



Equitable Photoplethysmography in Wearables: Accurate Data for All

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**Undiagnosed
disease in UK:**

**0.4
mill**

**atrial
fibrillation**

**1.4
mill**

**obstructive
sleep apnea**

**5.5
mill**

hypertension

Robles et al. *Int J Clin Cardiol*. 2018;137:1-7. <https://doi.org/10.1016/j.ijcc.2018.05.018>
Blipbeat. *Heart*. 2018;104(10):955-961. <https://doi.org/10.1136/heart-2017-012118>





St. Thomas' Hospital





(artistic license – this heart rate is fictional)

“**Equity** is the absence of unfair, avoidable or remediable differences among groups of people ...”

“**Equity** is the absence of unfair, avoidable or **remediable** differences among groups of people ...”

unfair: without favouritism or discrimination

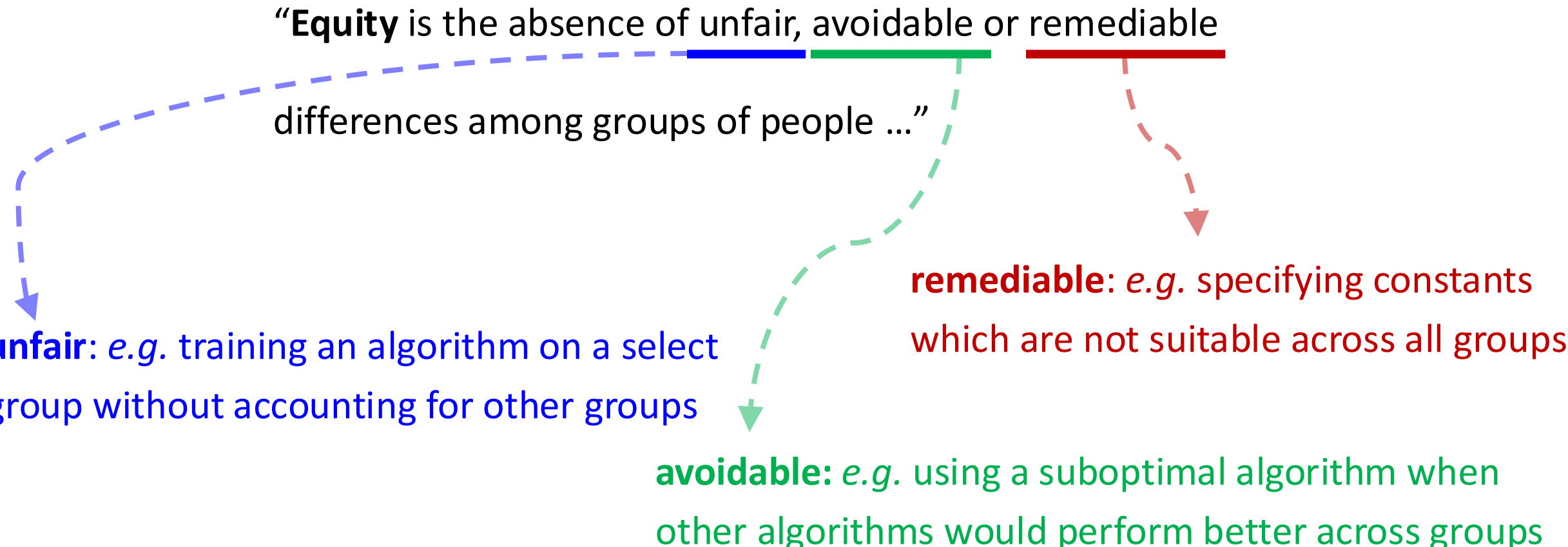
[Oxford English dictionary]

remediable: capable of being corrected

groups of people: “... whether those groups are defined socially, economically, demographically, or geographically or by other dimensions of inequality (e.g. sex, gender, ethnicity, disability, or sexual orientation).”

Examples from signal processing algorithms:

“**Equity** is the absence of unfair, avoidable or remediable differences among groups of people ...”



The diagram features a central quote: “**Equity** is the absence of unfair, avoidable or remediable differences among groups of people ...”. The words 'unfair', 'avoidable', and 'remediable' are underlined in blue, green, and red respectively. Dashed arrows of corresponding colors point from each underlined word to its definition below. A blue arrow points from 'unfair' to its definition on the left. A green arrow points from 'avoidable' to its definition at the bottom. A red arrow points from 'remediable' to its definition on the right.

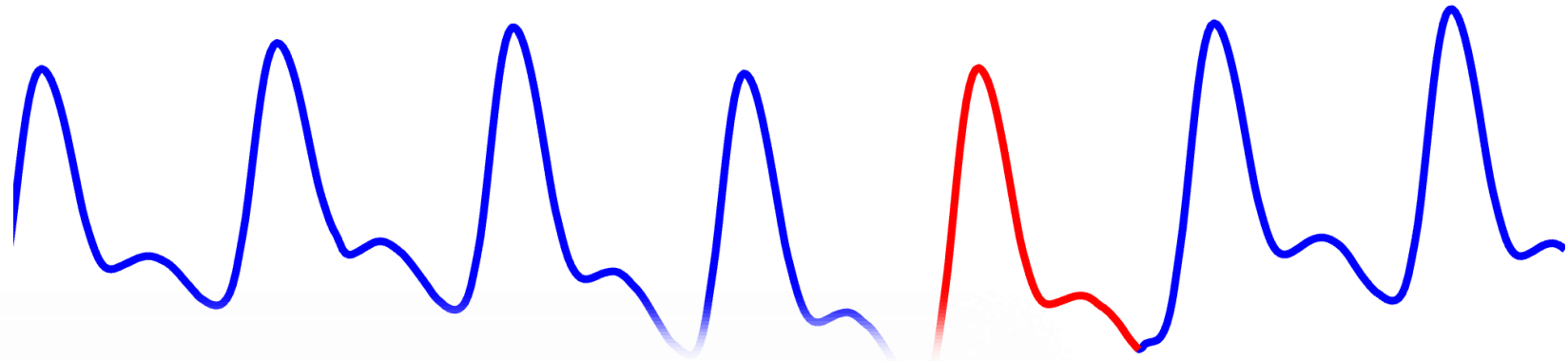
unfair: *e.g.* training an algorithm on a select group without accounting for other groups

avoidable: *e.g.* using a suboptimal algorithm when other algorithms would perform better across groups

remediable: *e.g.* specifying constants which are not suitable across all groups

PPG beat detection:

Adult:

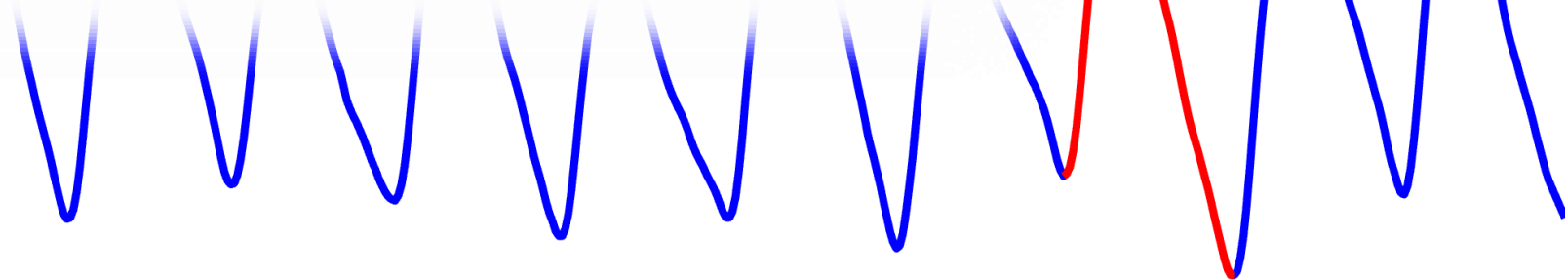


% Assumptions

```
typical_heart_rate = 90; % in beats per minute
```

Heart rate: ~97 bpm

Neonate:

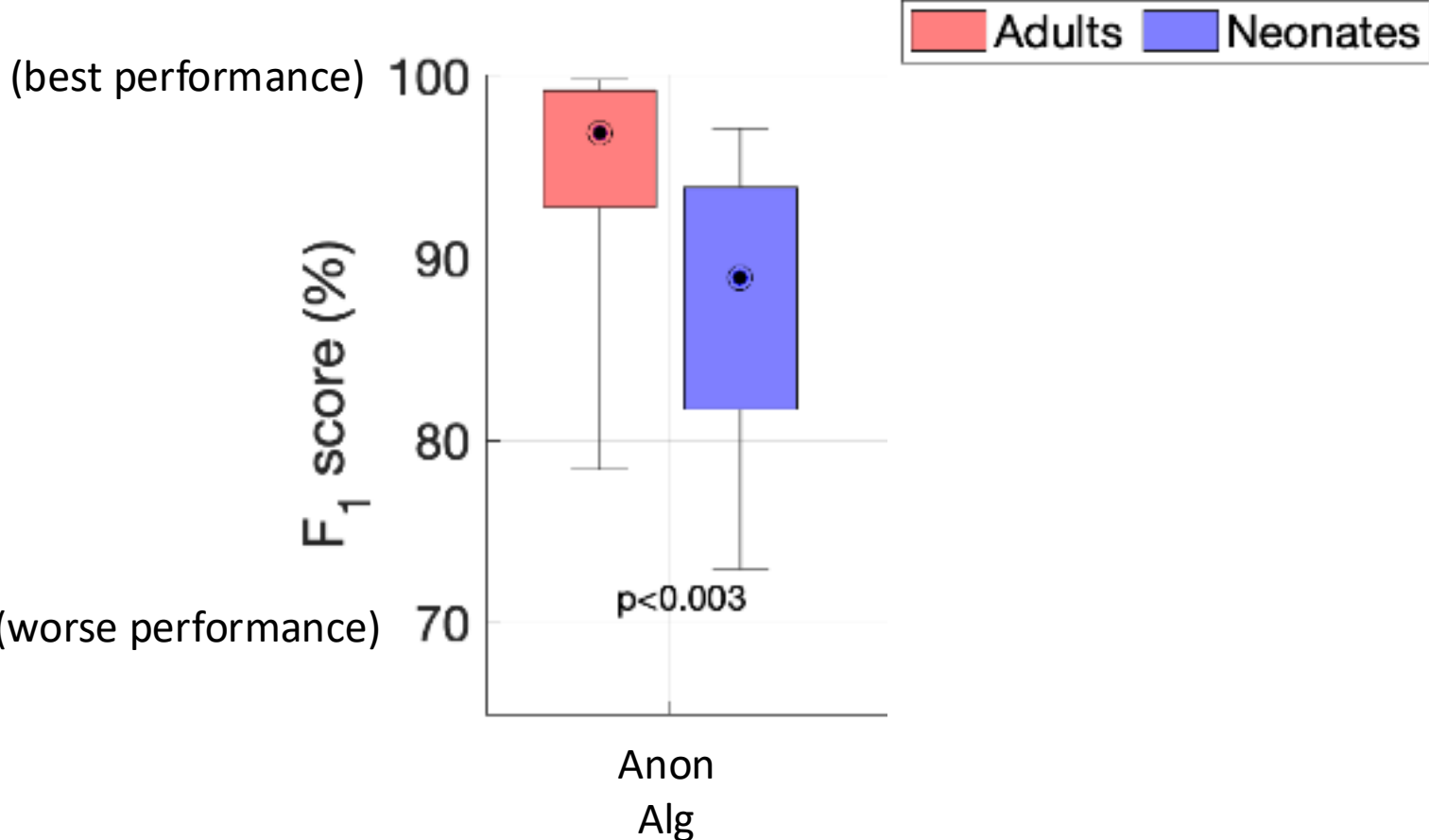


Heart rate: ~150 bpm

PPG beat detection:

%% Assumptions

typical_heart_rate = 90; % in beats per minute



PPG beat detection:

```
%% Assumptions
```

```
typical_heart_rate = 90; % in beats per minute
```

“**Equity** is the absence of unfair, avoidable or remediable differences among groups of people ...”

unfair: equal training provides performance benefit to all groups
get up with out diverse subjects? other groups

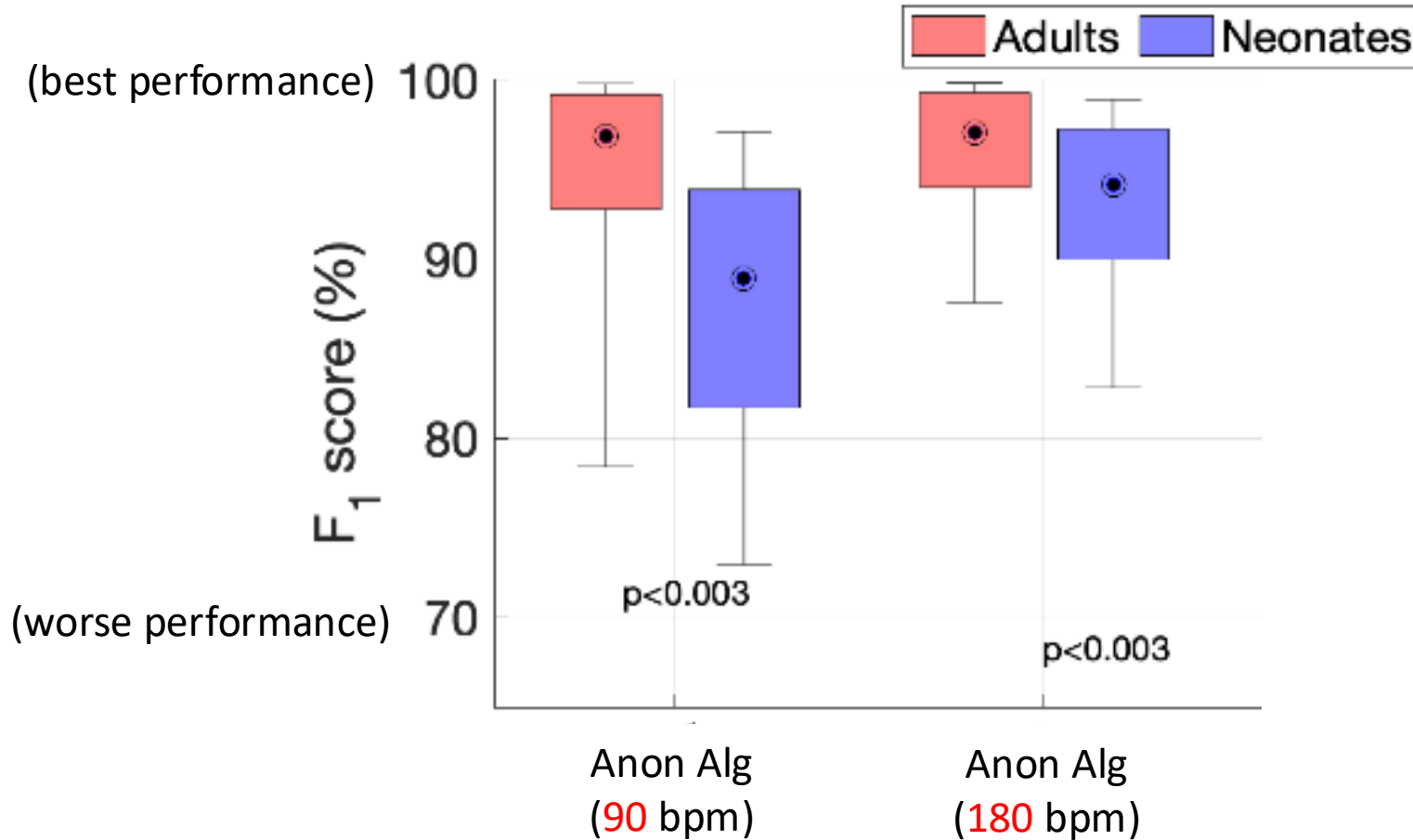
remediable: could we improve performance across groups by modifying this constant?

avoidable: could a better algorithm provide better performance across different groups

PPG beat detection:

%% Assumptions

typical_heart_rate = 180; % in beats per minute



PPG beat detection:

```
%% Assumptions
```

```
typical_heart_rate = 90; % in beats per minute
```

“**Equity** is the absence of unfair, avoidable or remediable differences among groups of people ...”

unfair: algorithm was trained on adults to identify a typical heart rate of 90 bpm

remediable: could we improve performance across groups by modifying this constant?

avoidable: could use an alternative algorithm which performs better across different groups

PPG beat detection:

```
%% Assumptions
```

```
typical_heart_rate = 90; % in beats per minute
```

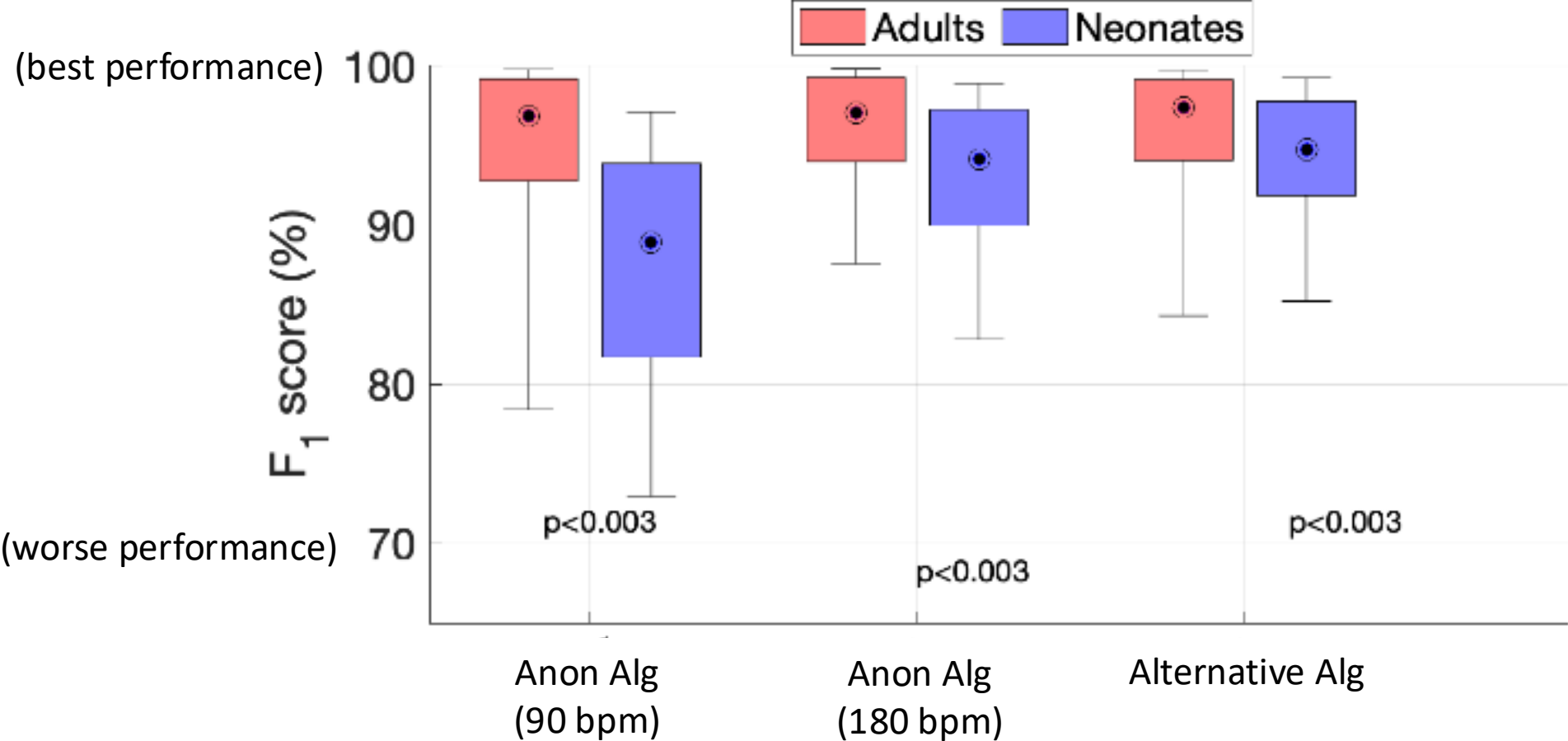
“**Equity** is the absence of unfair, avoidable or remediable differences among groups of people ...”

unfair: algorithm was trained on adults to identify a typical heart rate of 90 bpm

remediable: could we improve performance across groups by modifying this constant?

avoidable: could use an alternative algorithm which performs better across different groups

PPG beat detection:



“**Equity** is the absence of unfair, avoidable or remediable differences among groups of people ...”

