

Using consumer wearables in healthcare and research

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Pexels, https://pixabay.com/photos/man-heartache-chest-pain-hurt-pain-1846050/ (Pixabay Licence)





(artistic license – this heart rate is fictional)

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Consumer Wearables



Challenges & Opportunities

CFCF, https://commons.wikimedia.org/wiki/File:Normal-heart.jpg (CC0 1.0)

https://commons.wikimedia.org/wiki/File:Apple Watch user (Unsplash).jpg (CC0 1.0)

Luke Chesser,

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Next Steps

Brother UK,























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



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

The Photoplethysmogram



Photoplethysmogram (PPG) Sensor



Peter Charlton

The Photoplethysmogram

2 3 4 0 Time (s) Slower-heart rate (heart rate = 80 bpm) Photoplethysmogram (PPG) Sensor 0 2 3 4 Time (s) Atrial fibrillation (heart rate = 140 bpm) Wrist crosssection 2 3 0 1 4

Photoplethysmogram

Time (s)

(heart rate = 100 bpm)

5

5

5



Charlton PH et al., Wearable photoplethysmography for cardiovascular monitoring, Proc. IEEE, 2022. https://doi.org/10.1109/JPROC.2022.3149785 (CC BY 4.0)

Peter Charlton, PhD Thesis (CC BY 4.0)

The Photoplethysmogram





- Simple structure
- High resolution
- · Wide response range
- · Serious hysteresis
- Complex production process
- High temperature error

Piezoelectric

- Wide frequency response range
- High sensitivity
- High signal to noise ratio
- Simple structure
- · Easy leakage of electric charge
- Dynamic signals only

Capacitive

- High temperature stability
- Low electrostatic gravitational force
- Good dynamic response
- High impedance

0

- Poor load capacity
- · High parasitic capacitance effect



- High accuracy
- Long service life
- high reliability
- · High resolution
- High optical coupling
- Vulnerable to perspiration

Advantages:

- Not limited to arteries
- Oxygen saturation
- (non-contact)

Disadvantages:

- Power hungry
- Noisy

2 3 Time (s)

4

Triboelectric

- · High sensitivity
- Low production costs
- Simple construction
- Wide range of materials
- Dynamic signals only
- High inner impedance

Magnetoelastic

- Intrinsic waterproofness
- · High stretchability
- High biocompatibility
- Low inner impedence
- Susceptible to magnetic fields interference

Meng K et al., Wearable Pressure Sensors for Pulse Wave Monitoring, Advanced Materials, 2022. https://doi.org/10.1002/adma.202109357 (copyright)

Alternative sensing technologies

Potential alternatives to photoplethysmography:

- *Triboelectric sensing
- *Piezoresistive sensing
- *Piezoelectric sensing
- *Capacitive sensing
- *Magnetoelastic sensing
- Magnetic induction plethysmography
- Doppler ultrasound and acoustic sensing
- Impedance plethysmography
- Speckle plethysmography

* See: Meng K et al., <u>https://doi.org/10.1002/adma.202109357</u>

Alternative form factors

• E-textiles





lakovlev D et al., <u>https://doi.org/10.3390/bios8020033</u> (CC BY 4.0)

Alternative form factors

- E-textiles
- Miniaturize inorganic sensors





lakovlev D et al., <u>https://doi.org/10.3390/bios8020033</u> (CC BY 4.0)

Alternative form factors

- E-textiles
- Miniaturize inorganic sensors
- Organic, flexible sensors



Khan Y et al., https://doi.org/10.1109/ACCESS.2019.2939798 (CC BY 4.0)



Potential Applications





Peter Charlton, Wikimedia Commons, <u>https://commons.wikimedia.org/wiki/File:Max_Health_Band.jpg</u> (<u>CC BY 4.0</u>)

Further Reading

Broad overview:

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

Review article:

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

Textbook chapter:

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

Consumer Wearables

MMM

Challenges & Opportunities

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https://www.flickr.com/photos/brother-uk/31501281284/in/photostream/ (CC BY 2.0)

Next Steps

Brother UK,

The photoplethysmogram contains a wealth of cardiovascular information



Black – mean value for 25 year old; Red – 1 standard deviation above mean; Blue – 1 standard deviation below mean

Charlton PH et al., Wearable photoplethysmography for cardiovascular monitoring, Proc. IEEE, 2022. https://doi.org/10.1109/JPROC.2022.3149785 (CC BY 4.0)

Many people use wearables ...



