



Unobtrusive and Personalised Monitoring of Parkinson's Disease Using Smartphones

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It's all about people

You might know them









and another 5.2 million...

Technology-based monitoring



In the lab



Uncomfortable



Motor Focused



Short & Sporadic



Intrusive



Population based

The potential of the digital world





Personal



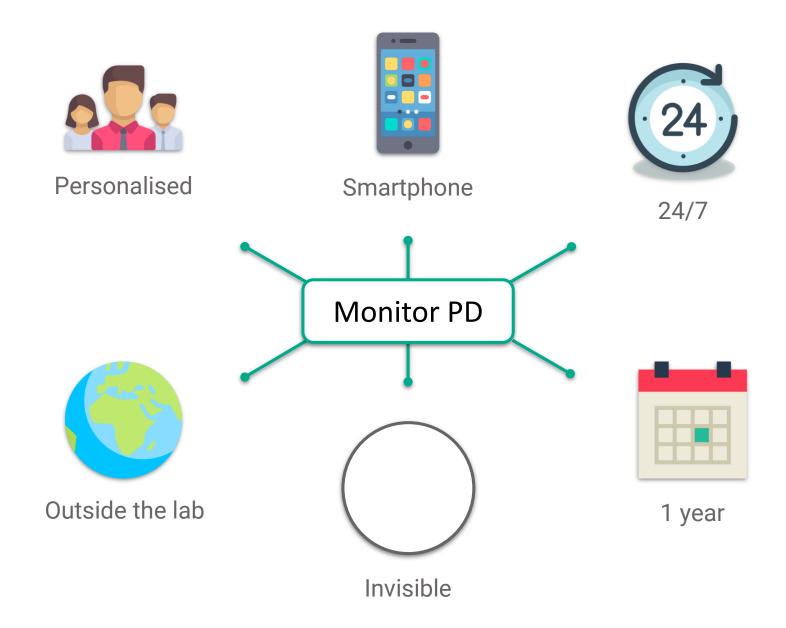


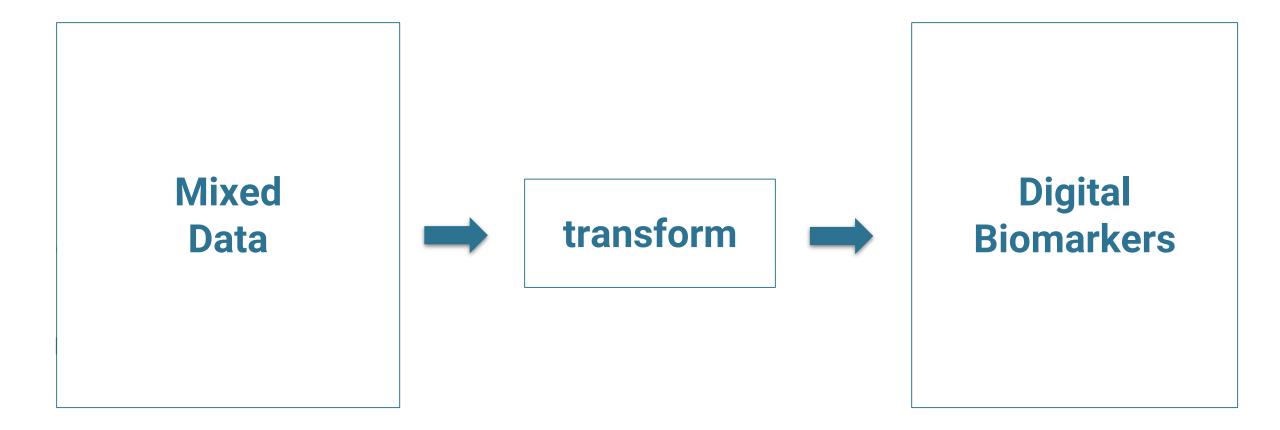
Internet

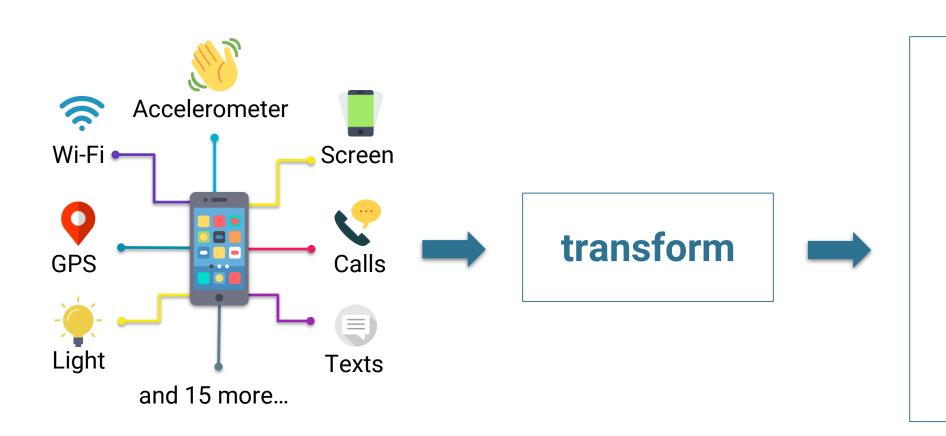


Ambient

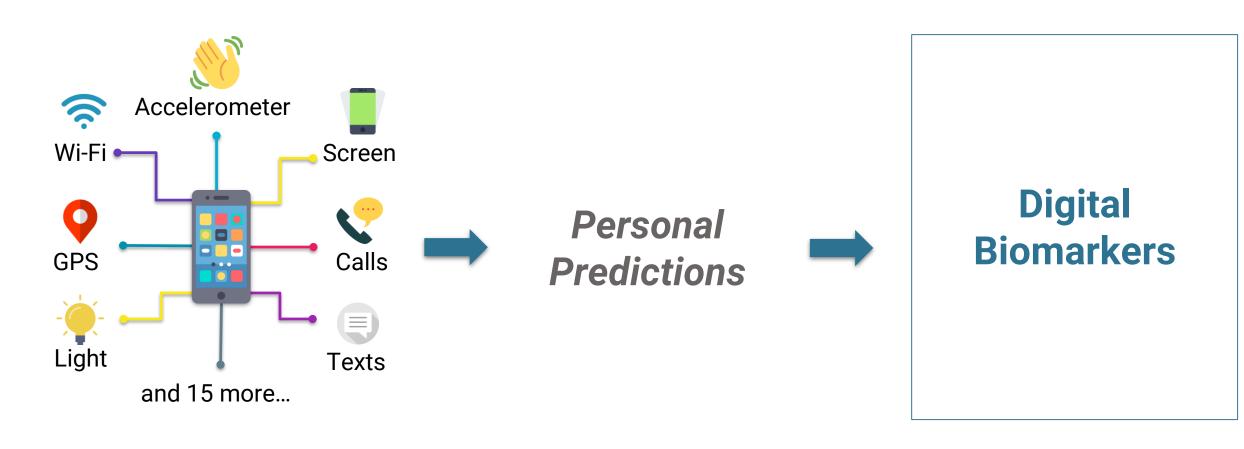
Our goal

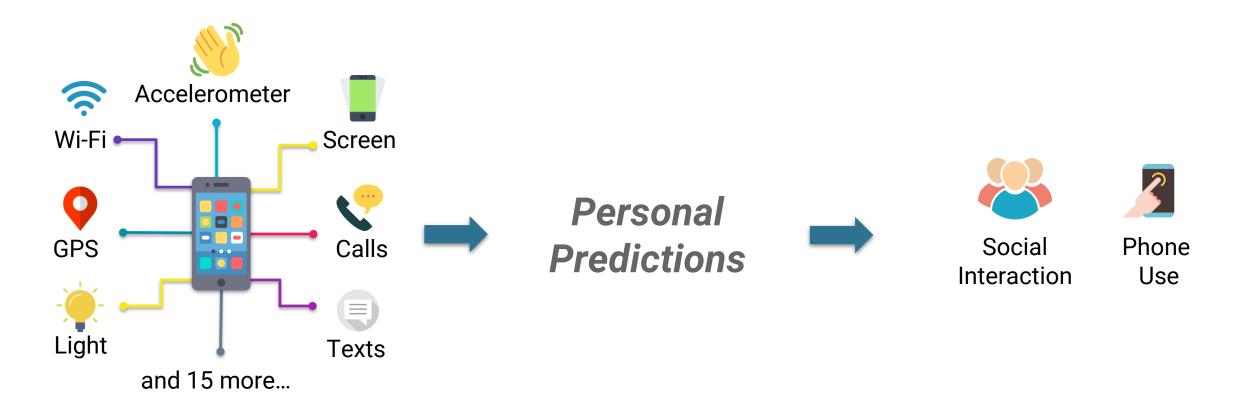


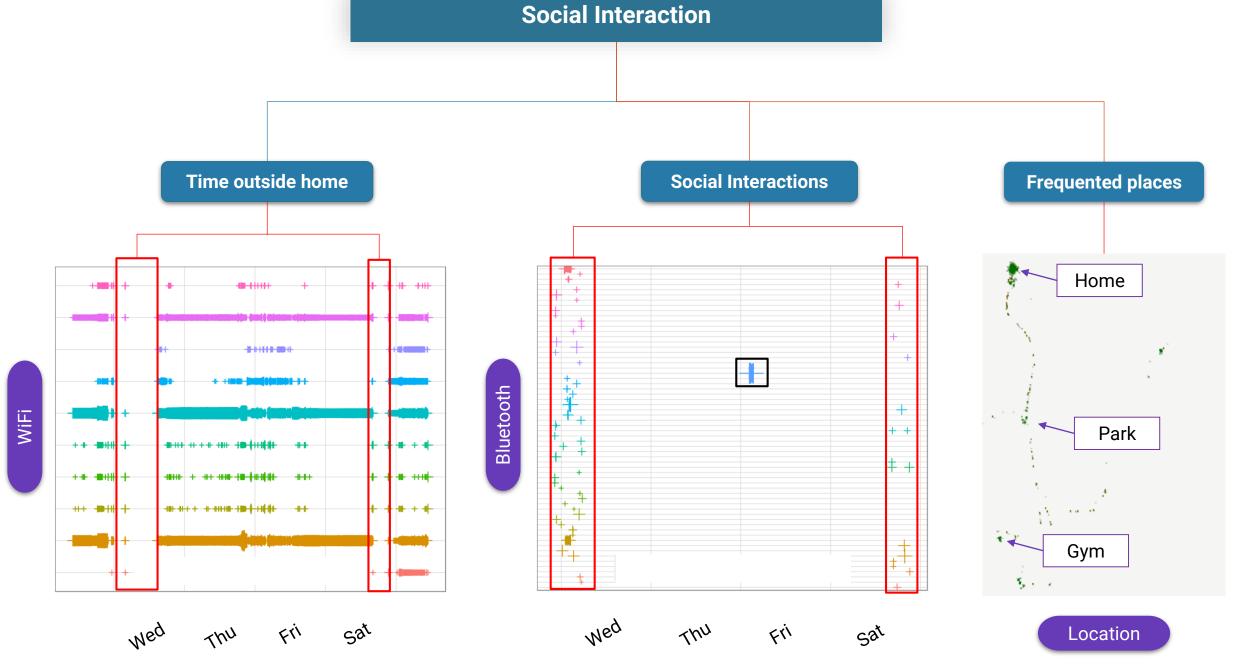




Digital Biomarkers







Our protocol







11 participants

22 sensors per participant

6-week visits

and nine months to get ethical approval...

Validation

Compare smartphone inferences vs symptom changes



Every 6 weeksClinical Scores



Every daySelf-Reporting

Start designing, implementing, executing, early

Go tech!







But tech can fail...