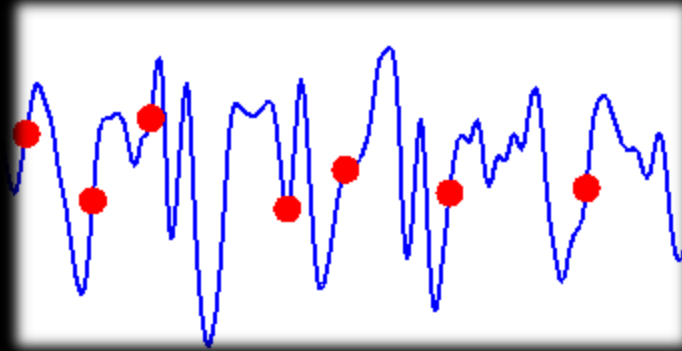
Many, many factors contribute to equity, *e.g.*

- Performance of wearables
- Acceptability of wearables

1. Introduction to wearables



2. Our contributions



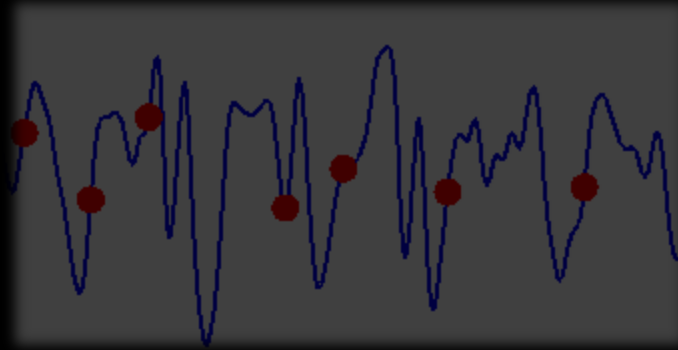
3. Perspectives



1. Introduction to wearables



2. Our contributions



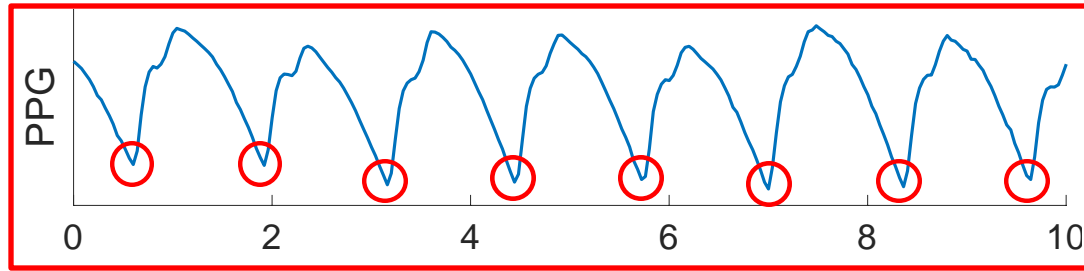
3. Perspectives



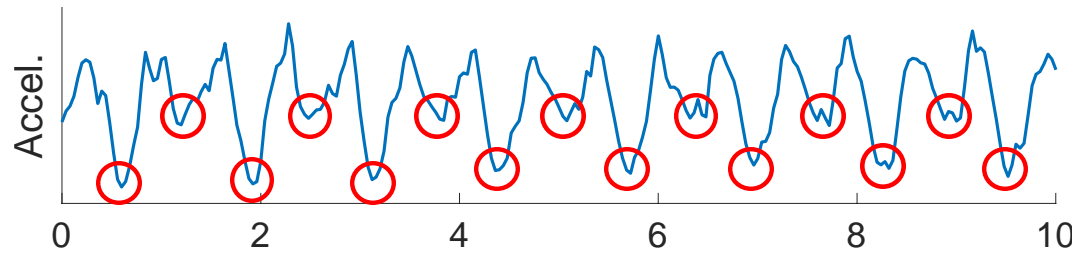


Source: [Charlton et al.](#) Individual images: [P. Charlton](#) under [CC BY 4.0](#); cropped from [image by Marco Verch](#) ([CC BY 2.0](#)); cropped image from [Passler et al.](#) under [CC BY 4.0](#); cropped from [image by GEEK KAZU](#) ([CC BY 2.0](#)); cropped from [image by Pixels](#) ([Pixabay License](#)); cropped from [image by Luke Chesser](#) ([CC0 1.0](#)).

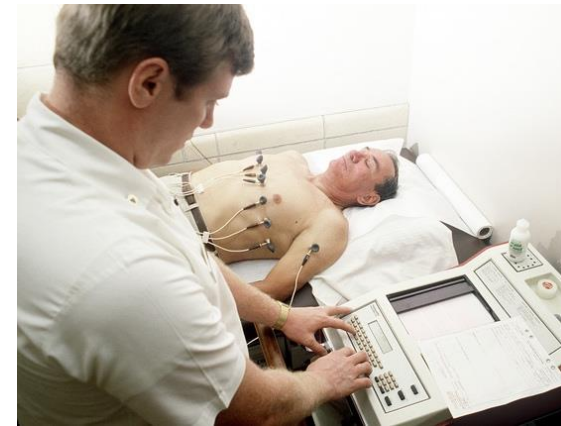
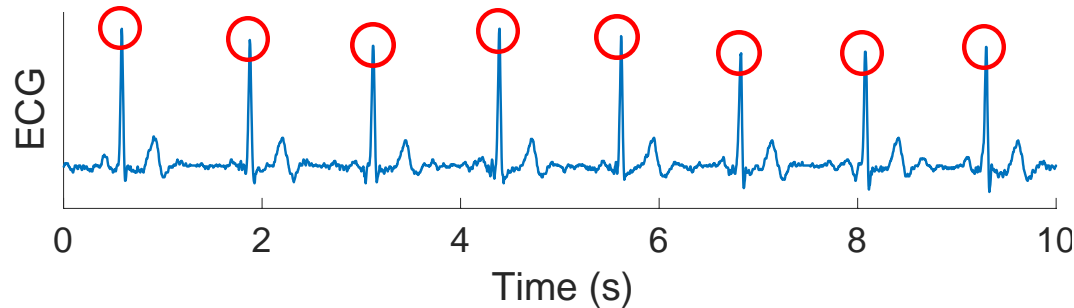
A **Fitness tracker** which acquires photoplethysmography (PPG) and accelerometry (Accel.) signals



○ **Pulse waves** used to:
- estimate heart rate
- identify an irregular pulse



○ **Steps** used to:
- estimate step count

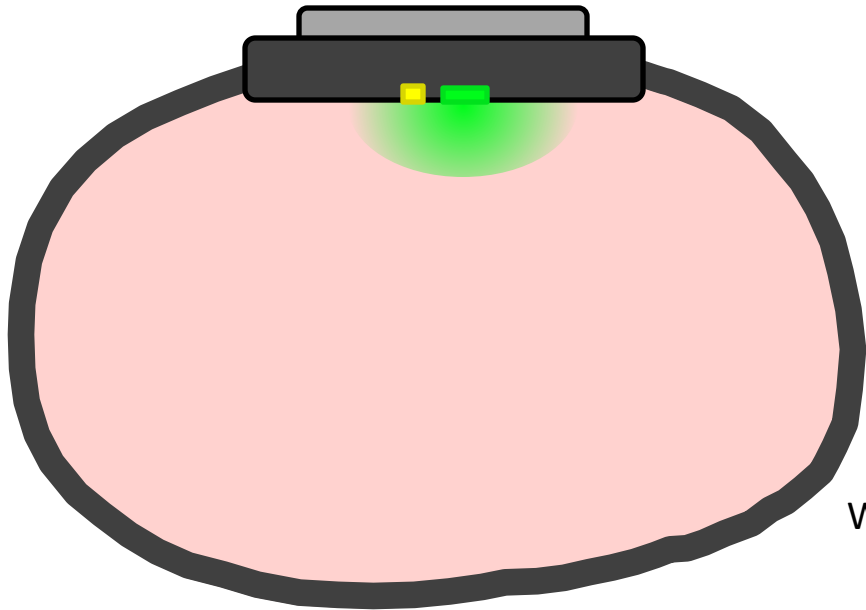


A **Smartwatch** which acquires electrocardiography (ECG) and accelerometry (Accel.) signals

The Photoplethysmogram



Photoplethysmogram (PPG) Sensor

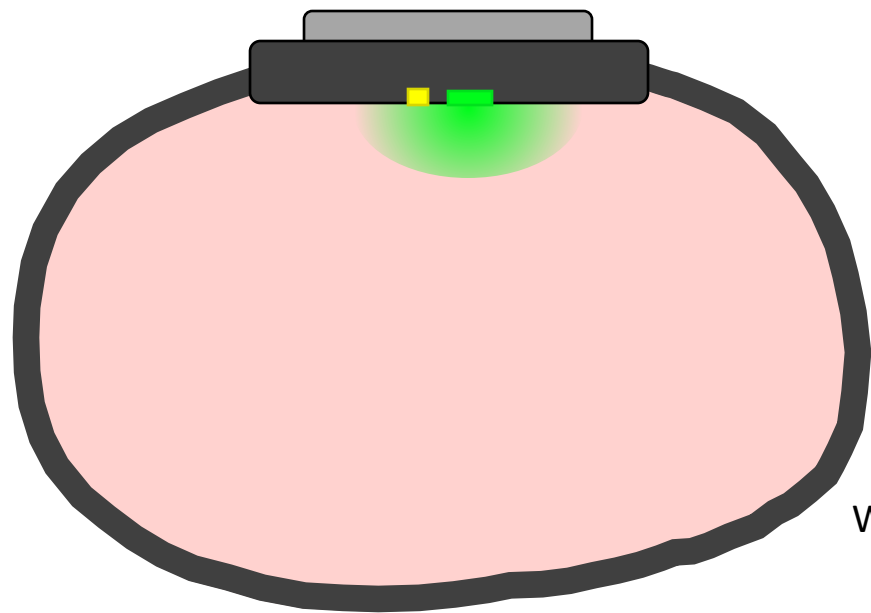


Wrist cross-section

The Photoplethysmogram



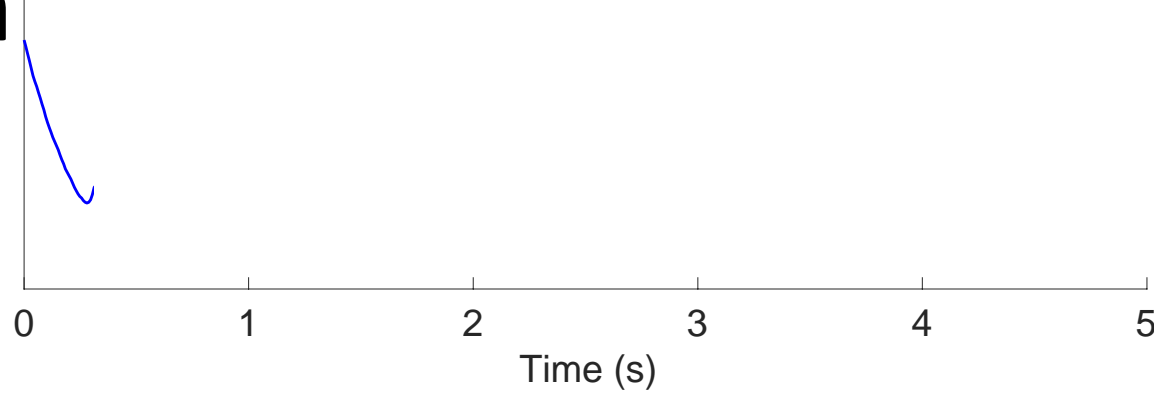
Photoplethysmogram (PPG) Sensor



Wrist cross-section

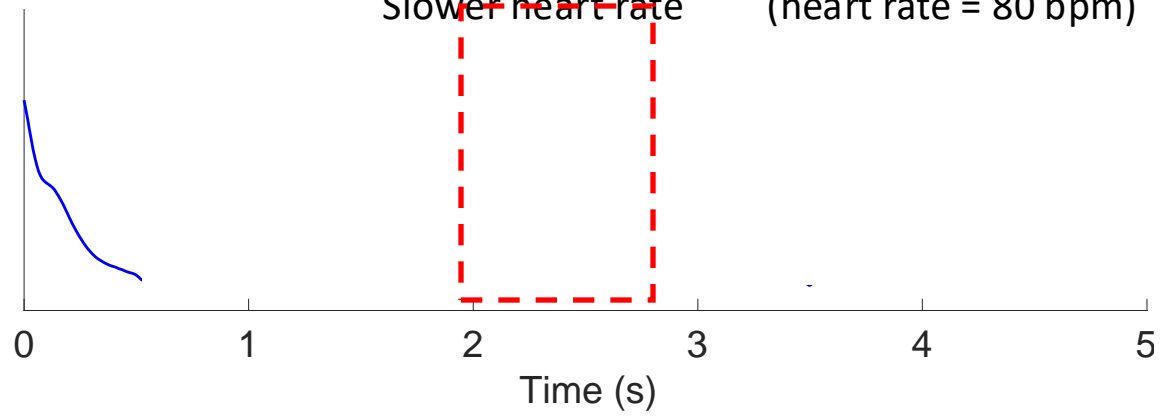
Photoplethysmogram

(heart rate = 100 bpm)



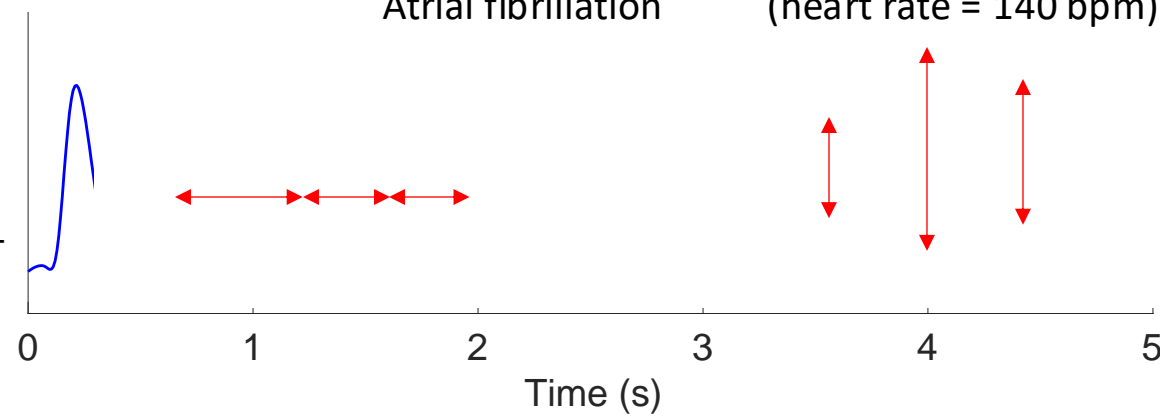
Slower heart rate

(heart rate = 80 bpm)

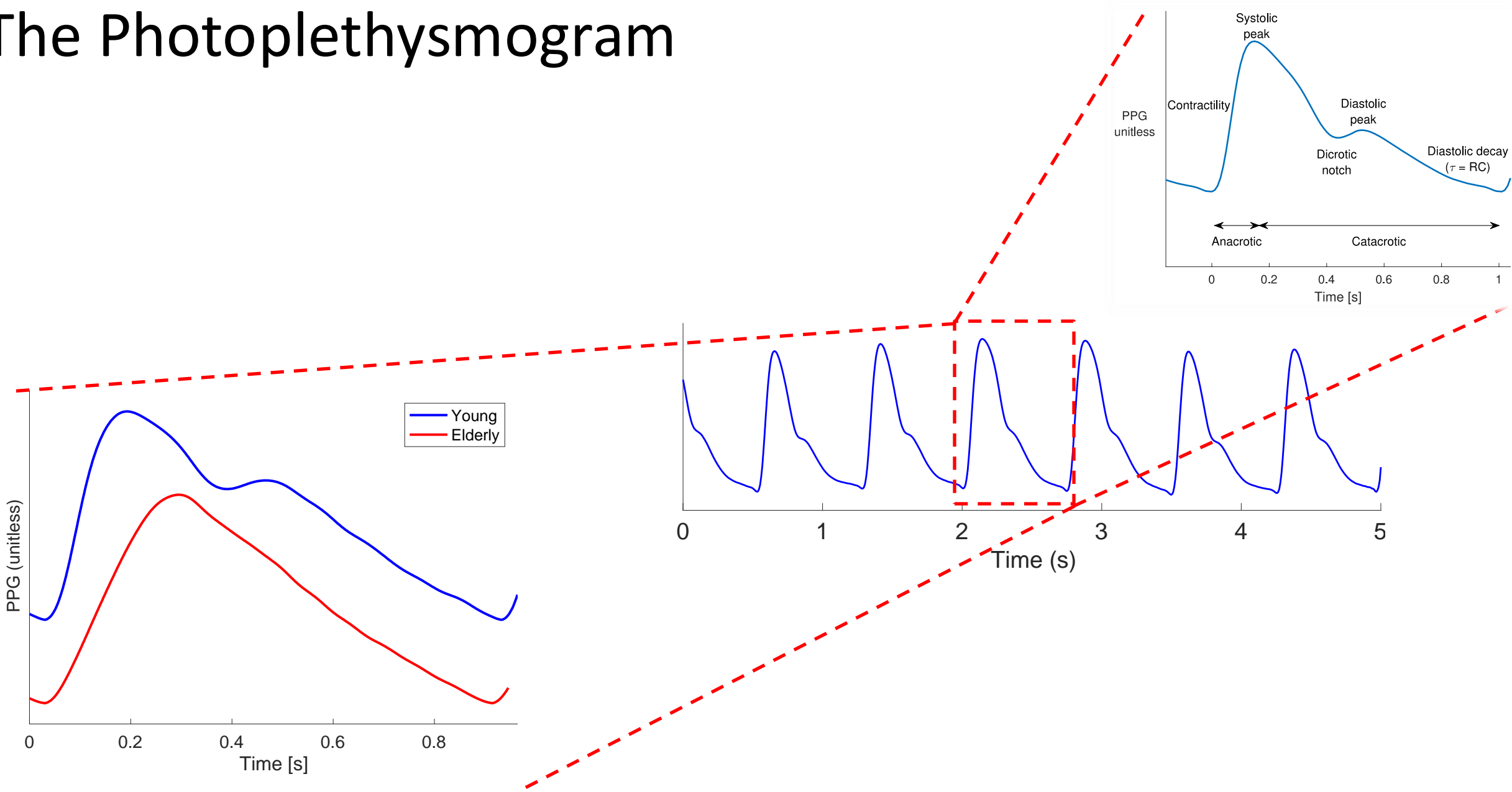


Atrial fibrillation

(heart rate = 140 bpm)



