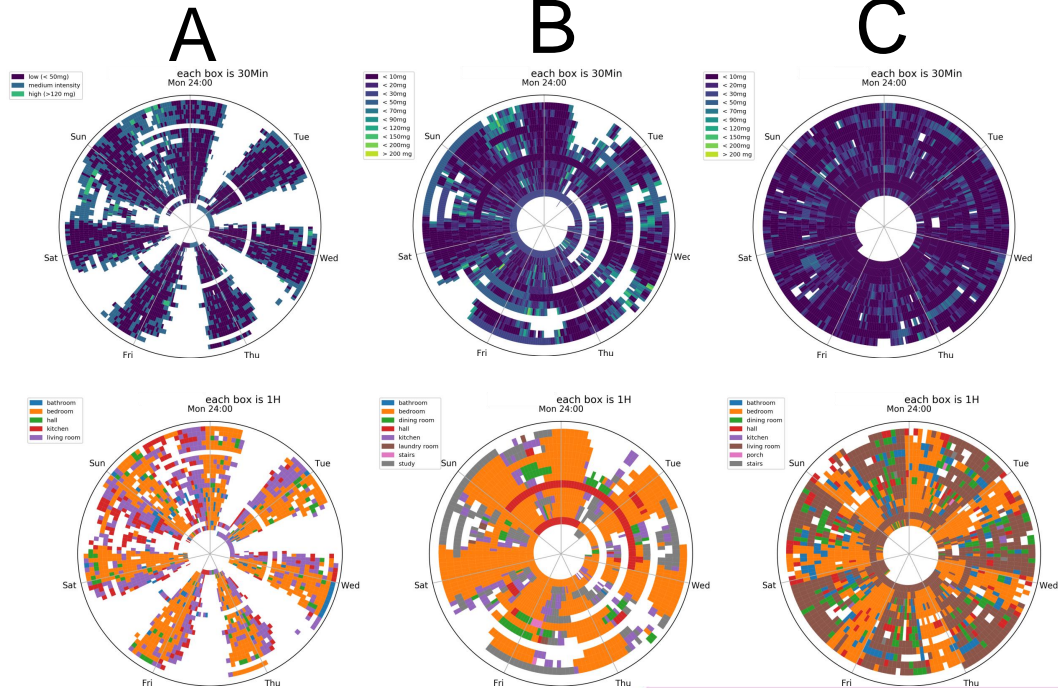


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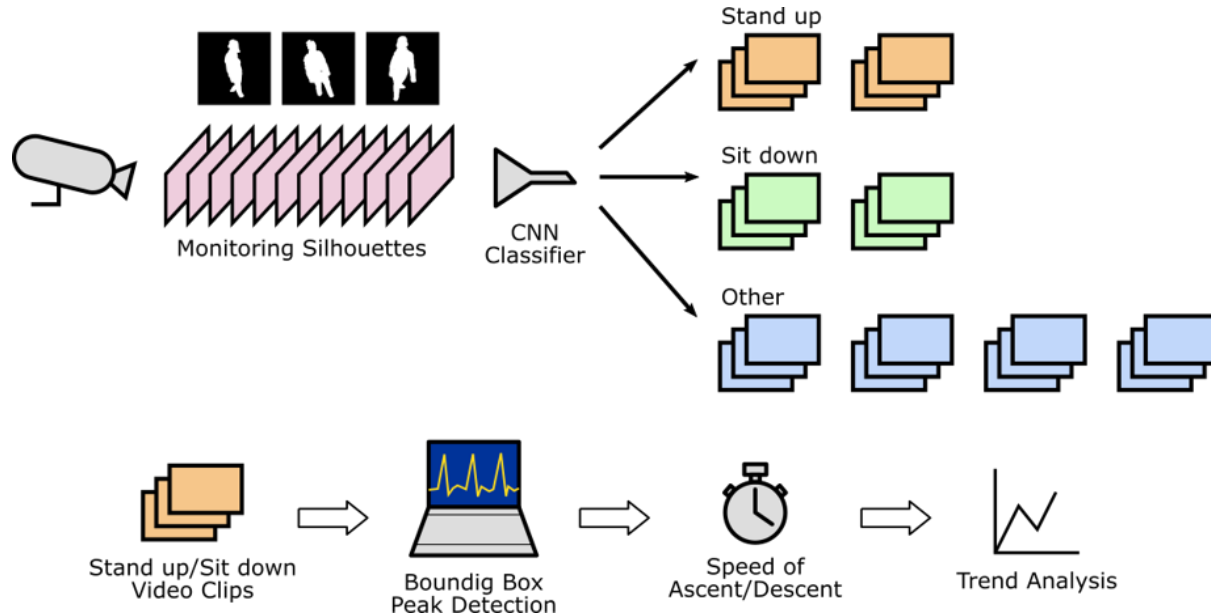
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Activity and Locations overview