Trying even more tech



Cube with NFC tags

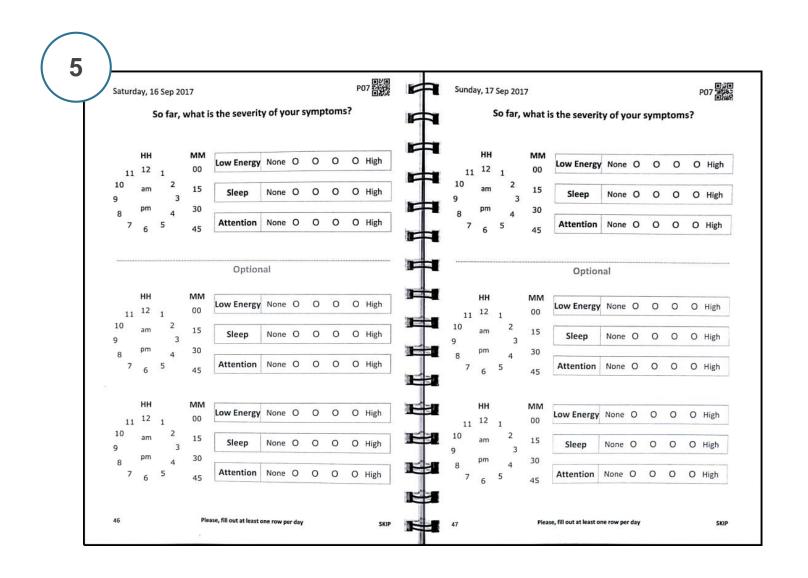
Trying even more tech

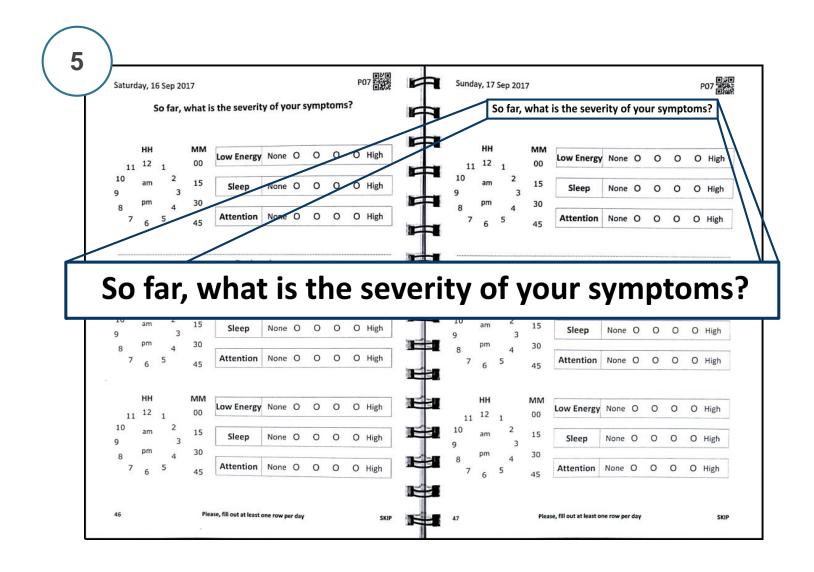


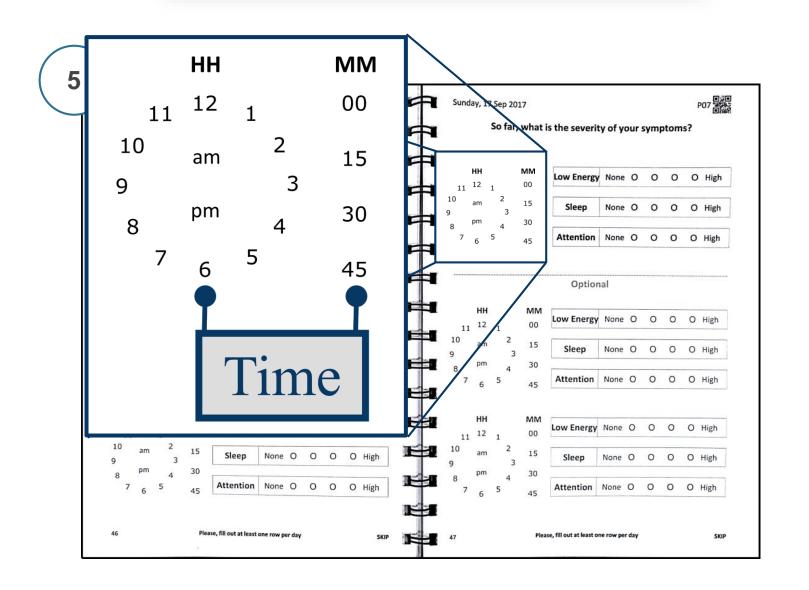
Micro:Bit