The Photoplethysmogram



Further Reading on Wearable Photoplethysmography

Broad overview:

Charlton P.H. *et al.*, **The 2023 wearable photoplethysmography roadmap**, *Phys Meas*, 2023, <https://doi.org/10.1088/1361-6579/acead2>

Review article:

Charlton P.H. *et al.*, **Wearable Photoplethysmography for Cardiovascular Monitoring**, *Proc. IEEE*, 2022, <https://doi.org/10.1109/JPROC.2022.3149785>

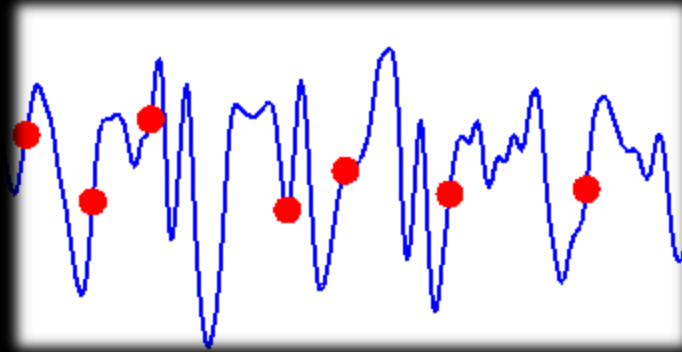
Textbook chapter:

Charlton P.H. and Marozas V., **Wearable photoplethysmography devices**, *Photoplethysmography*, 2021, <https://doi.org/10.1016/B978-0-12-823374-0.00011-6>

1. Introduction to wearables

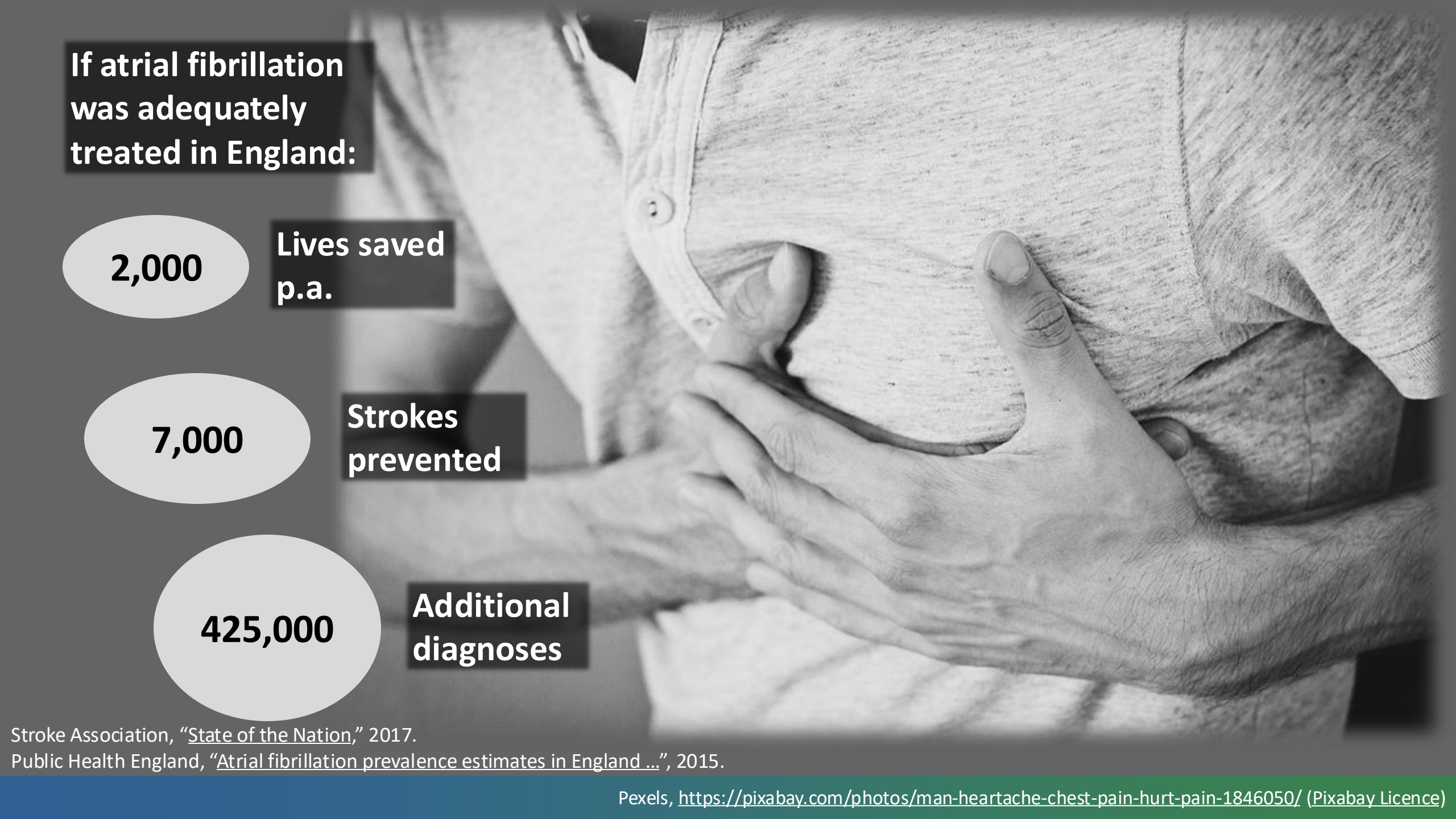


2. Our contributions



3. Perspectives





**If atrial fibrillation
was adequately
treated in England:**

2,000

**Lives saved
p.a.**

7,000

**Strokes
prevented**

425,000

**Additional
diagnoses**

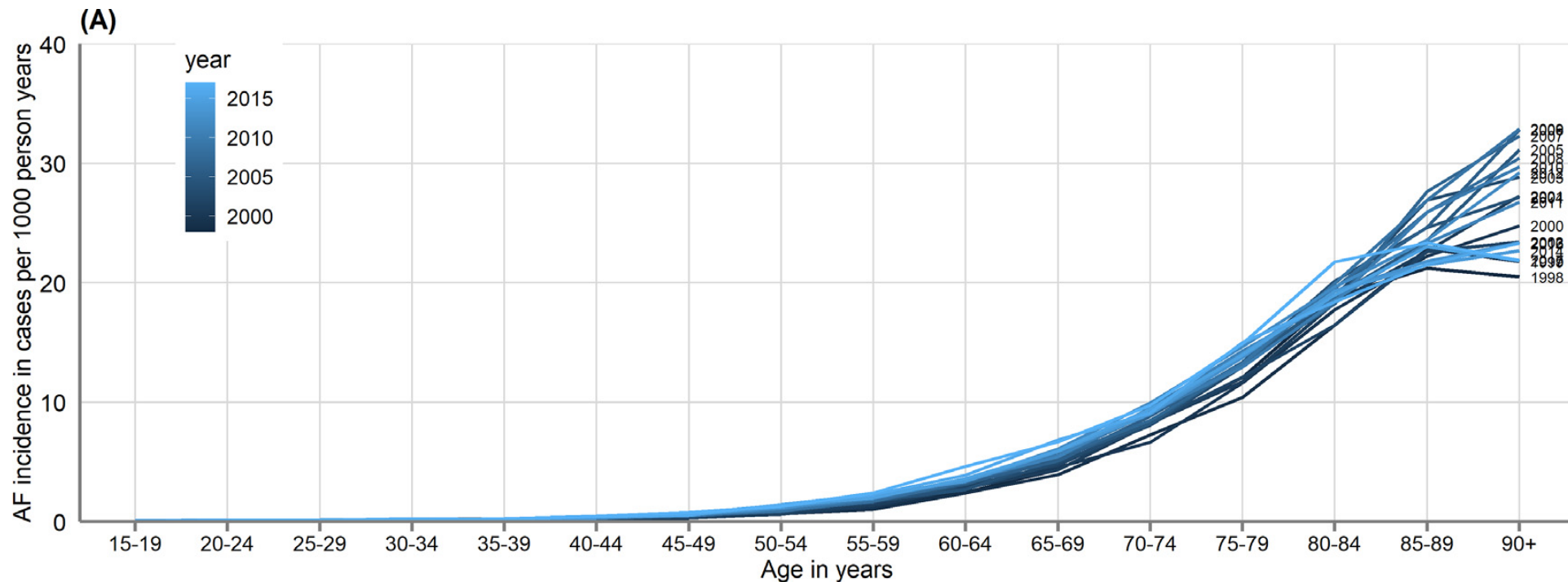
Stroke Association, "[State of the Nation](#)," 2017.

Public Health England, "[Atrial fibrillation prevalence estimates in England ...](#)", 2015.

Target population and target setting

Older adults because:

- AF incidence increases with age



From primary and secondary electronic health records of 3.4 million individuals in England

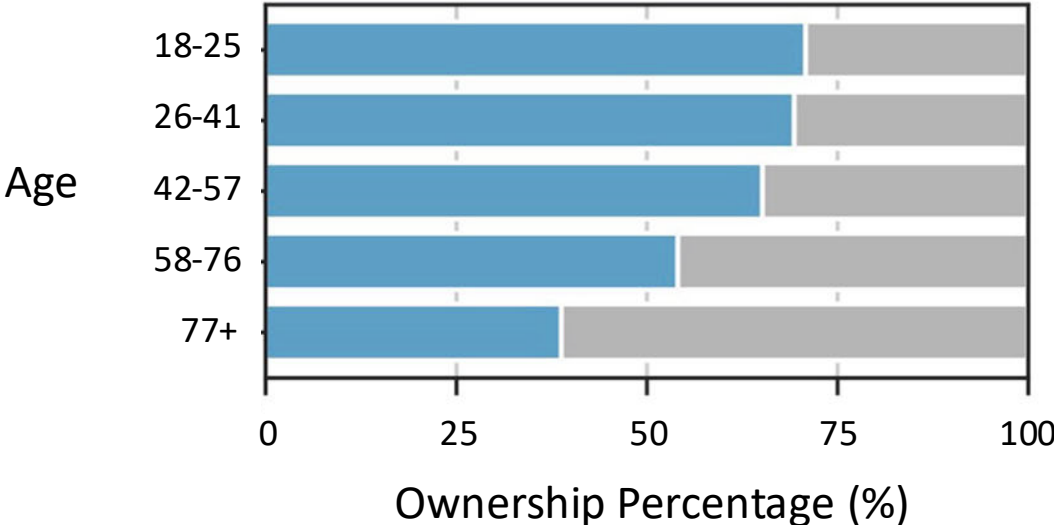
In daily life because:

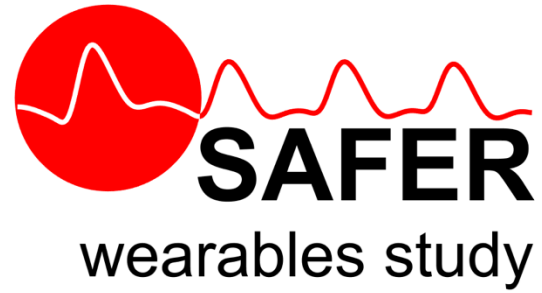
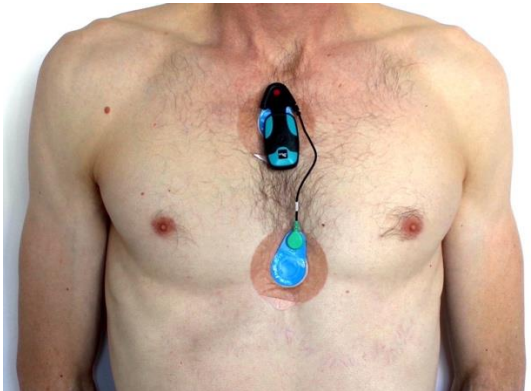
- AF can occur infrequently

Smartwatch ownership

Not everyone has a smartwatch, particularly older adults

An electronic survey of 1,368 patient advisory group members:





SAFER

wearables study

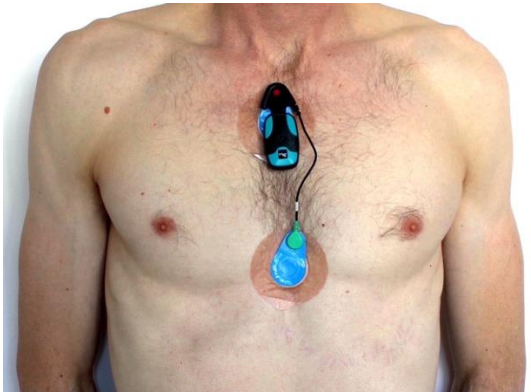
Aim: Assess performance and acceptability of wearables for detecting atrial fibrillation (AF)

Methods:

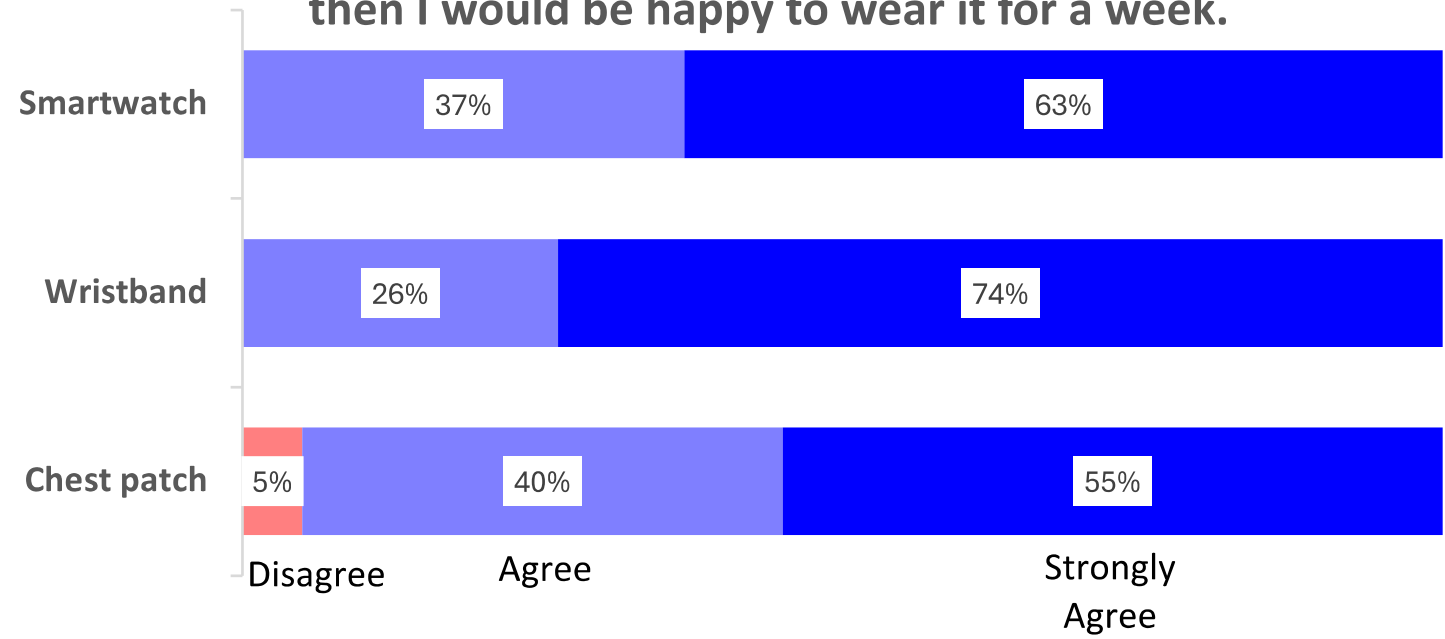
- In 130 older adults, aged 65+, half of whom have AF
- Two wrist-worn devices: smartwatch, wristband
- Reference ECG chest patch
- Questionnaire

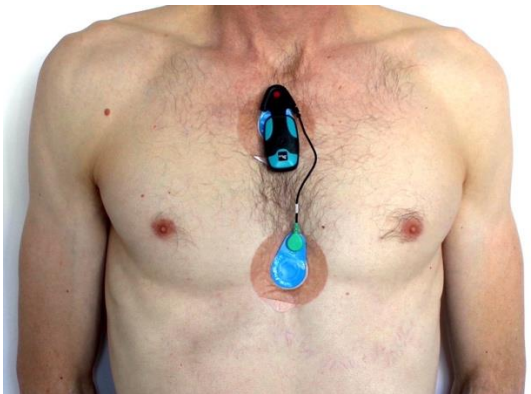
Progress:

- 21 out of 130 participants to date.