- A is a participant from the control group
- B and C are recovering from a hip replacement surgery
- B and C spend more time at home, and are generally less active

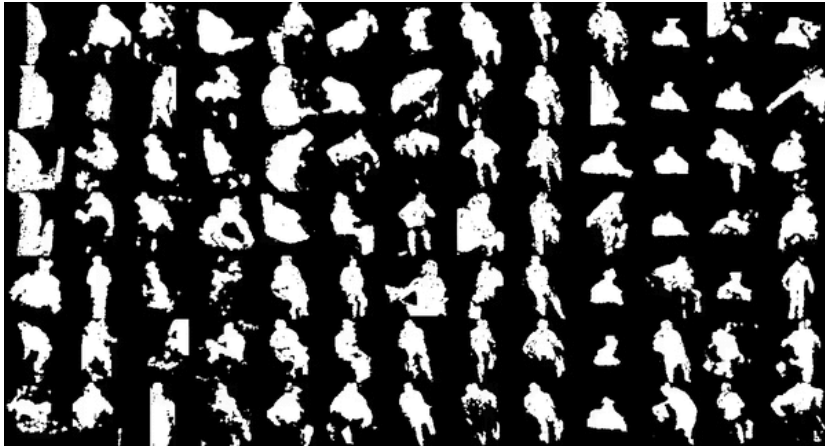


Sit to Stand as a surrogate of recovery

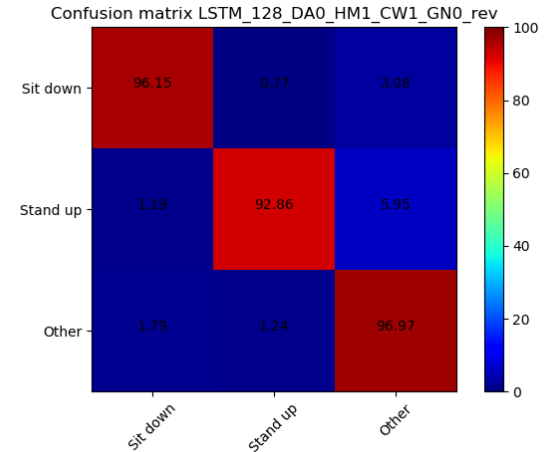


Sit to Stand Detector (Classifier)

- Using MuViLab, we have annotated 4 months of video data for STS transitions.
- Using a convolutional neural network we were able to achieve >90% accuracy in classifying video sequences:

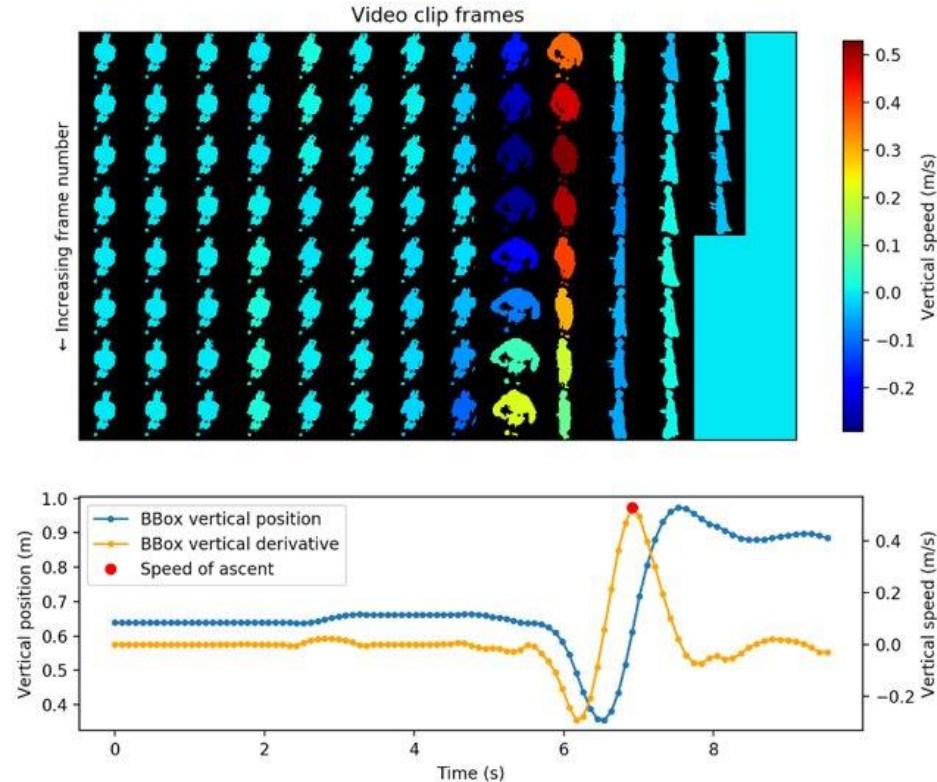


- Examples of 4 seconds video sequences detected as stand up transitions.



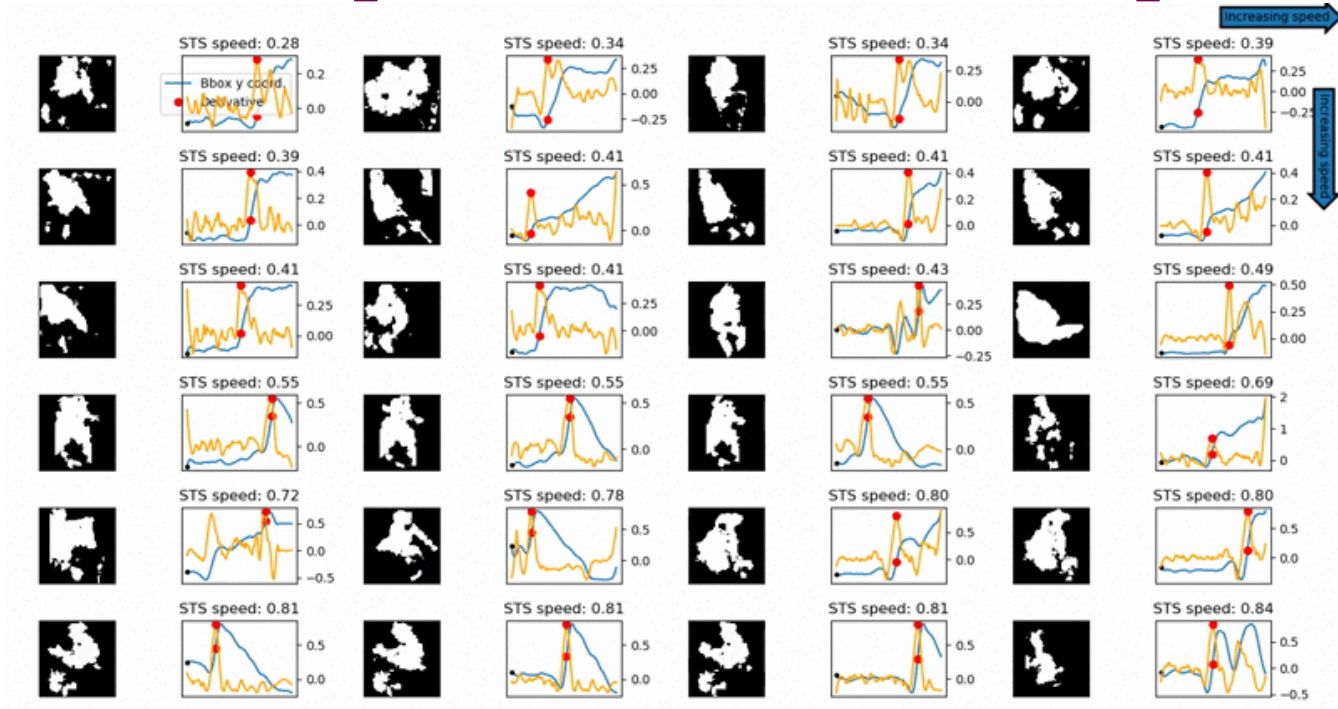
Bounding box peak detection

- The figure shows an example of analysis of sit-to-stand transition using silhouettes.
- On top are the frames of the video sequence of a person standing up, on the bottom the analysis derived from their movement.
- The red dot shows the speed of ascent, i.e. the maximum transfer velocity from sitting to standing position.