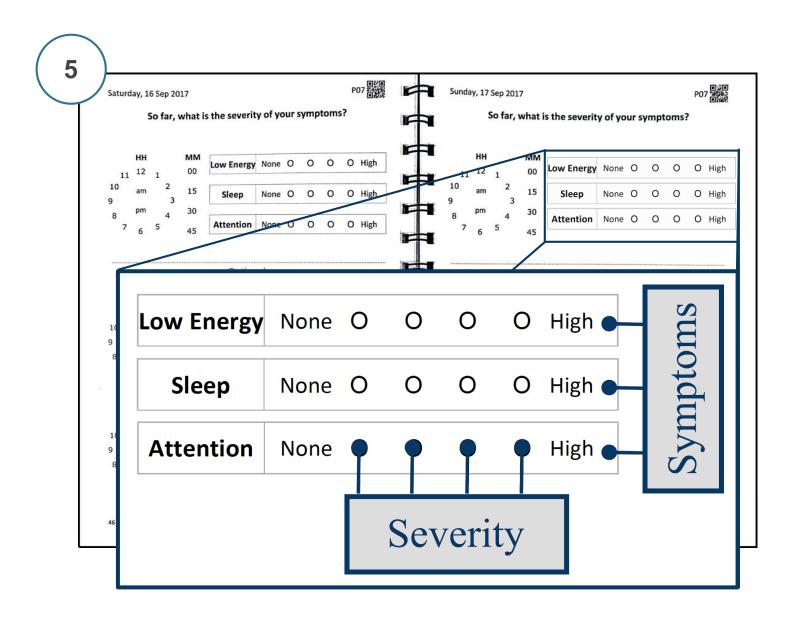
BRCK-IIII TO PALOGUE

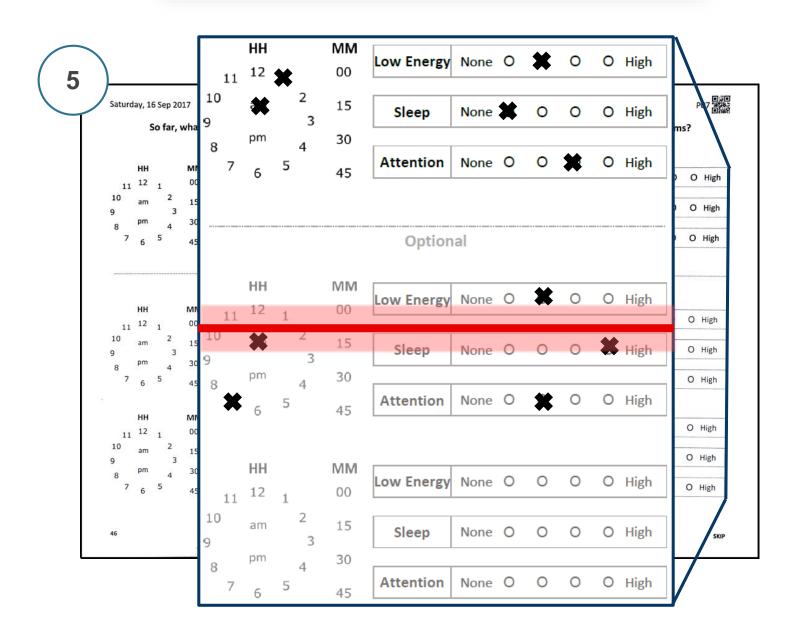




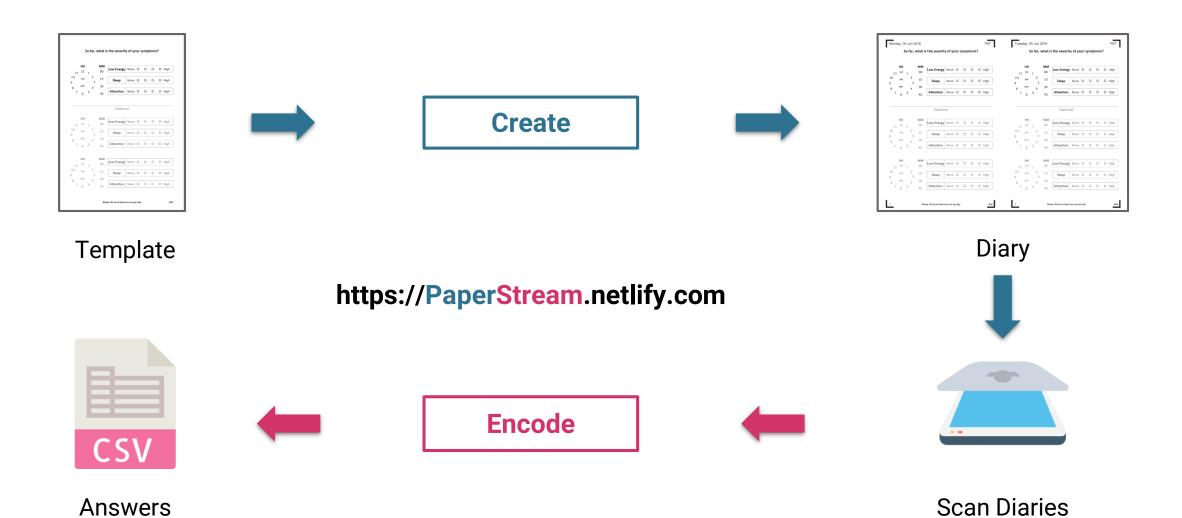








Create and encode paper diaries or surveys



Answers

Did the diary work?