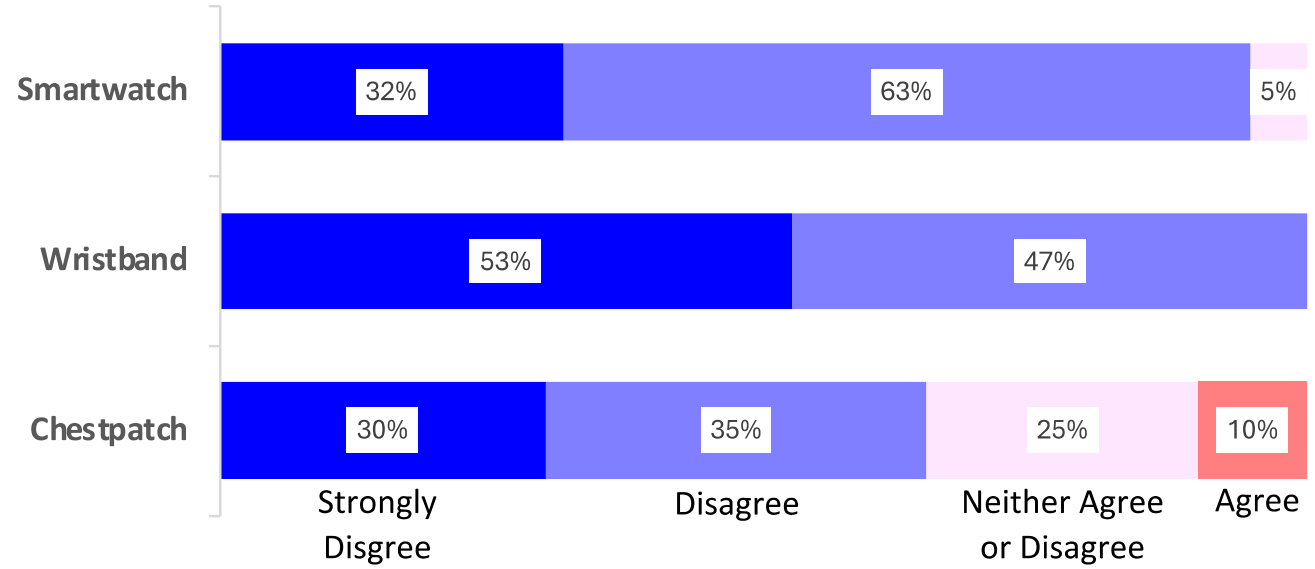


If the device was regularly used to check people's health then I would be happy to wear it for a week.

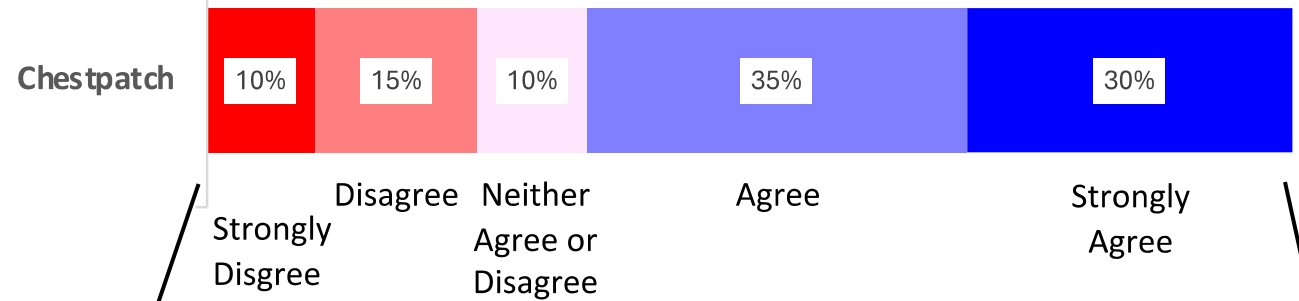




The device was uncomfortable



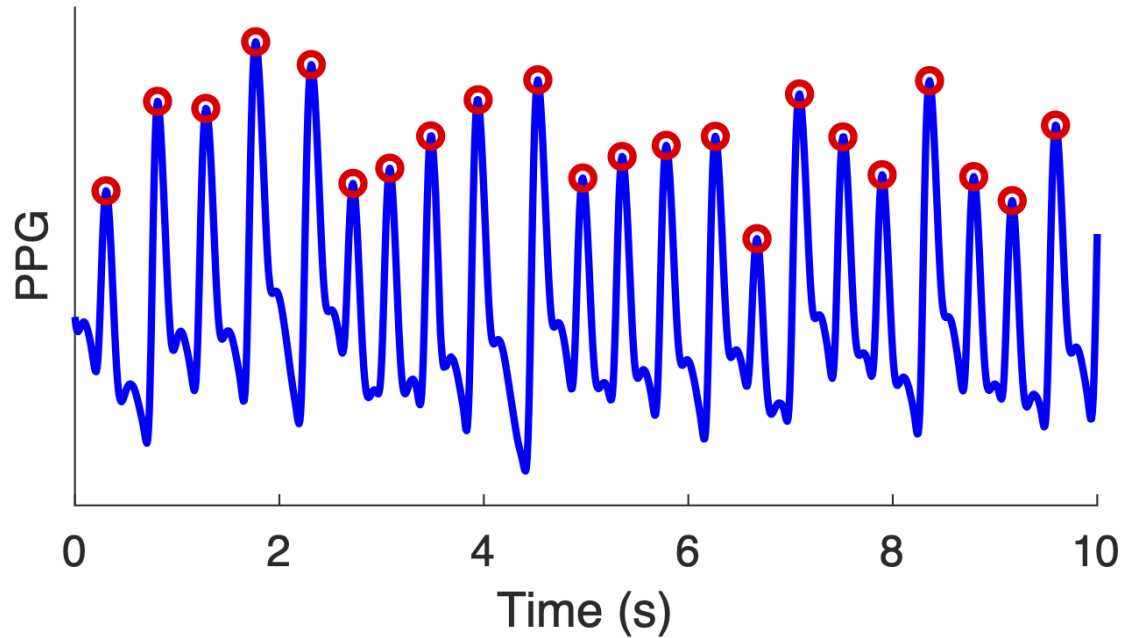
The sticky dots on my chest didn't bother me.



"looked as though I'd been attacked with an octopus with a round ring"

"I forget that's there"

Benchmarking photoplethysmogram beat detectors

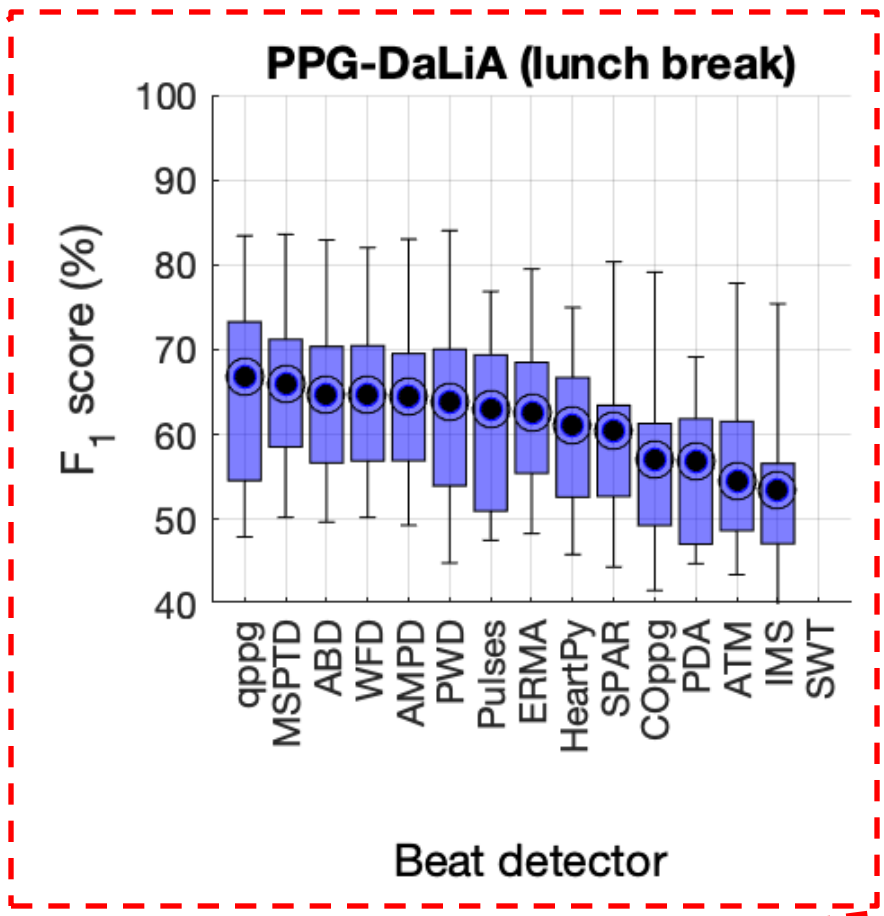
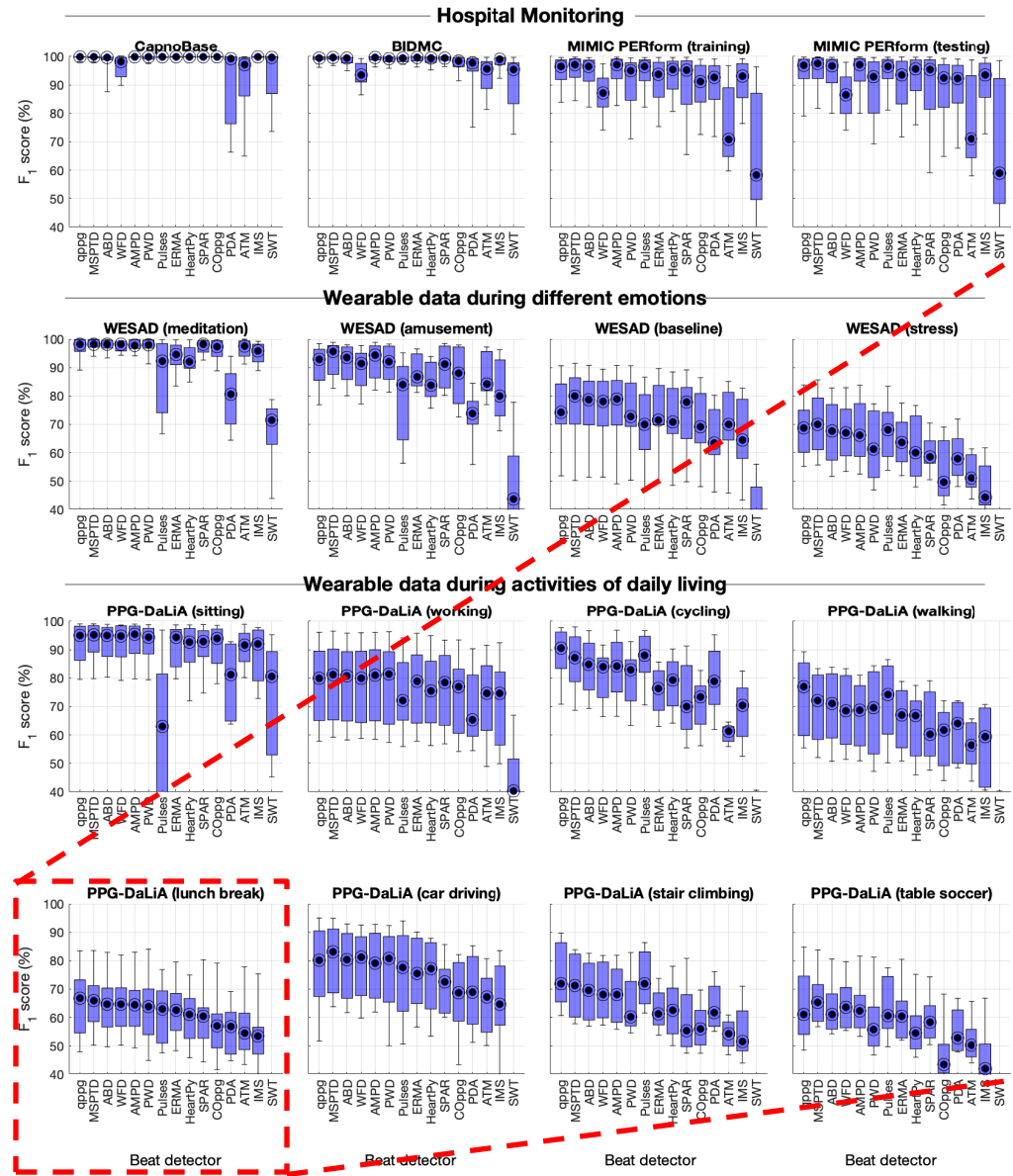


Aim: To identify the best algorithm to detect heart beats in photoplethysmography signals.

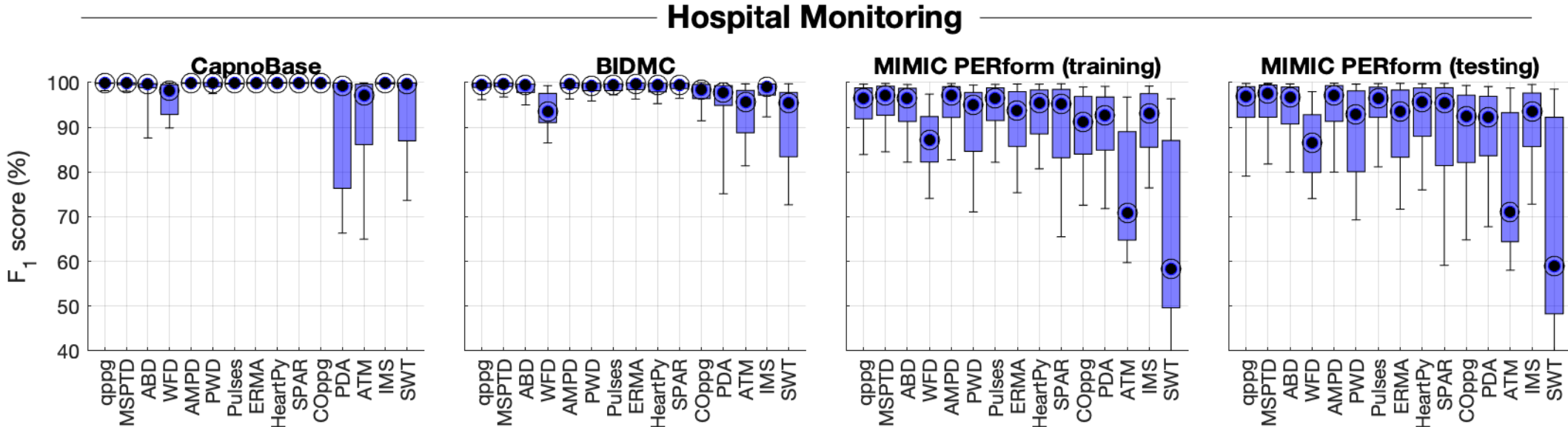
Methods:

- Fifteen open-source beat detection algorithms
- Assessed against electrocardiogram-derived heartbeats
- On eight datasets.

Benchmarking photoplethysmogram beat detectors



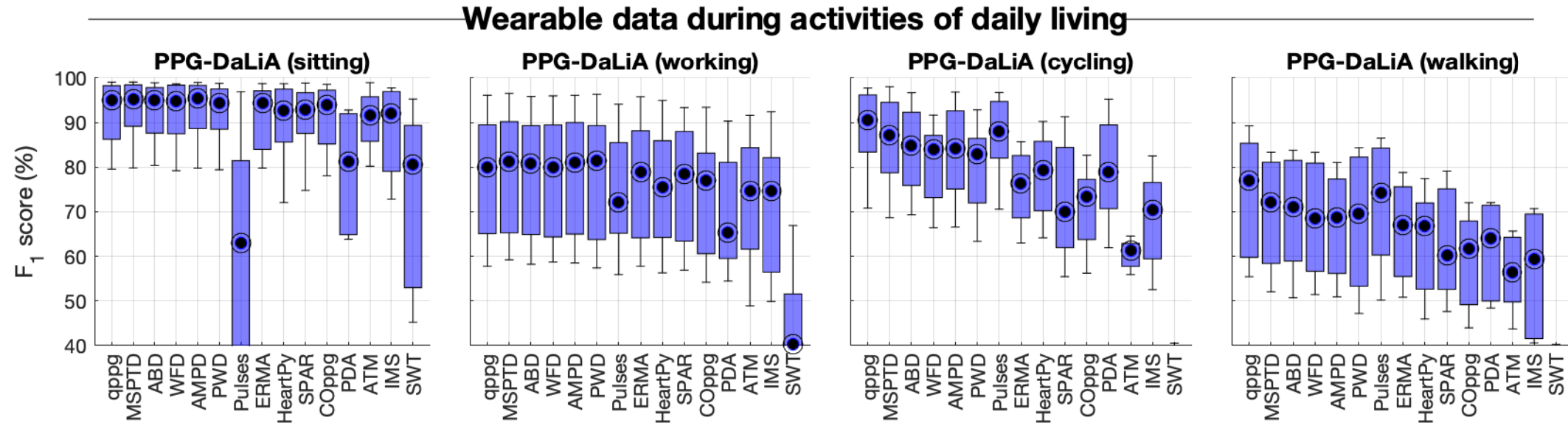
Benchmarking photoplethysmogram beat detectors



Eight beat detectors performed well at rest in the absence of movement

- F1 scores of $\geq 90\%$ on hospital data and wearable data

Benchmarking photoplethysmogram beat detectors



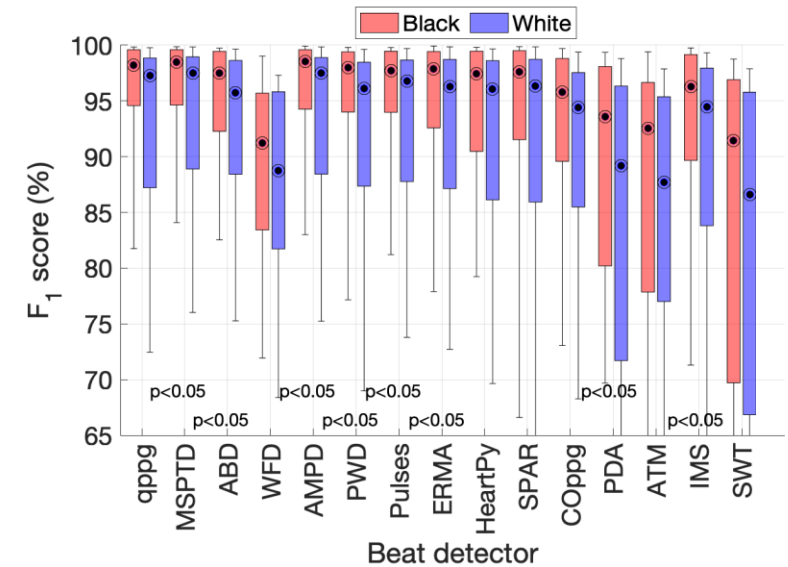
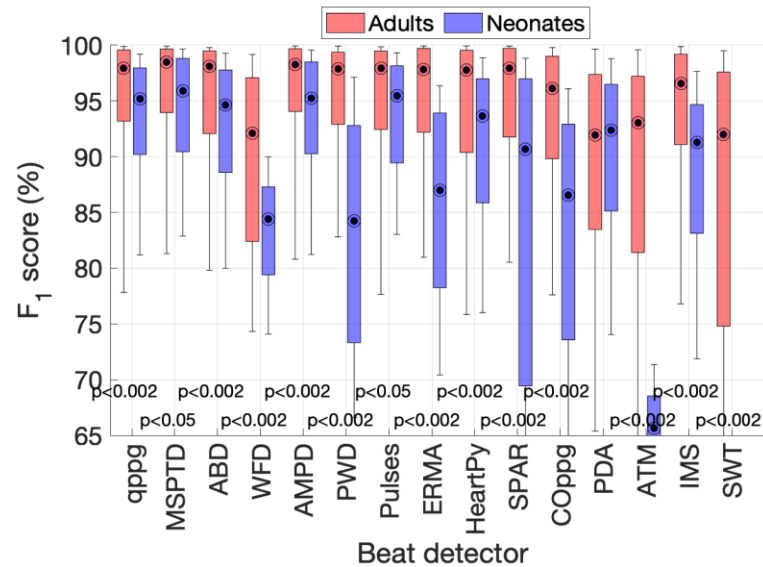
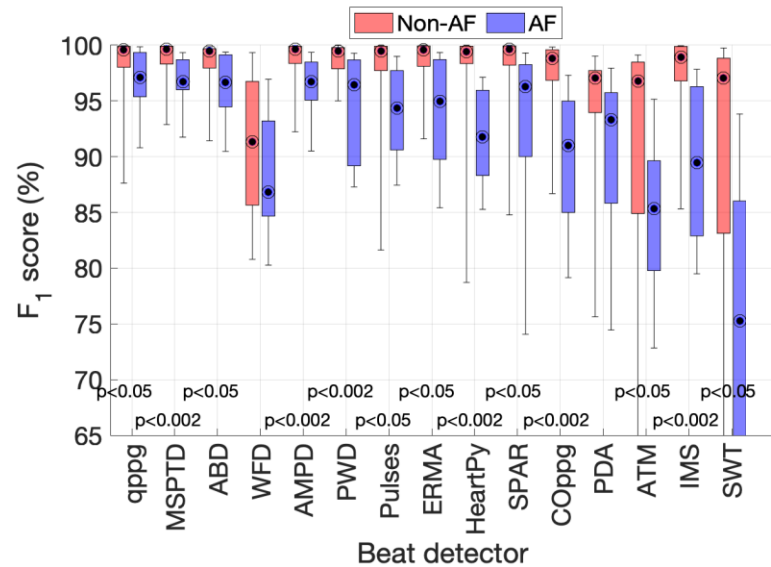
Eight beat detectors performed well at rest in the absence of movement