Many people use wearables ... in daily life



Source: Charlton et al., Acquiring wearable photoplethysmography data in daily life: the PPG Diary Study, https://doi.org/10.3390/ecsa-7-08233 (CC BY 4.0)

Opportunities afforded by wearables in research:



Opportunities afforded by wearables in healthcare:



Adapted from: Balayla J, Prevalence threshold (φe) and the geometry of screening curves, PLoS ONE, https://doi.org/10.1371/journal.pone.0240215 (CC BY 4.0)

Opportunities afforded by wearables in healthcare:



Assessing performance and acceptability of wearables for detecting atrial fibrillation

3. If you removed any device before the end, then why?				
I didn't remove them before the end	Chest-patch			
I removed the chest-patch because:				
I removed the wristband because:	Wristband			
I removed the watch because:	Watch OR			



Adapted from: Balayla J, Prevalence threshold (φe) and the geometry of screening curves, PLoS ONE, https://doi.org/10.1371/journal.pone.0240215 (CC BY 4.0)

Many people use wearables ... in daily life



Our Contributions:

- Established threshold levels of activity below which interbeat intervals can be accurately measured from the photoplethysmogram.
- Compared beat detection algorithms to identify the best performing algorithm for use in daily life.



P. H. Charlton, "Accelerometry-Guided Inter-Beat-Interval Assessment from Wrist Photoplethysmography," https://doi.org/10.5281/zenodo.8403222 (CC BY 4.0)

Many people use wearables

Performance across different subjects

Our Contribution:

Al-Halawani R. *et al.*, **A review of the effect of skin pigmentation on pulse oximeter** accuracy, *Phys Meas.*, 2023, <u>https://doi.org/10.1088/1361-6579/acd51a</u>

Cardiovascular determinants of the photoplethysmogram



Black – mean value for 25 year old Red – 1 standard deviation above mean

Blue – 1 standard deviation below mean

Case Study: Detecting hypertension



- Much research on estimating blood pressure from the photoplethysmogram
- Mixed performance
- Estimating absolute blood pressure vs. detecting potentially elevated blood pressure

Case Study: Detecting obstructive sleep apnea



Key publications:

 Behar *et al.*, <u>Feasibility of single channel</u> <u>oximetry for mass screening of</u> <u>obstructive sleep apnea</u>, *EClinicalMedicine*, 2019
Behar *et al.*, <u>Single-channel oximetry</u> <u>monitor versus in-lab polysomnography</u> <u>oximetry analysis: Does it make a</u> <u>difference?</u>, *Physiological Measurement*, 2020





fancycrave1, https://commons.wikimedia.org/wiki/File:Apple Watch-.jpg (CC0 1.0)



Perez M.V. *et al.* Large-scale assessment of a smartwatch to identify atrial fibrillation. *N. Engl. J. Med.* **2019**, *381*, 1909–1917, doi:<u>10.1056/NEJMoa1901183</u>



fancycrave1, https://commons.wikimedia.org/wiki/File:Apple Watch-.jpg (CC0 1.0)



Lubitz S. *et al.* Detection of Atrial Fibrillation in a Large Population using Wearable Devices: The Fitbit Heart Study. *Circulation.* **2022**, doi: <u>10.1161/CIR.000000000001041</u>



fancycrave1, https://commons.wikimedia.org/wiki/File:Apple Watch-.jpg (CC0 1.0)

Our Contributions:

• Using photoplethysmography to detect inter-beat intervals

Our Contributions:

- Using photoplethysmography to detect inter-beat intervals
- Identifying AF reliably from inter-beat intervals



Our Contributions:

- Using photoplethysmography to detect inter-beat intervals
- Identifying AF reliably from inter-beat intervals
- Diagnosis of AF from single-lead ECGs



Case Study Summary

Case Study	Technology	Clinical Setting	Confirmation	Potential benefits
Detecting hypertension	Difficult to estimate blood pressure from photoplethysmography	established – detection of hypertension in daily life	Clear – subsequent cuff measurement	Ubiquitous Early detection In daily life
Detecting obstructive sleep apnea	Challenges in oxygen saturation assessment (reflectance, skin type)	new – opportunistic detection at home	Unclear	Ubiquitous Early detection
Atrial fibrillation (AF) detection	(relatively) Straight- foward implementation	(relatively) new – opportunistic screening for asymptomatic AF	Emerging – simultaneous ECG on wearable device	Ubiquitous Early detection Asymptomatic AF

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Next Steps

Brother UK,

Selecting a clinical use case

Consider:

- Would a wearable add value to current practice? *e.g.*
 - Physiological assessment where it would not otherwise be performed
 - Frequent monitoring where measurements would otherwise be intermittent
- Would the results be actionable? *e.g.*
 - Prompt further assessment (relatively safe)
 - Diagnosis (and treatment) (higher risk)
- Could it be integrated into a clinical pathway?
- Is it cost-effective?
- Many use cases focus on "prevention of avoidable illness and its exacerbations" (NHS Long Term Plan)