7 participants



~380 days (at least once a day)



Answer rate 96%

Why 96%? It was simple, flexible, and physical

Put your users first

Data Analysis

(a.k.a. data cleaning)

Key factor

Personalisation



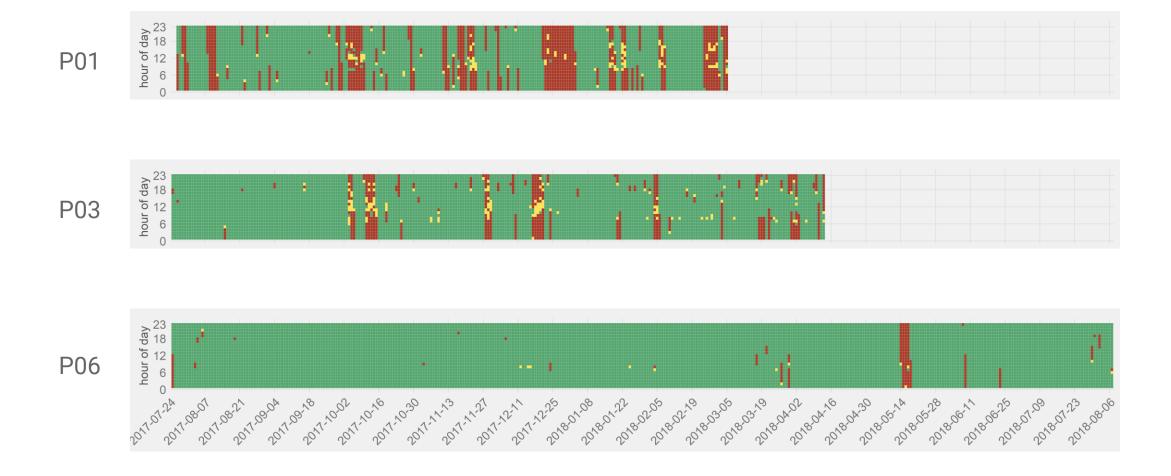
Precision Medicine

Generalisation



Population Methods

Smartphone data



Diary data



Our data

Smartphone data

- 14 location (Canzian '15, Barnett '18)
- 2 Activity Recognition (Google API)
- 4 calls
- 2 screen
- 2 sleep
- 1 Wi-Fi
- 1 light, 2 SMS, 5 keyboard (only Android)

Groundtruth data

- Daily self-reported top 3 personal symptoms
- Clinical (every 6 weeks)
 - MDS-UPDRS: Gold standard
 - PDQ-39: Quality of life
 - NMS: Non-motor symptoms
 - ACE: cognition
 - Six laptop-based cognitive tests

Be aware of the benefits and limitations of technology

The bottleneck

... a main bottleneck in the current digital phenotyping work is not due to technical challenges but more due to the lack of sufficient statistical methodology...

Lisa A. Marsh Opportunities and needs in digital phenotyping Nature Neuropsychopharmacology April '18