- F1 scores of $\geq 90\%$ on hospital data and wearable data

Their performance was poorer during exercise:

- F1 scores of 55%–91%

Benchmarking photoplethysmogram beat detectors



Eight beat detectors performed well at rest in the absence of movement

- F1 scores of $\geq 90\%$ on hospital data and wearable data

Their performance was poorer during exercise:

- F1 scores of 55%–91%

Performance was:

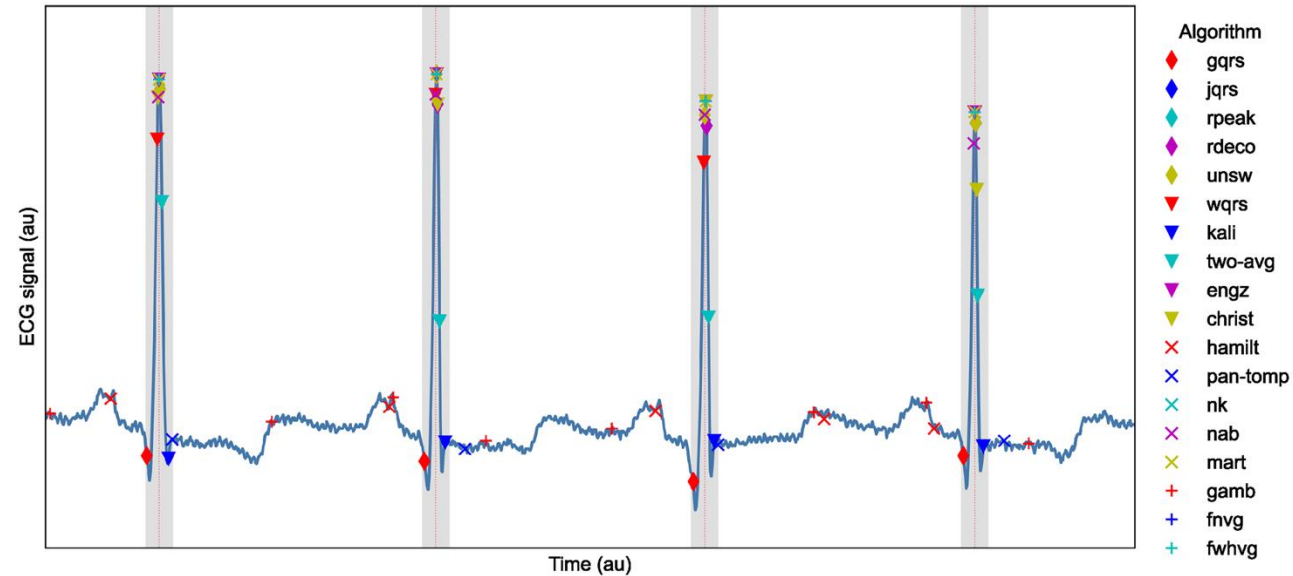
- Poorer in AF
- Poorer in neonates than adults
- Not associated with ethnicity (Black compared with White)

Concluded that 'MSPTD' and 'qppg' performed best, although this is somewhat subjective.

Using photoplethysmography to prompt single-lead ECG



Benchmarking ECG beat detectors

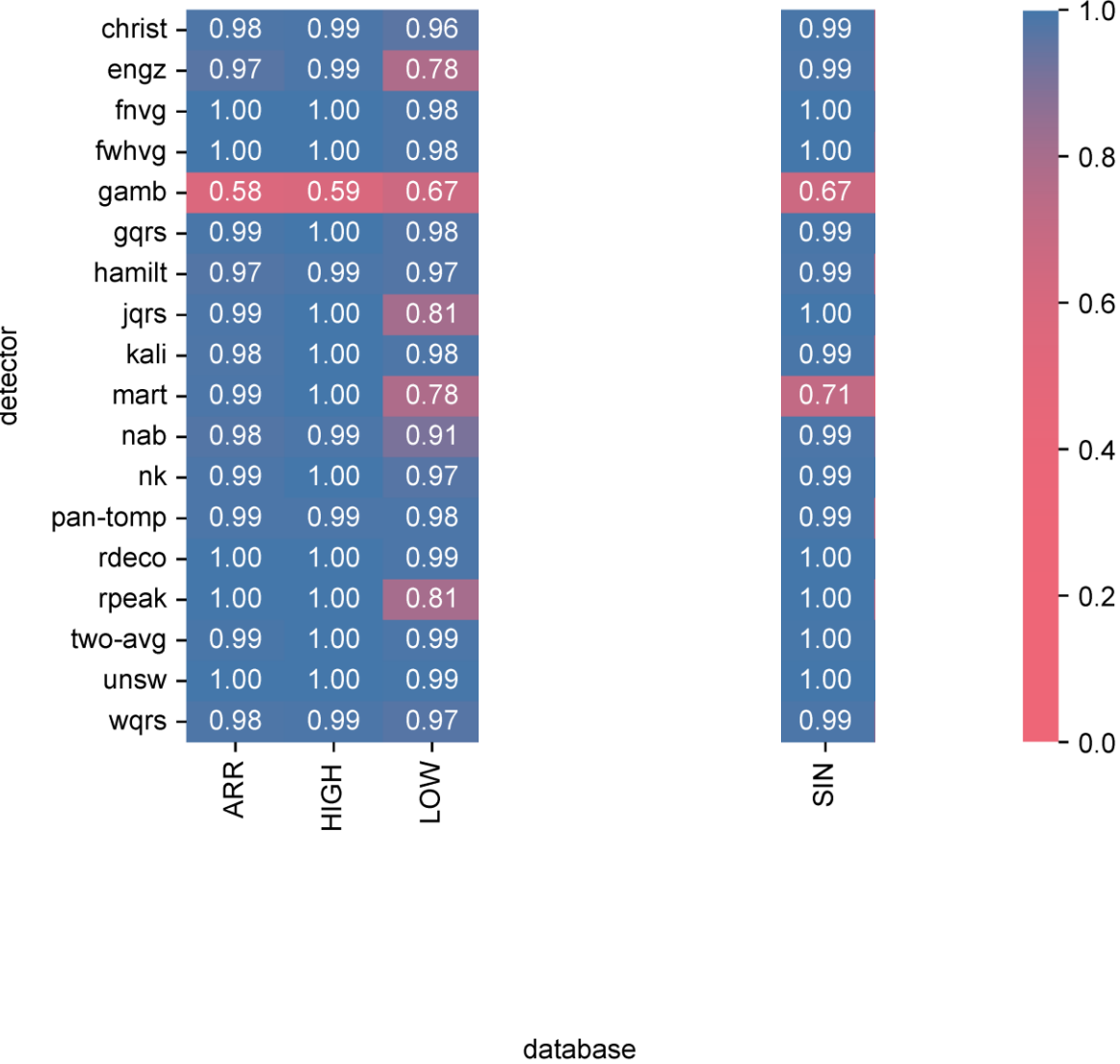


Aim: To identify the best-performing open-source QRS detector for use with telehealth ECGs

Methods:

- 18 open-source beat detection algorithms
- Assessed against manual annotations
- On six datasets.
- Performance expressed as F1-score (ranges from 0 to 1, with 1 being best).

Benchmarking ECG beat detectors



Results:

Using 'F1 score ≥ 0.96 ' as the criteria for determining whether an algorithm performed well:

- 12 algorithms performed well on ECGs collected under clinical supervision.
- Fewer performed well on telehealth ECGs:
 - Five performed well on the TELE dataset
 - Six performed well on high-quality SAFER data
 - Performance was poorer on low-quality SAFER data.

Respiratory rate (RR) (breaths per minute)

Elevated RR

associated with:

Diagnosis of COVID-19

Intensive care admission
with COVID-19

Death in hospital with
COVID-19

Abnormal

24

20

Normal

12

9

Abnormal

kalhh, Pixabay, <https://pixabay.com/illustrations/upper-body-lung-copd-disease-944557/> Royal College of Physicians, 'National Early Warning Score (NEWS) 2: Standardising the assessment of acute-illness severity in the NHS', 2017

DOI: [10.1016/S2589-7500\(20\)30274-0](https://doi.org/10.1016/S2589-7500(20)30274-0) ; DOI: [10.1001/jama.2020.6775](https://doi.org/10.1001/jama.2020.6775)

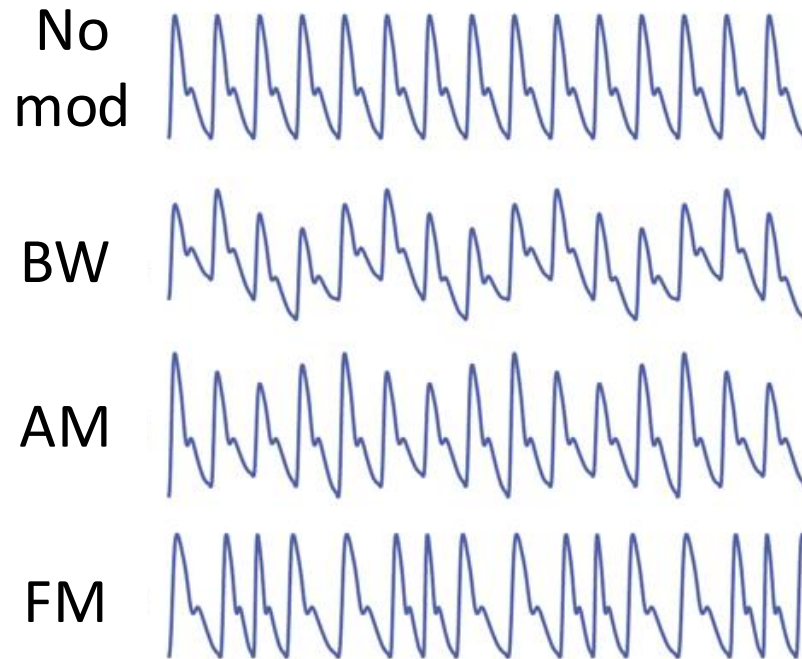
DOI: [10.1002/emp2.12350](https://doi.org/10.1002/emp2.12350)

DOI: [10.1017/ice.2020.461](https://doi.org/10.1017/ice.2020.461)