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Several Examples of Sit to Stand Speed

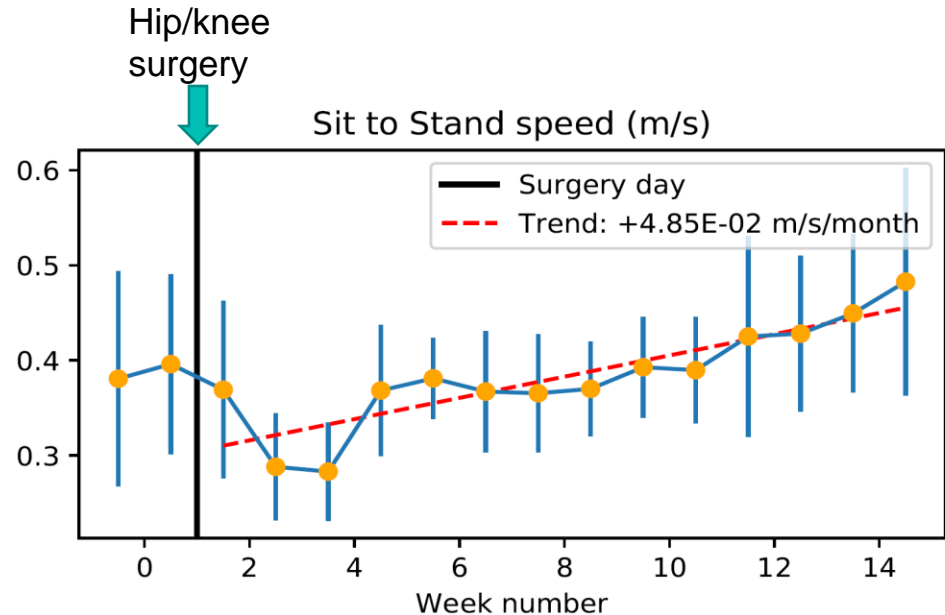


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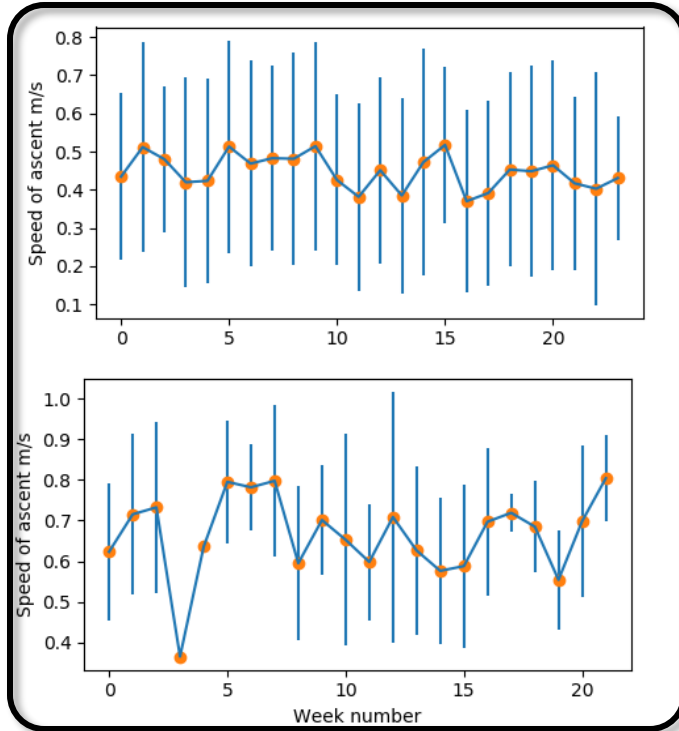
What Does This Look Like for a Patient?

- Recover from hip/knee replacement
- Sit to Stand speed decreases abruptly after surgery
- After that, the speed gradually increases