Personal Predictions

Create Prediction



2 Personalise Prediction

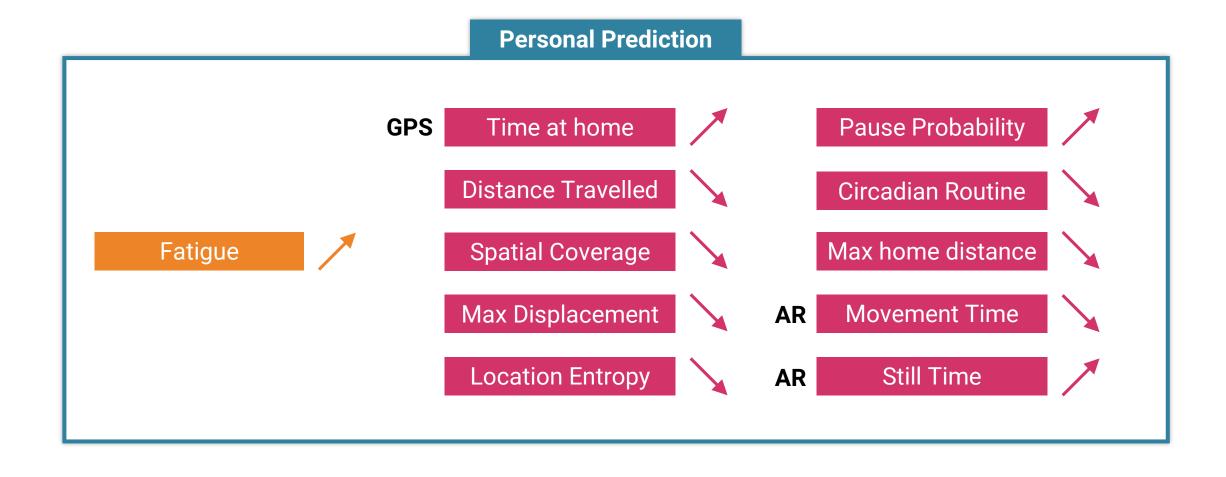


3 Test Prediction

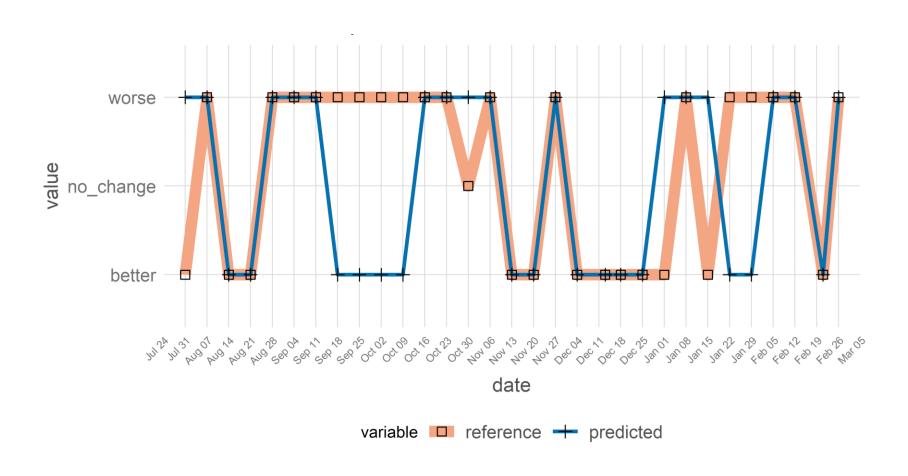
Define relation between smartphone features and human behaviour Adapt smartphone features to each person's symptom fluctuations

Compare adapted features to a random baseline

Example of prediction



What features are relevant to each person?

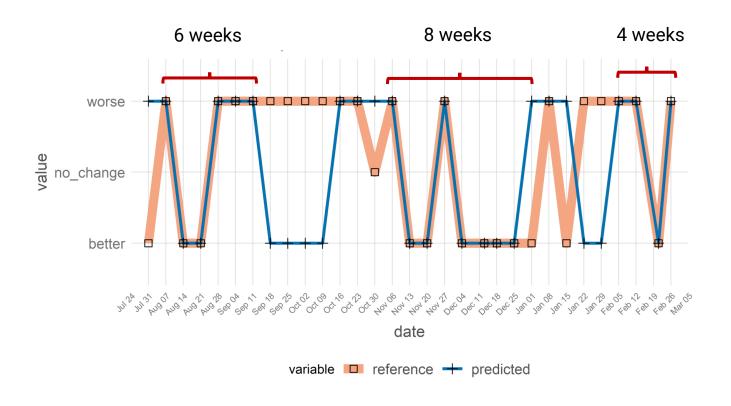


P01 - Fatigue

Personal Prediction on 100% of data

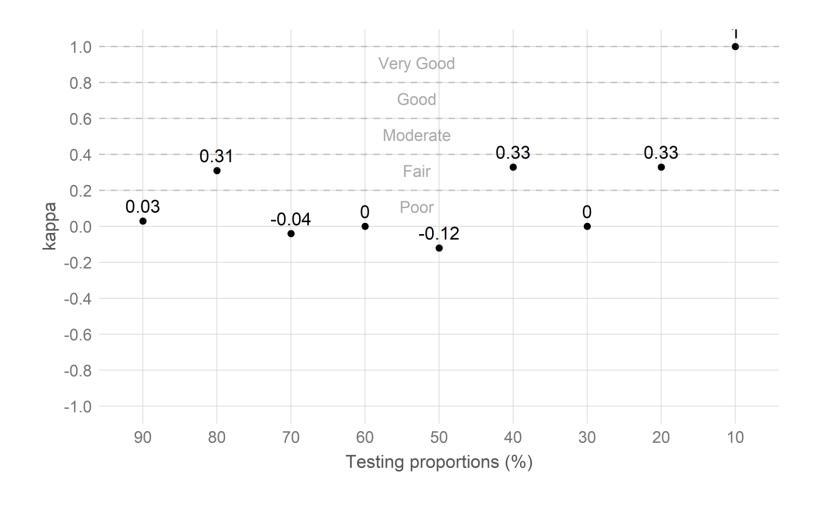
Participant	Groundtruth	weeks	Agreement
P01	Slow Walk	31	0.53
P01	Fatigue	31	0.37
P03	Low Energy	31	0.29
P03	Pain	31	0.35
P03	Freezing	31	0.23
P04	Fatigue	37	0.45
P06	Gait	42	0.18
P07	Low Energy	46	0.21
P08	Gait	43	0.34
P08	Pain	43	0.35