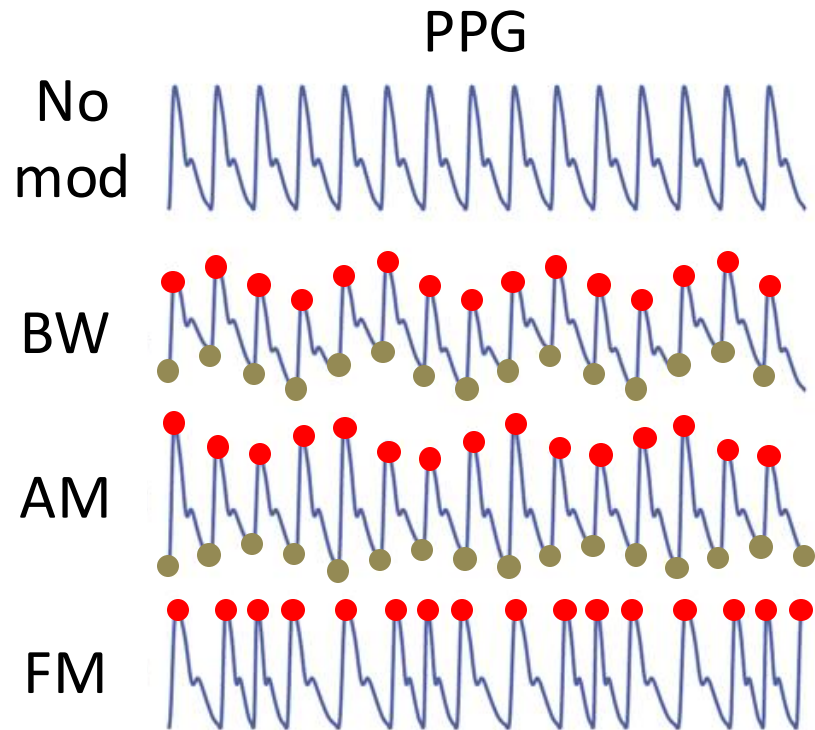
Structure of Algorithms



PPG

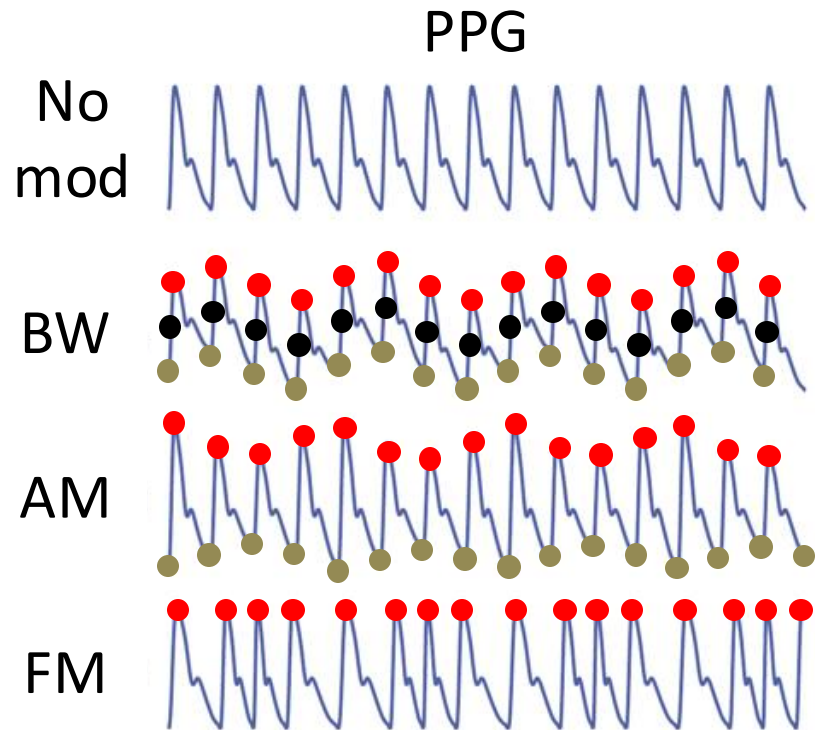


Structure of Algorithms

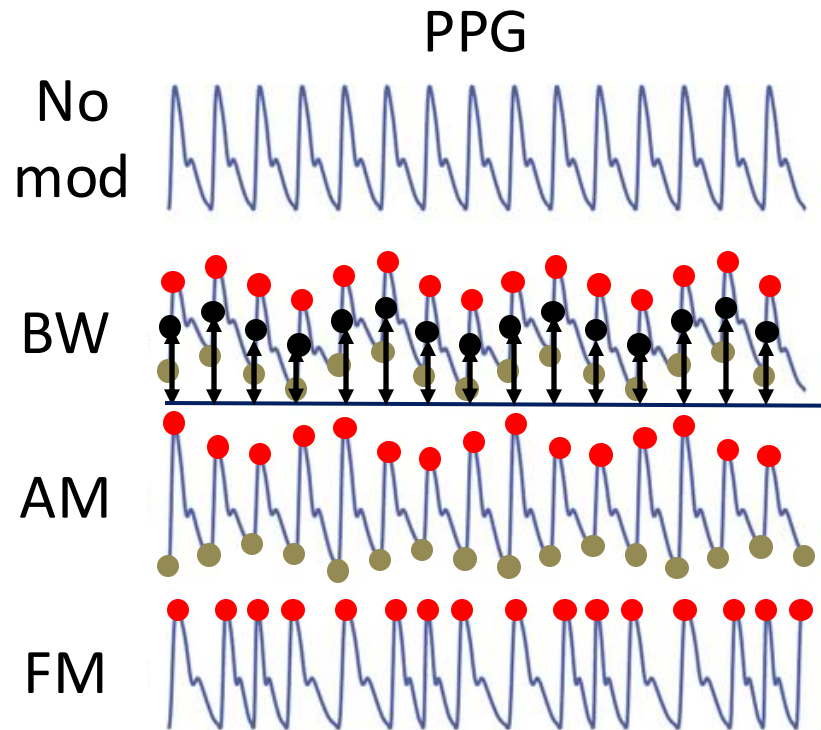


Identify
fiducial
points

Structure of Algorithms

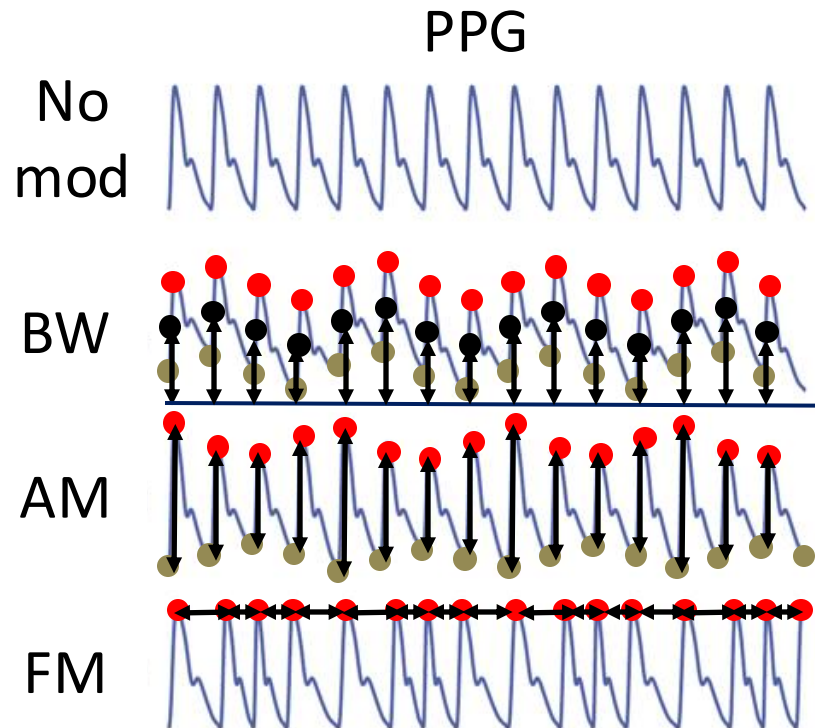


Structure of Algorithms



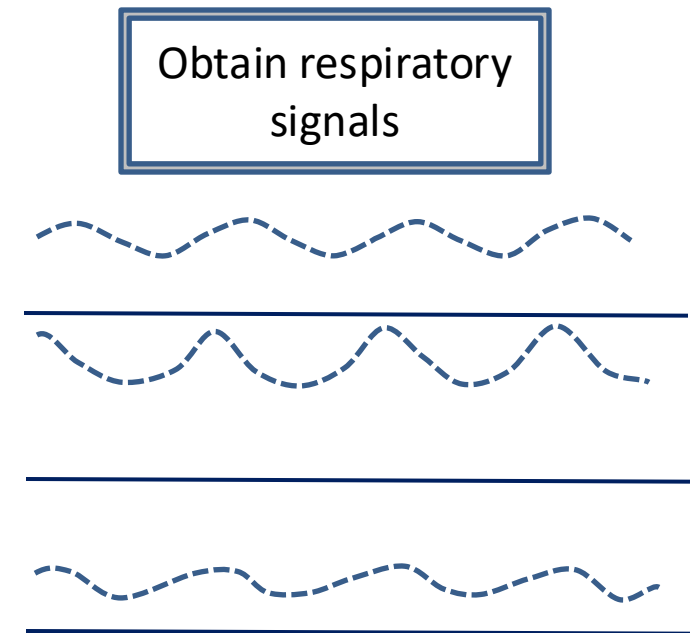
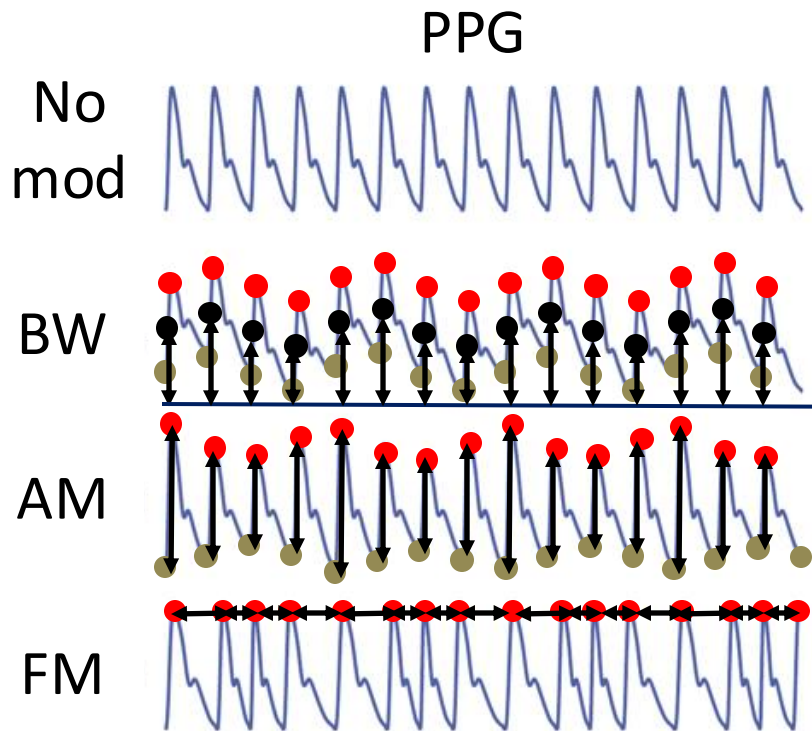
Find baseline

Structure of Algorithms

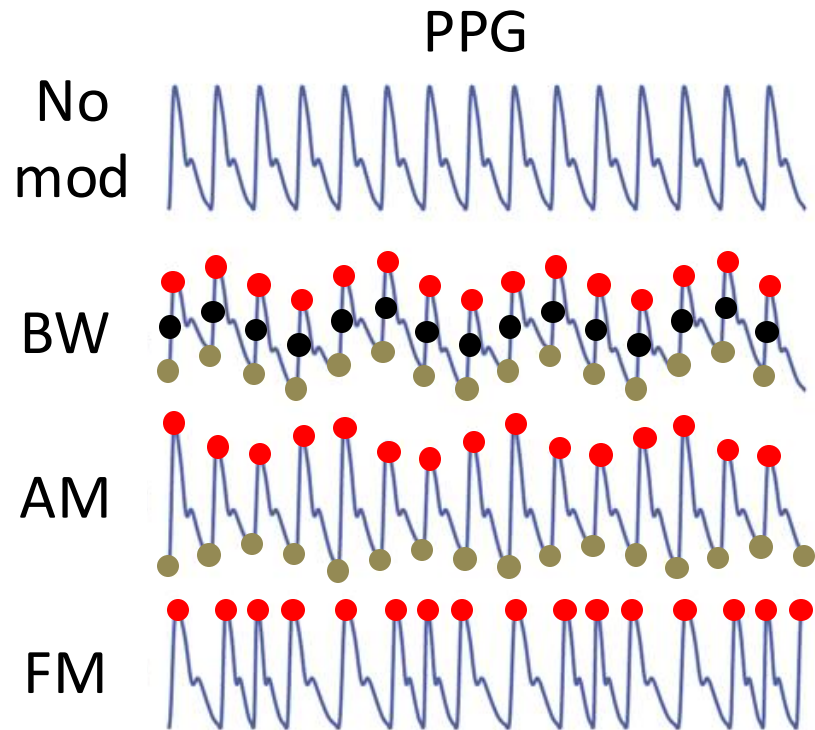


Measure amplitudes and intervals

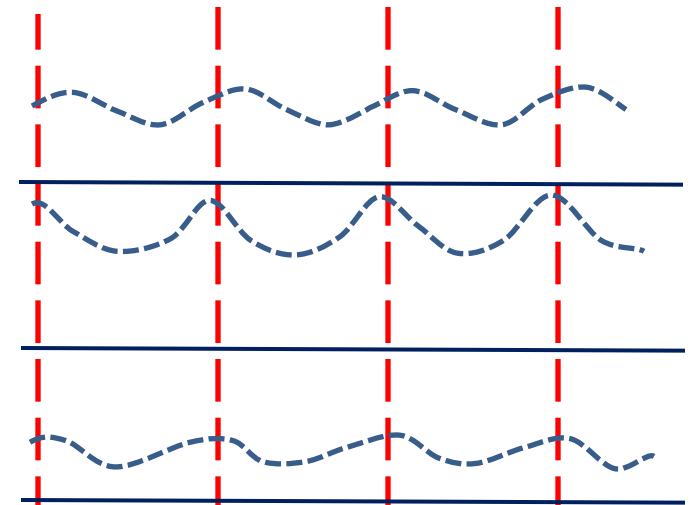
Structure of Algorithms



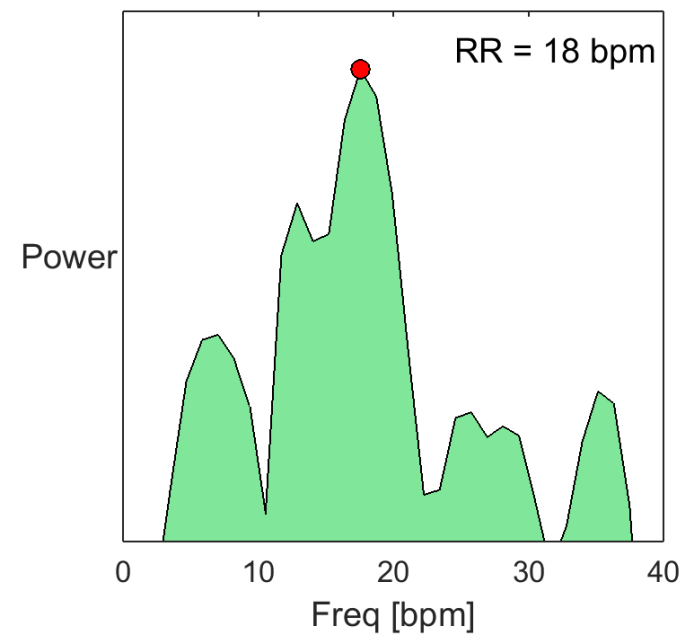
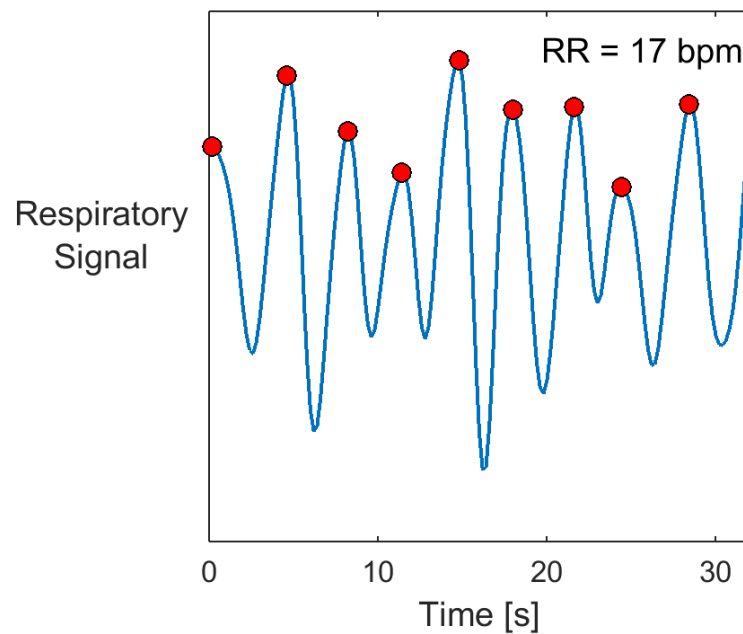
Structure of Algorithms



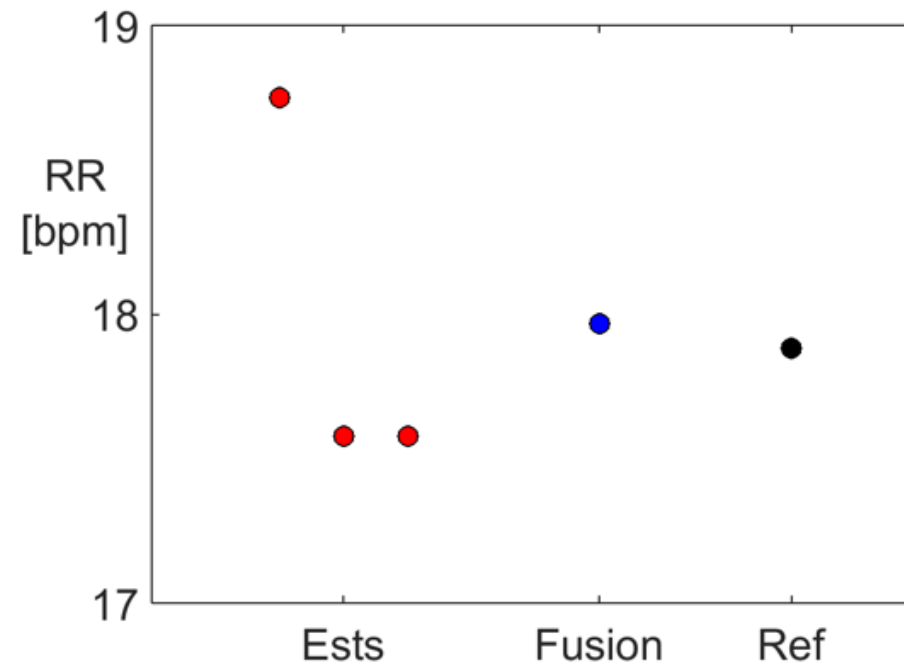
breaths



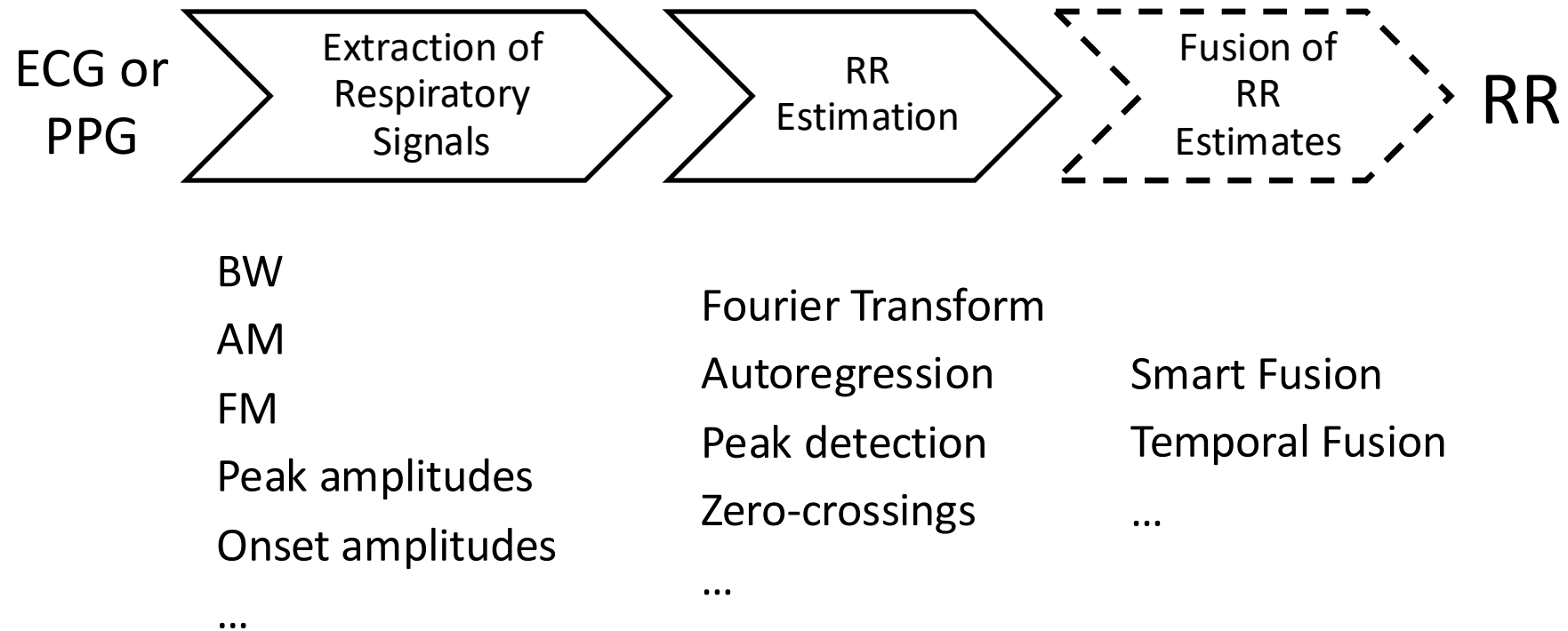
Structure of Algorithms



Structure of Algorithms



Implementing Algorithms



Further reading:

Charlton PH *et al.*, Breathing rate estimation from the electrocardiogram and photoplethysmogram: a review. *IEEE Rev. Biomed. Eng.* **2018**, *11*, 2–20.
doi:[10.1109/RBME.2017.2763681](https://doi.org/10.1109/RBME.2017.2763681)

Implementing Algorithms



BW

AM

FM

Peak amplitudes

Onset amplitudes

...

Fourier Transform

Autoregression

Peak detection

Zero-crossings

...

Smart Fusion

Temporal Fusion

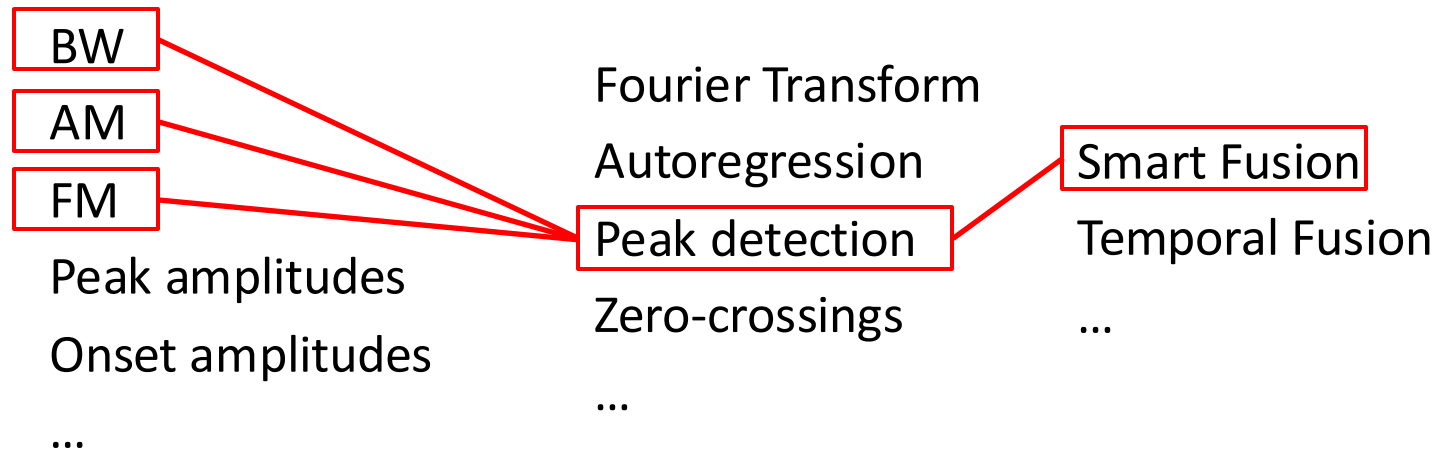
...