Advancing the technology

Several potential directions:

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

Open-source algorithms:

PPG-beats Home Toolbox -	Datasets • Functions • Tutorials •	Q Search	← Previous Next →			
PPG Beat Detectors Adaptive Threshold Beat Detector Automatic Beat Detection	PPG Beat Detectors Algorithms to detect beats in photoplethysmogram (PPG) signals.					
Automatic Multiscale-based Peak Detection	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.					
Percentile Peak Detector Event-Related Moving Averages	Adaptive Threshold Beat Detector					
HeartPy Incremental Merge Segmentation	Original publication: Shin HS et al., Adaptive th Biol Med 2009; 39: 1145-52. DOI: 10.1016/j.com		9 9			
Multi-Scale Peak and Trough Detection	Description: Link: atmax_beat_detector (see also atmin_beat.	ព្រំព័ ំ	Î P Po			
Peak Detection Algorithm	Licence: MIT Licence					
PPG Pulses Detector		VUYYY	U VYY VUUY			
Adapted Onset Detector Stationary Wavelet Transform Beat	Decontellary	,				
Detector Symmetric Projection Attractor	Link: abd_beat_detector 0 2 4	6 (s)	8 1			
Reconstruction Detector	Licence: GNU GPL Licence	10 (3)				

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

Respiratory Rate Estimation

Research into estimation of respiratory rate from physiological signal

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



https://peterhcharlton.github.io/RRest

Validating devices

The INTERLIVE Network:

Goal: "To develop gold standard protocols for the validation of wearables in order to improve the accuracy and reliability of physical activity patterns assessment."

Validation protocols published for:

- Step counts
- Heart rate
- Energy expenditure
- Maximal oxygen consumption

See: https://www.interlive.org/

Cuffless blood pressure estimation:

- New IEEE standard being developed
- Requiring a new approach

See: Mukkamala et al., Hypertension, 2020

Understanding the needs of stakeholders



Identifying potential clinical pathways

- Screening
- Patient-led measurements to prompt clinical assessment *e.g.*
 - Bradycardia as sign of possible heart block: https://doi.org/10.1016/j.jaccas.2019.11.087
 - Smartwatch ECG capturing ventricular tachycardia: https://doi.org/10.1016/j.cjco.2021.12.003
 - Incidental findings during follow-up of smartwatch-detected tachycardia:

https://doi.org/10.1177/20542704211068651

- Identifying side-effects of treatments, e.g.
 - Bradycardia or arrhythmia after medication: <u>https://doi.org/10.1177/23247096211069761</u>
- Population-level surveillance
- Self-directed health monitoring

Large-scale studies

Heart rate variability with photoplethysmography in 8 million individuals: a cross-sectional study

Aravind Natarajan, Alexandros Pantelopoulos, Hulya Emir-Farinas, Pradeep Natarajan

https://doi.org/10.1016/S2589-7500(20)30246-6

ORIGINAL ARTICLE

Large-Scale Assessment of a Smartwatch to Identify Atrial Fibrillation

Marco V. Perez, M.D., Kenneth W. Mahaffey, M.D., Haley Hedlin, Ph.D., John S. Rumsfeld, M.D., Ph.D., Ariadna Garcia, M.S., Todd Ferris, M.D., Vidhya Balasubramanian, M.S., Andrea M. Russo, M.D., Amol Rajmane, M.D., Lauren Cheung, M.D., Grace Hung, M.S., Justin Lee, M.P.H., Peter Kowey, M.D., Nisha Talati, M.B.A., Divya Nag, Santosh E. Gummidipundi, M.S., Alexis Beatty, M.D., M.A.S., Mellanie True Hills, B.S., Sumbul Desai, M.D., Christopher B. Granger, M.D., Manisha Desai, Ph.D., and Mintu P. Turakhia, M.D., M.A.S., for the Apple Heart Study Investigators*

https://doi.org/10.1056/NEJMoa1901183

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EPSRC British Heart Foundation

The 50 co-authors of 'The 2023 wearable photoplethysmography roadmap', who greatly helped shape my thinking on this topic.

Wearables can provide a wealth of physiological information unobtrusively.

Much work is required to understand how to extract physiological data reliably, and how to use it in healthcare and research.

Perhaps in the future, wearables could be as essential in maintaining health as a rope is for climbing safely.



Using consumer wearables in healthcare and research

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

Slides available at: <u>https://doi.org/10.5281/zenodo.10517773</u>

Peter Charlton, Wikimedia Commons, https://commons.wikimedia.org/wiki/File:Max_Health_Band.jpg (CC BY 4.0)