P01 Fatigue vs Max Home Distance

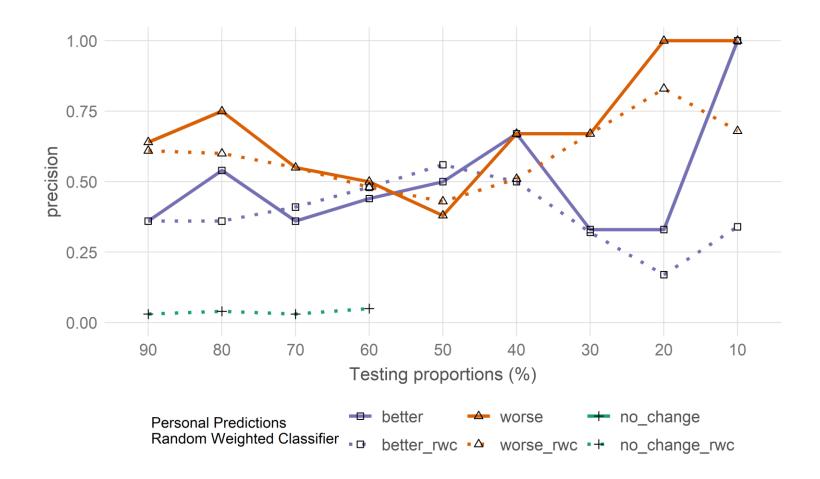


Class	Precision (diff over chance)	Recall (diff over chance)
worse	0.75 (+0.17)	0.67 (+0.09)
no change	0.03 (NA)	0.04 (-0.04)
better	0.60 (+0.22)	0.75 (+0.38)

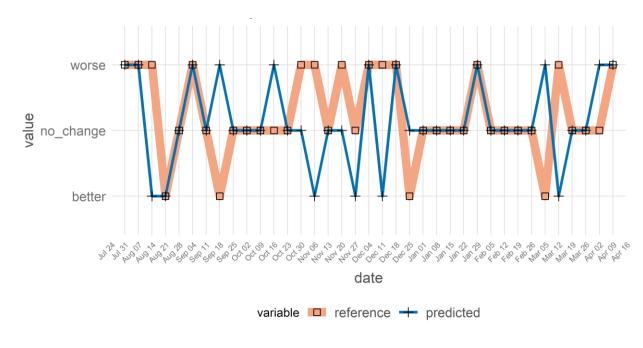
P01 Fatigue (agreement)



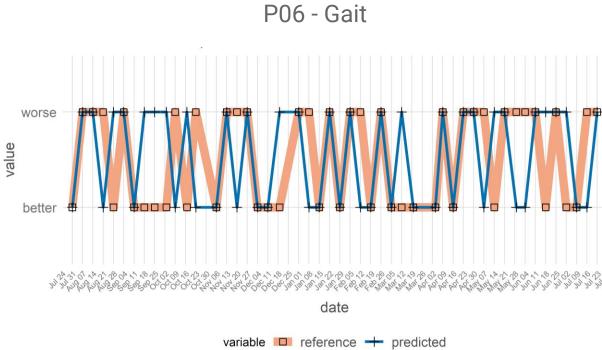
P01 Fatigue (precision vs chance)



Other participants



P04 - Fatigue



Insights

Our method **personalised** smartphone features to people's symptoms (not on unseen data)

Our method **adapted** to patients with different weekly trends

Our method tracked contiguous fluctuations (blocks of behaviour)

The personalisation/testing split is not suitable for a real-world deployment

Limitations (opportunities)



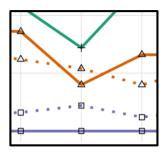
Validity of smartphone features



Self-reporting drawbacks



Confounding factors in people's behaviour



Clinical validity and usefulness of weekly trends

Creating new questions is as good as getting answers





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