Implementing Algorithms

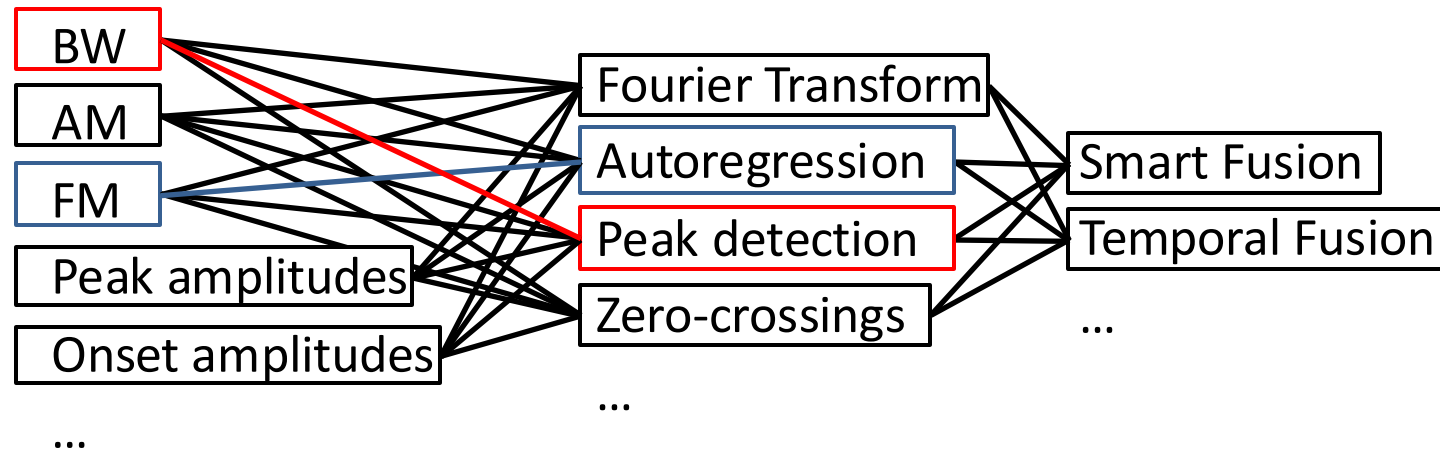


- BW
 - AM
 - FM
 - Peak amplitudes
 - Onset amplitudes
 - ...
- Fourier Transform
 - Autoregression
 - Peak detection
 - Zero-crossings
 - ...
- Smart Fusion
 - Temporal Fusion
 - ...

Implementing Algorithms



Implementing Algorithms



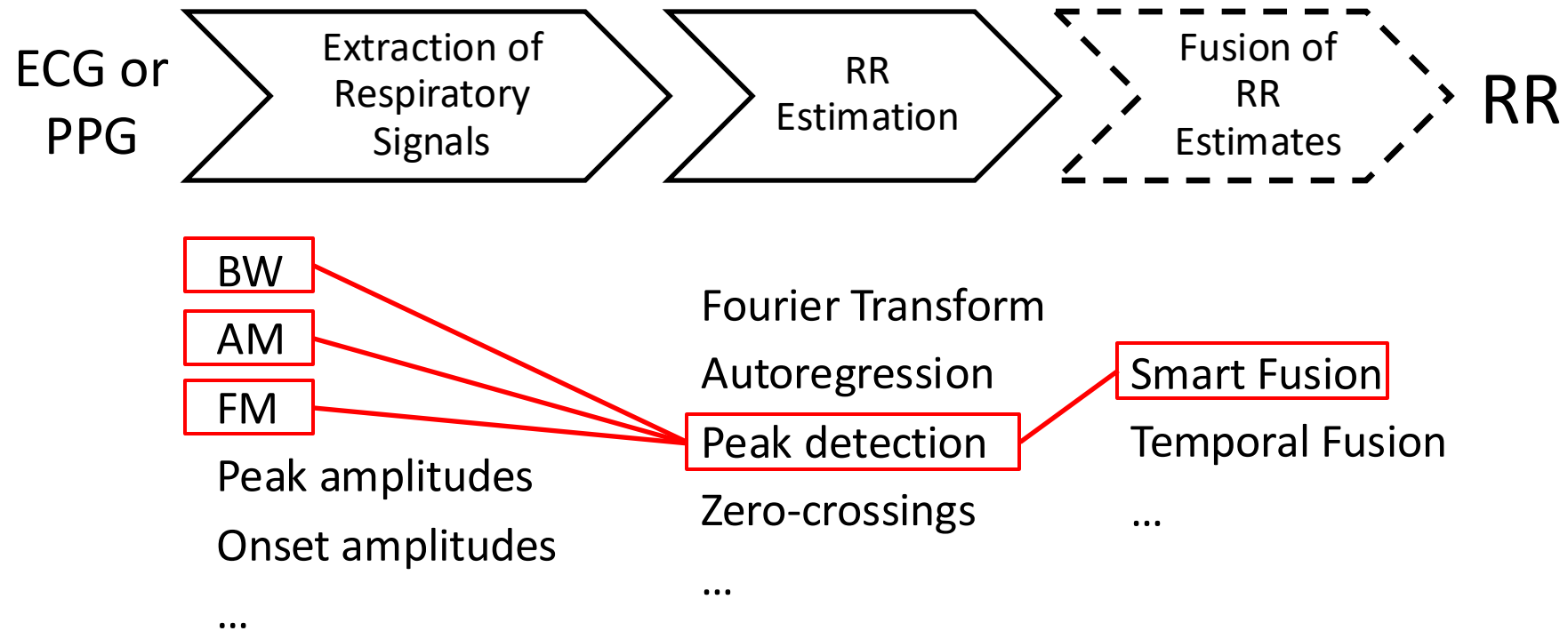
14 techniques

12 techniques

5 techniques

370 algorithms

The winner...



... on young subjects

The strength of respiratory modulations



BW

AM

FM

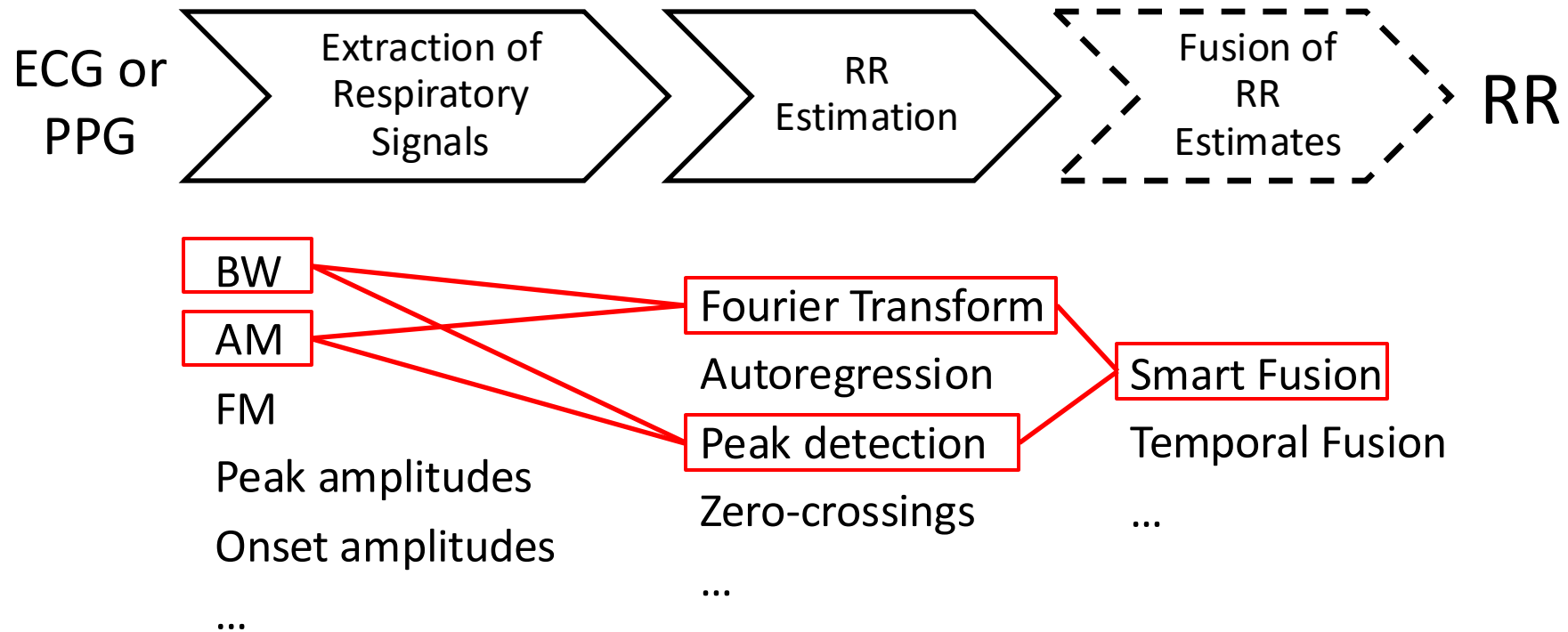
Peak amplitudes

Onset amplitudes

...

Respiratory modulation	Strength (CC): Young adults	Strength (CC): Older adults
BW	0.48 (0.35 – 0.56)	0.49 (0.36 – 0.63)
AM	0.48 (0.39 – 0.57)	0.47 (0.35 – 0.57)
FM	0.64 (0.50 – 0.73)	0.38 (0.32 – 0.55)

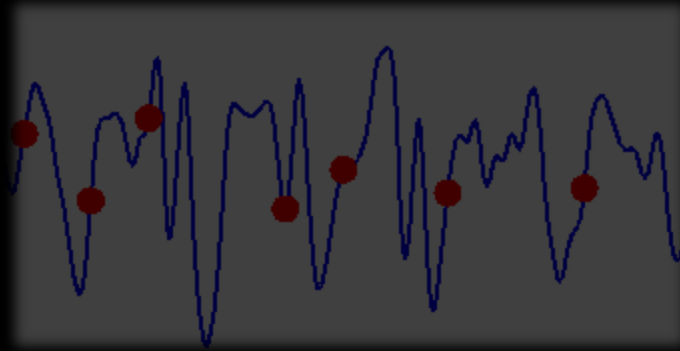
The (revised) winner...



1. Introduction to wearables



2. Our contributions



3. Perspectives



Perspectives

Areas for inequity:

- Many, many potential sources of inequity for wearables. See:
 - D. M. Kaplan, M. Greenleaf, and W. A. Lam, 'Wear With Care: A Call for Empirical Investigations of Adverse Outcomes of Consumer Health Wearables', *Mayo Clinic Proceedings: Digital Health*, vol. 1, no. 3, pp. 413–418, Sep. 2023, doi: [10.1016/j.mcpdig.2023.06.014](https://doi.org/10.1016/j.mcpdig.2023.06.014).

Equity in academic research: Open science

- In my academic role, I value sharing our work. *e.g.*

Open science

- Providing open-source toolboxes of algorithms

MATLAB Toolbox

Respiratory Rate Estimation

Research into estimation of respiratory rate from physiological signals

The Respiratory Rate Estimation project.

The aim of the Respiratory Rate Estimation project is to develop and assess methods for automated respiratory rate (RR) monitoring. It consists of a series of studies of different algorithms for RR estimation from clinical data, complimented by the provision of publicly available datasets and resources.

News: The BIDMC PPG and Respiration dataset is now publicly available here.

Background

The rationale for estimating RR from physiological signals

Datasets

Datasets for evaluation of RR algorithms

Algorithms

A toolbox of RR algorithms to aid research into estimation of RR

Publications

A selection of publications arising from or related to the project

Resources

A selection of additional resources related to project publications

Contributions

Details of how to contribute to this ever-growing resource

Acknowledgments

Thank you to all who have helped make this possible

<https://peterhcharlton.github.io/RRest>

Open science

- Providing open-source toolboxes of algorithms
 - Ease of use

MATLAB Toolbox

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Open science

- Providing open-source toolboxes of algorithms
 - Ease of use

The screenshot shows the PPG-beats website interface. At the top, there is a navigation bar with links for Home, Toolbox, Datasets, Functions, and Tutorials, along with a search bar and navigation arrows. A sidebar on the left lists various PPG Beat Detectors, including Adaptive Threshold Beat Detector, Automatic Beat Detection, and others. The main content area displays the title 'PPG Beat Detectors' and a brief description. Below this, two specific detectors are highlighted: 'Adaptive Threshold Beat Detector' and 'Automatic Beat Detection'. Each entry includes its original publication, a description, a link to the detector, and its license.

PPG Beat Detectors
Algorithms to detect beats in photoplethysmogram (PPG) signals.

PPG-beats contains several algorithms to detect beats in the photoplethysmogram (PPG). This page provides an overview of these beat detectors. Follow the links for further details on each one, and see [this tutorial](#) for an example of how to use them.

Adaptive Threshold Beat Detector

Original publication: Shin HS et al., Adaptive threshold method for the peak detection of photoplethysmographic waveform. *Comput Biol Med* 2009; 39: 1145-52. DOI: [10.1016/j.compbimed.2009.10.006](https://doi.org/10.1016/j.compbimed.2009.10.006)

Description:

Link: [atmax_beat_detector](#) (see also [atmin_beat_detector](#))

Licence: MIT Licence

Automatic Beat Detection

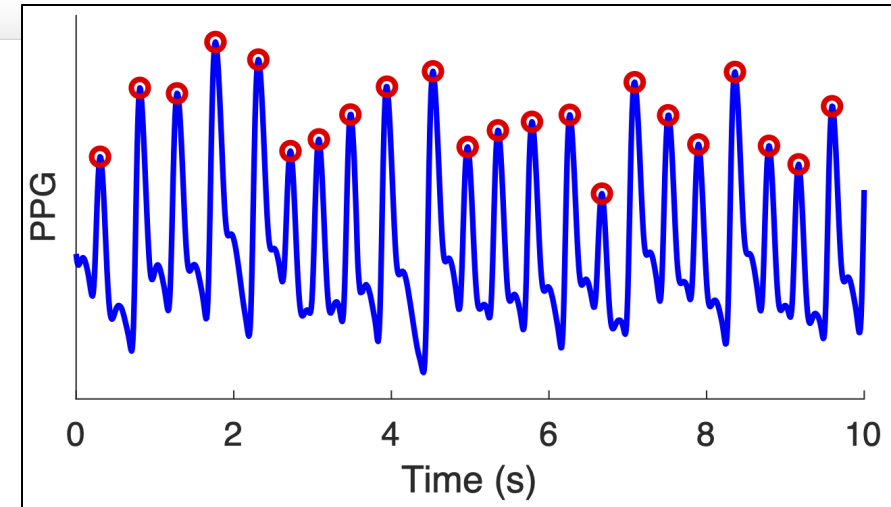
Original publication: Aboy M et al., An automatic beat detection algorithm for pressure signals. *IEEE Trans Biomed Eng* 2005; 52: 1662-70. DOI: [10.1109/TBME.2005.855725](https://doi.org/10.1109/TBME.2005.855725)

Description:

Link: [abd_beat_detector](#)

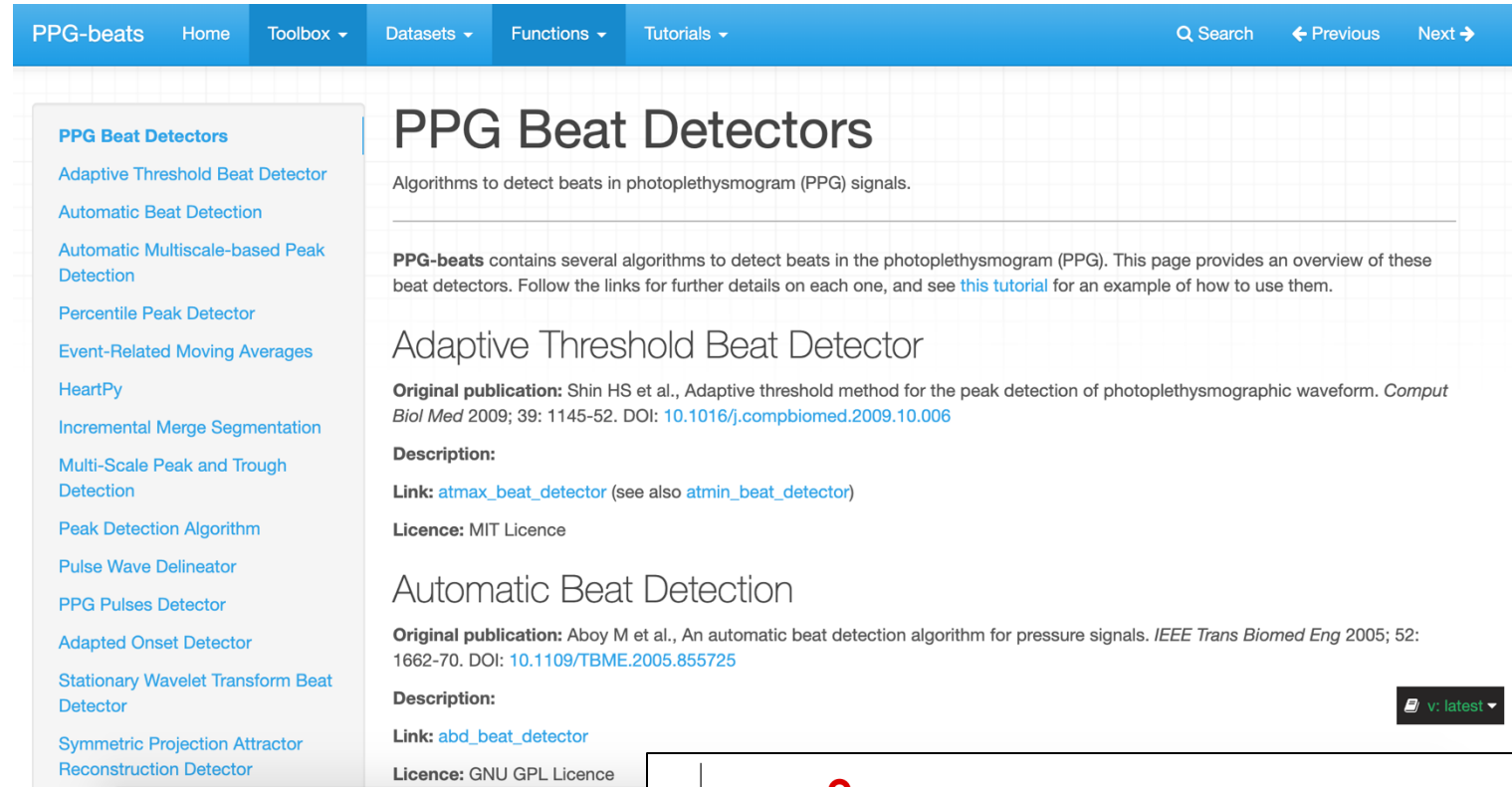
Licence: GNU GPL Licence

<https://ppg-beats.readthedocs.io/>



Open science

- Providing open-source toolboxes of algorithms
 - Ease of use
 - Programming language



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Description:

Link: [atmax_beat_detector](#) (see also [atmin_beat_detector](#))