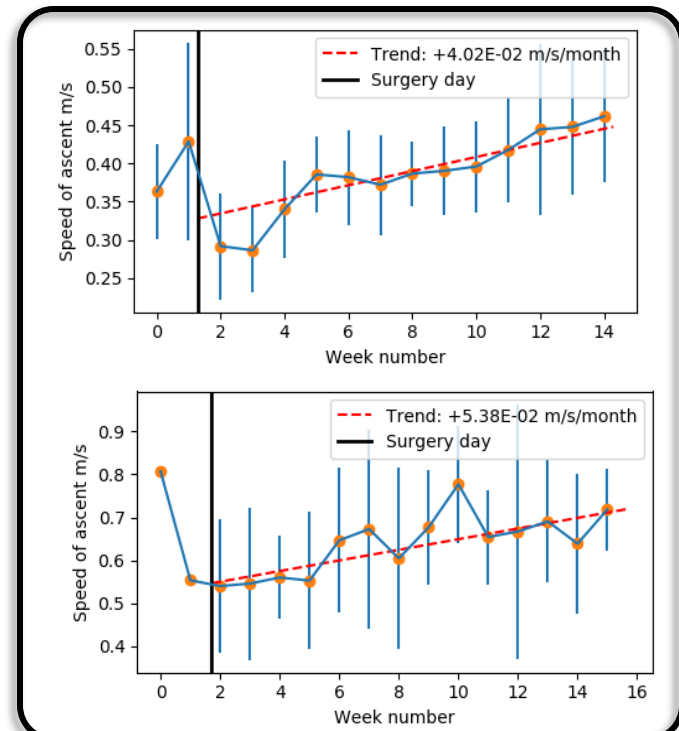


Comparison With Healthy Control

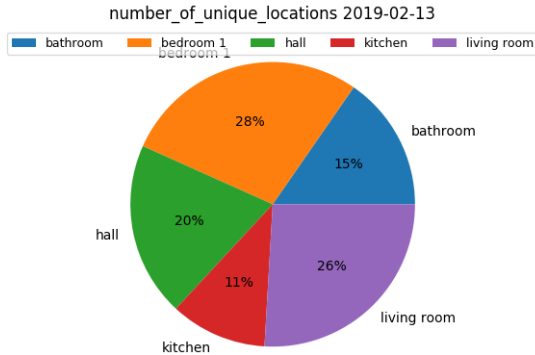
Healthy control



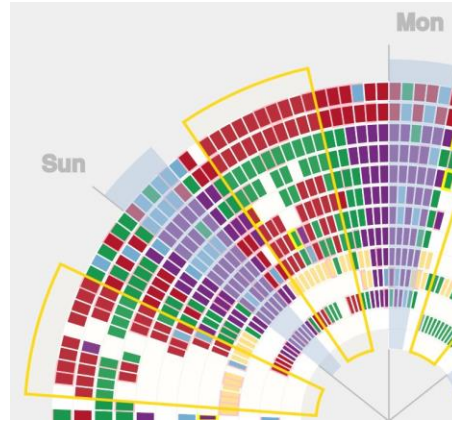
Surgery patients



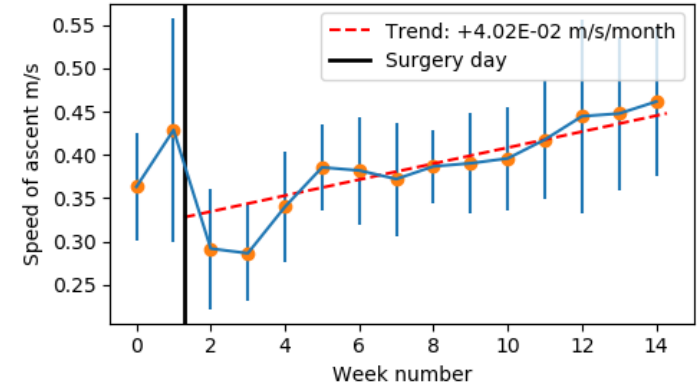
Other examples of available data



- Proportion of time spent in each room in one day (or several)
- Or Number of transitions between rooms



- Actigraphy for the study of sleep and circadian rhythms. Cut out of a circular plot showing the 10 most active hours (yellow box) and 5 less active (light blue) [2]



- Speed of ascent from sit to stand for a patient recovering from a hip replacement [3]

References

- [1] M. Holmes, H. Song, E. Tonkin, M. P. Nieto, S. Grant, and P. Flach, “**Analysis of patient domestic activity in recovery from hip or knee replacement surgery: Modelling wrist-worn wearable RSSI and accelerometer data in the wild,**” in CEUR Workshop Proceedings, 2018, vol. 2148.
- [2] M. Holmes, M. P. Nieto, H. Song, E. Tonkin, S. Grant, and P. Flach, “**Modelling Patient Behaviour Using IoT Sensor Data: a Case Study to Evaluate Techniques for Modelling Domestic Behaviour in Recovery from Total Hip Replacement Surgery,**” J. Healthc. Informatics Res., vol. 4, no. 3, pp. 238–260, Sep. 2020.
- [3] Masullo, Alessandro, et al. “**No Need for a Lab: Towards Multi-sensory Fusion for Ambient Assisted Living in Real-world Living Homes.**” VISIGRAPP (5: VISAPP). 2021.

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