Licence: MIT Licence

Automatic Beat Detection

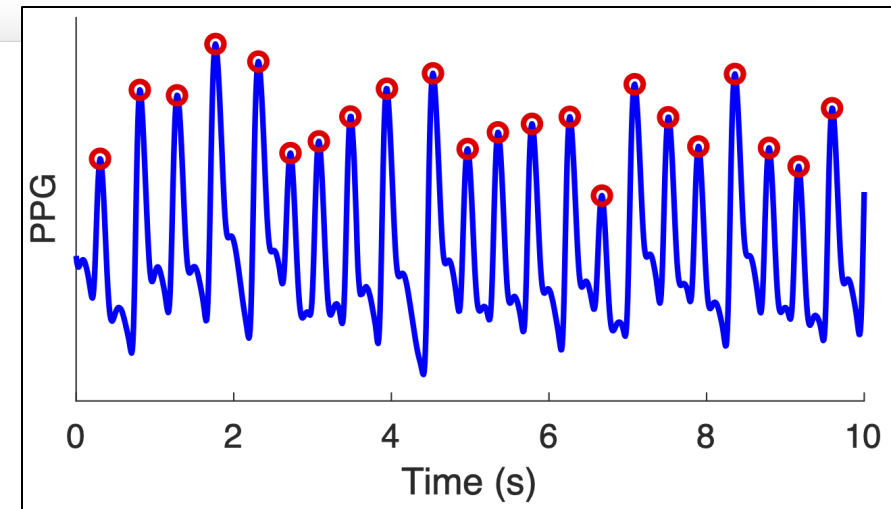
Original publication: Aboy M et al., An automatic beat detection algorithm for pressure signals. *IEEE Trans Biomed Eng* 2005; 52: 1662-70. DOI: [10.1109/TBME.2005.855725](https://doi.org/10.1109/TBME.2005.855725)

Description:

Link: [abd_beat_detector](#)

Licence: GNU GPL Licence

<https://ppg-beats.readthedocs.io/>

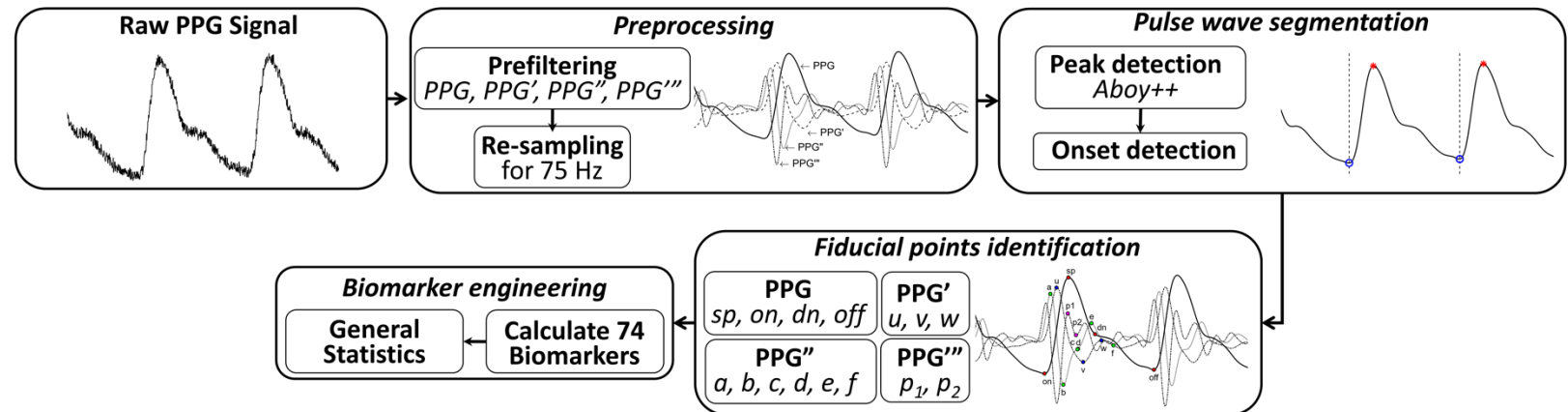


Open science

- Providing open-source toolboxes of algorithms
 - Ease of use
 - Programming language

pyPPG toolbox

A toolbox for finger photoplethysmogram (PPG) analysis, including beat detection, fiducial point detection, and comprehensive assessment of standard biomarkers.



<https://pyppg.readthedocs.io/>

Open science

- Providing open-source toolboxes of algorithms
 - Ease of use
 - Programming language
- Providing open datasets

Pulse Wave Database

A database of simulated pulse waves

Background

The rationale for a database of simulated pulse waves

The Database

Download the pulse wave database

Algorithms

Algorithms used to create and analyse the pulse wave database

Publications

Publications arising from or related to the project

Contributions

Details of how to contribute to the project

Acknowledgment

Thank you to all who have helped make this possible

News: The Pulse Wave Database can now be accessed here.

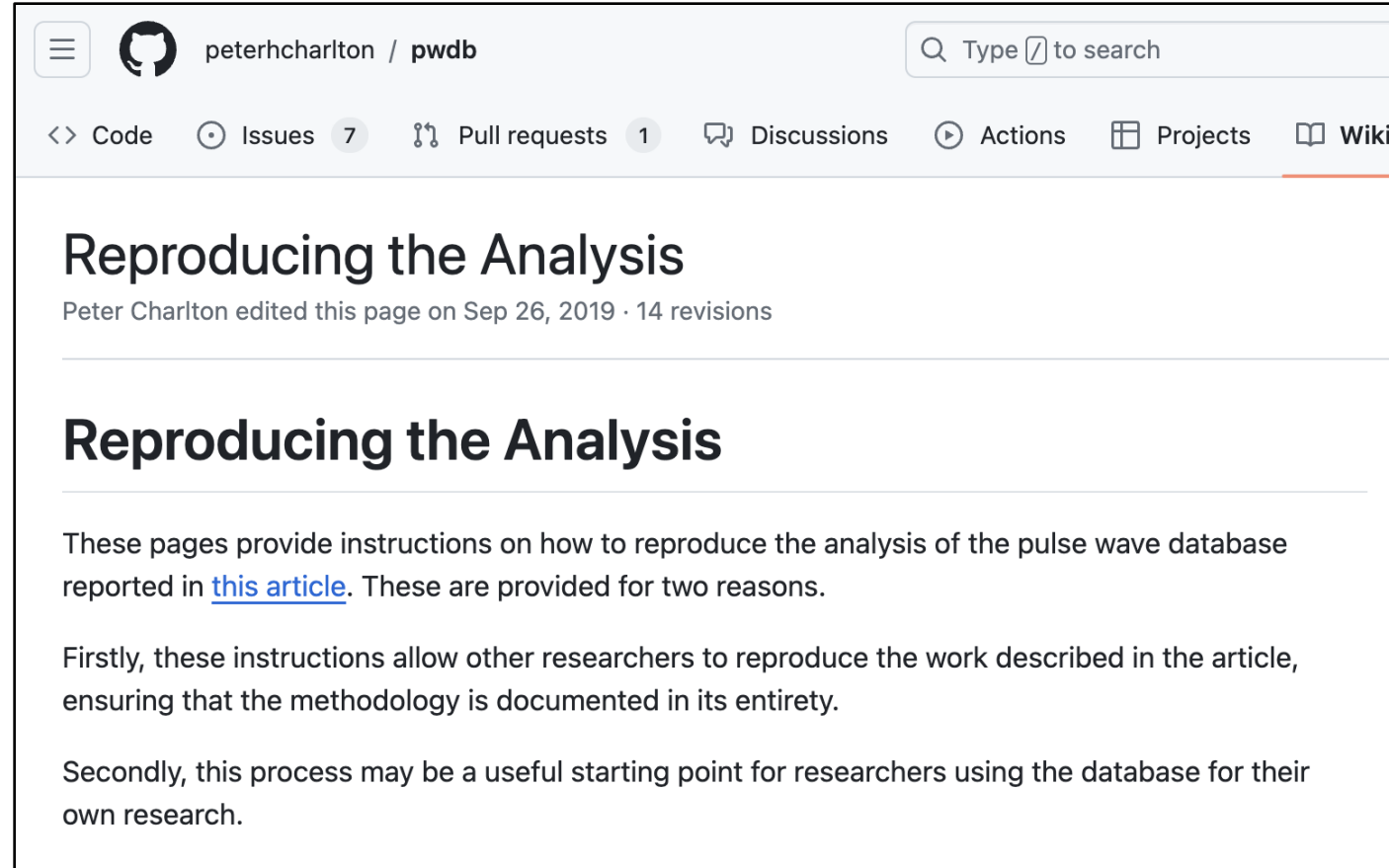
The Pulse Wave Database project

The aim of this project is to develop a database of simulated pulse waves for in silico testing of pulse wave analysis algorithms.

<https://peterhcharlton.github.io/pwdb>

Open science


- Providing open-source toolboxes of algorithms
 - Ease of use
 - Programming language
- Providing open datasets
- Reproducible research



The screenshot shows a GitHub repository page for 'peterhcharlton / pwdb'. The repository has 7 issues and 1 pull request. The main content is a README file titled 'Reproducing the Analysis', which was last edited by Peter Charlton on Sep 26, 2019, with 14 revisions. The README text reads: 'These pages provide instructions on how to reproduce the analysis of the pulse wave database reported in [this article](#). These are provided for two reasons. Firstly, these instructions allow other researchers to reproduce the work described in the article, ensuring that the methodology is documented in its entirety. Secondly, this process may be a useful starting point for researchers using the database for their own research.'

Open science

- Providing open-source toolboxes of algorithms
 - Ease of use
 - Programming language
- Providing open datasets
- Reproducible research
- Educational resources



Signal Processing and Learning for Wearables

Search this book...

- Introduction
- Overview
- Background
- Resources
- Tutorials
- Case Studies
- Summary
- About

Powered by [Jupyter Book](#)

Introduction

This book presents an introduction to processing wearable sensor data using signal processing techniques and machine learning.




Fig. 1 **A wearable device.** Photo by [Luke Chesser on Unsplash](#)

The book includes:

- **Overview:** The aims of the book, and details of accompanying workshops.
- **Background:** The background to wearable devices, including the signals they measure, the physiology behind the signals, and their applications.
- **Resources:** Details of the datasets and code used in the book.
- **Tutorials:** Interactive tutorials on signal processing and machine learning techniques.
- **Case Studies:** A case study on cuffless blood pressure estimation using the MIMIC Database.
- **Summary**

<https://peterhcharlton.github.io/bsp-book/>

Open science

- Providing open-source toolboxes of algorithms
 - Ease of use
 - Programming language
- Providing open datasets
- Reproducible research
- Educational resources
- Open access publications

Overview

This new textbook, titled *Photoplethysmography*, is due to be published in late 2021. Several experts in the field have contributed to the textbook, which is intended to provide a comprehensive summary of the theory, principles and technology of photoplethysmography.

Sample Chapters

The following sample chapters are available:

1. **Photoplethysmography signal processing and synthesis**: A comprehensive overview of signal processing techniques for the photoplethysmogram signal.
2. **Wearable photoplethysmography devices**: A comprehensive overview of the state-of-the-art of wearable photoplethysmography devices.

Link to Published Version

The published version of the textbook is available [here](#), and can be previewed [here](#). (Note that some of the images in the preview version are in black and white, whereas the sample chapters above contain full colour images.)

atrial fibrillation

cardiovascular monitoring

photoplethysmography

SAFER Wearables Study

wearables



Open science: an example

- Providing open-source toolboxes of algorithms
 - Ease of use
 - Programming language
- Providing open datasets
- Reproducible research
- Educational resources
- Open access publications

qppg

- Openly available
- GNU GPL Licence
- Highly efficient

MSPTD

- Openly available
- MIT Licence
- Less efficient

MSPTDfast



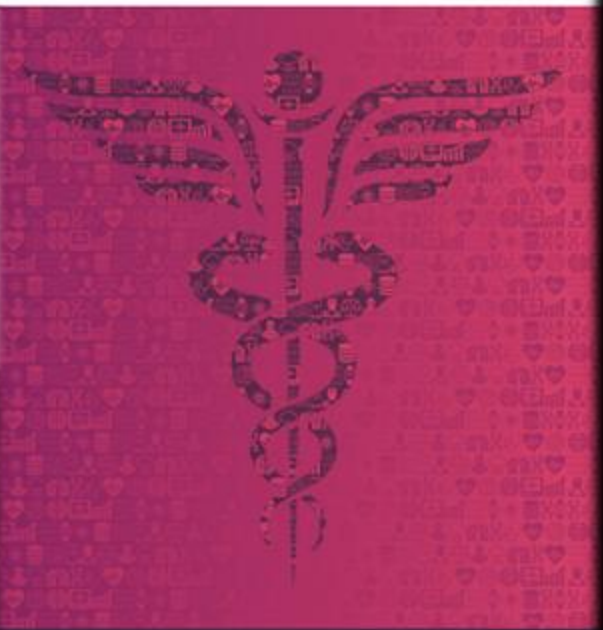
The NEW ENGLAND JOURNAL of MEDICINE

CORRESPONDENCE



Racial Bias in Pulse Oximetry Measurement

Equity in Medical Devices:
Independent Review



Report 2023

With thanks to...

Prof Jonathan Mant
Prof Panicos Kyriacou

The SAFER Research Team
University of Cambridge
City, University of London
King's College London

British Heart Foundation
NIHR
EPSRC

... and many, many others



Photoplethysmography is now widely used in wearable devices, with many potential applications.

Photoplethysmography-based wearables have shown great promise in certain applications such as detecting atrial fibrillation.

However, there is much work to do to ensure the potential benefits of photoplethysmography are available to all, making photoplethysmography-based devices as reliable as a climbing rope.



Equitable Photoplethysmography in Wearables: Accurate Data for All

Dr Peter H. Charlton

<https://peterhcharlton.github.io>

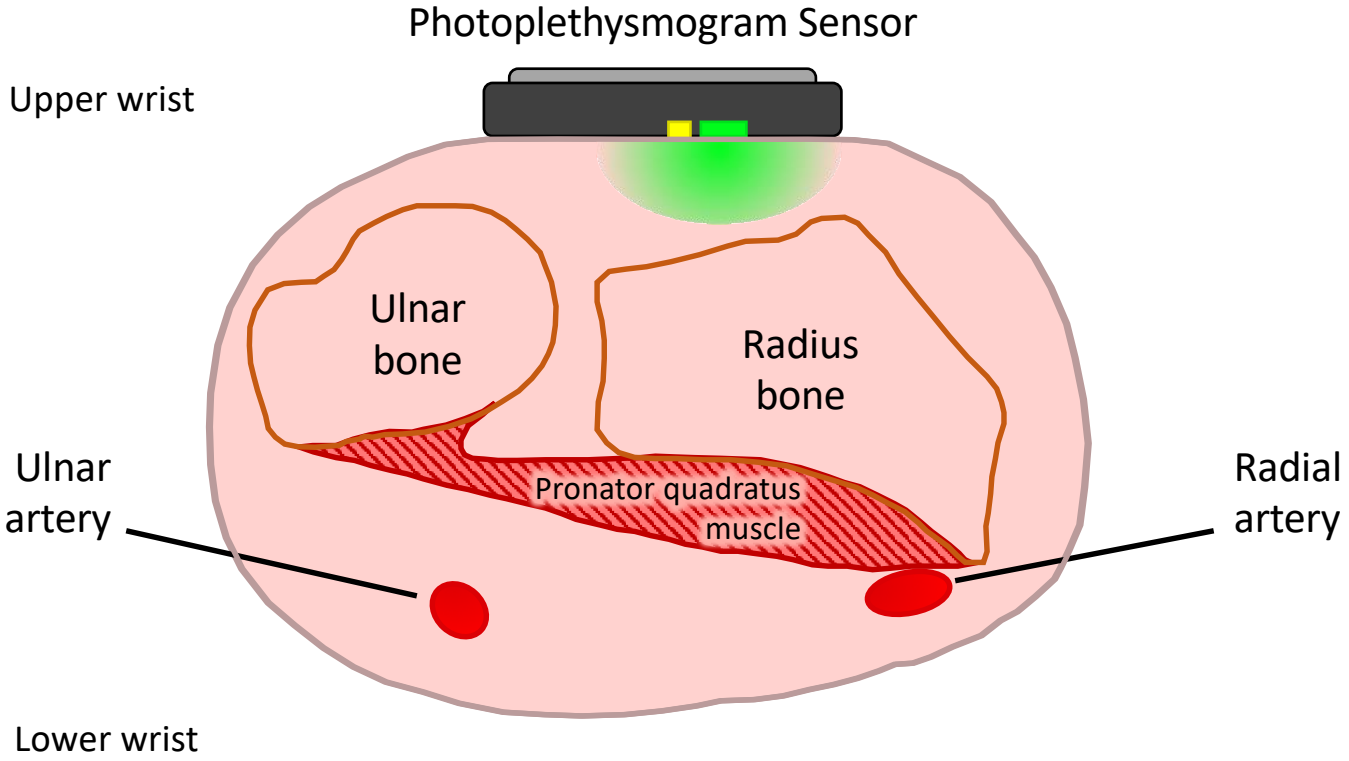
pc657@cam.ac.uk

Slides available at: <https://doi.org/10.5281/zenodo.13833790> (CC BY 4.0)

Wrist photoplethysmography: limitations



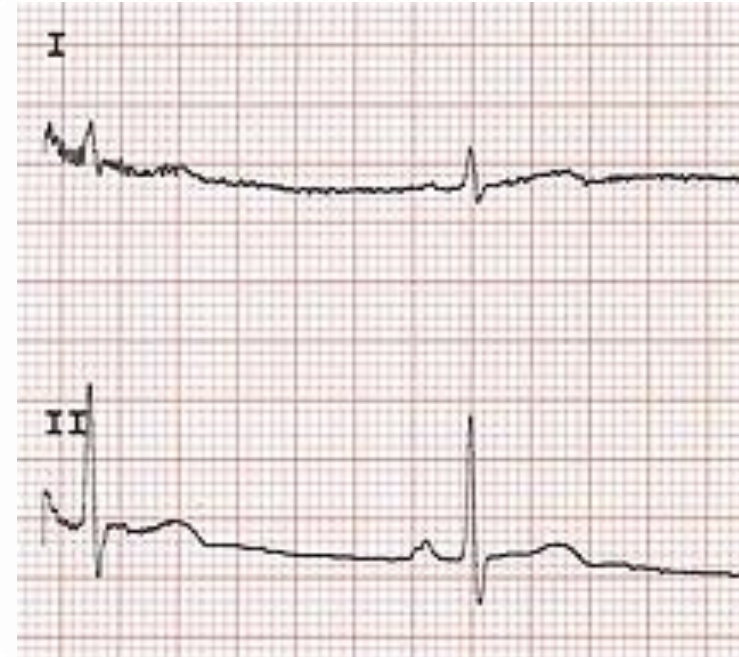
Photoplethysmogram (PPG) Sensor



Single-lead ECG: limitations



- Dry electrodes
- Recorded without clinical supervision
- Lead I



“Sinus P waves are usually most prominently seen in leads II and V1”

Meek S and Morris F, 'ABC of clinical electrocardiography: Introduction. II—basic terminology', <https://doi.org/10.1136/bmj.324.7335.470>