




# Wearable photoplethysmography devices for health monitoring

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Slides available at: <https://doi.org/10.5281/zenodo.7301839> (CC BY 4.0)



**Every 30 mins  
in UK:**

**6**

**hospital  
admissions**

**10**

**deaths**

**€500,  
000**

**healthcare  
costs**

Stress level  
VO2 max  
Heart rate  
Respiratory rate  
Blood pressure  
Vascular age  
Arrhythmias  
Oxygen saturation  
Sleep assessment  
Energy expenditure



Infamous diseases  
Sepsis  
Heart failure  
COPD  
Asthma  
Atrial fibrillation  
Orthostatic hypotension  
Obstructive sleep apnea  
Seizure detection  
Sleep monitoring  
Pre-eclampsia  
Mental stress  
Chronic kidney disease  
Vascular age  
Menstrual cycle  
Clinical deterioration  
Health insurance  
CV risk prediction  
Biometric authentication  
Fitness tracking





St. Thomas' Hospital



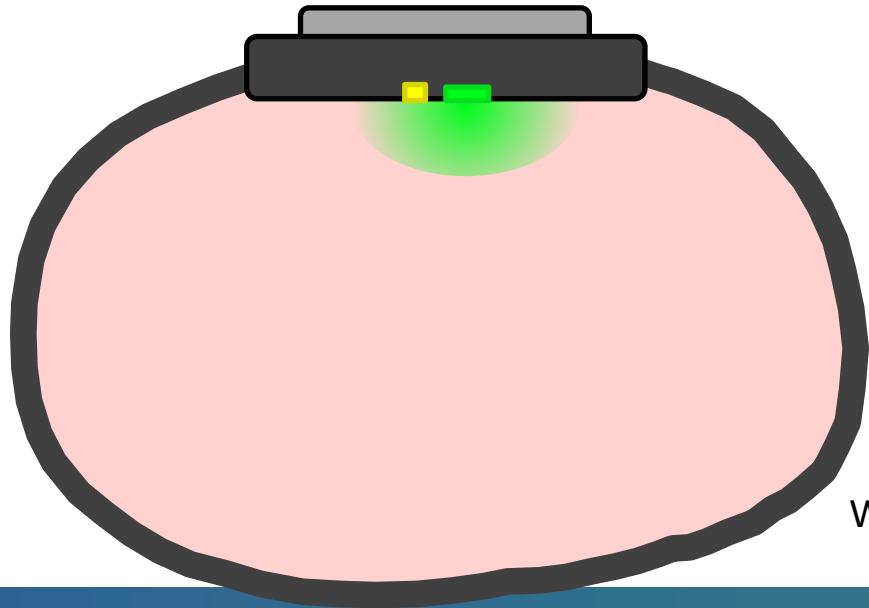


(artistic licence – the device didn't show this heart rate)

# The Photoplethysmogram



Photoplethysmogram (PPG) Sensor

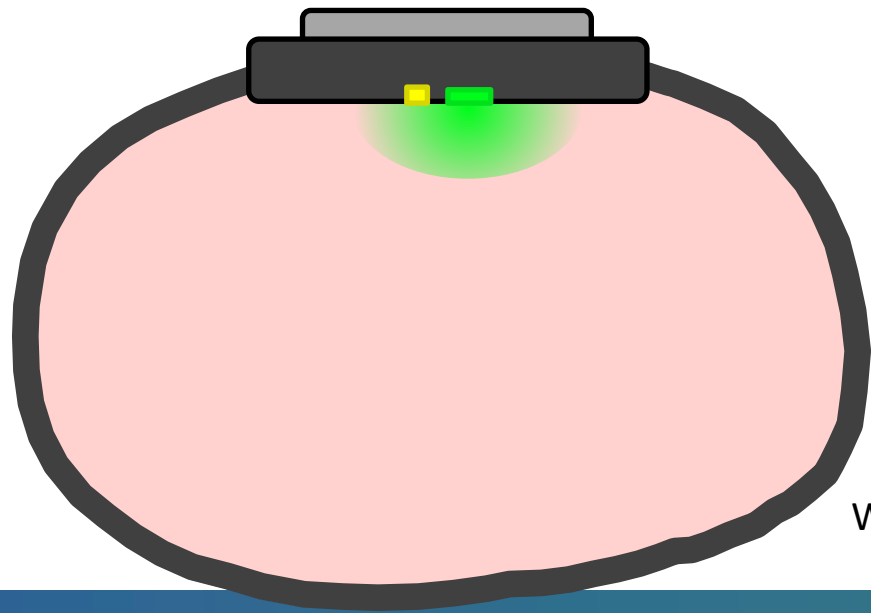


Wrist cross-section

# The Photoplethysmogram



Photoplethysmogram (PPG) Sensor



Wrist cross-section

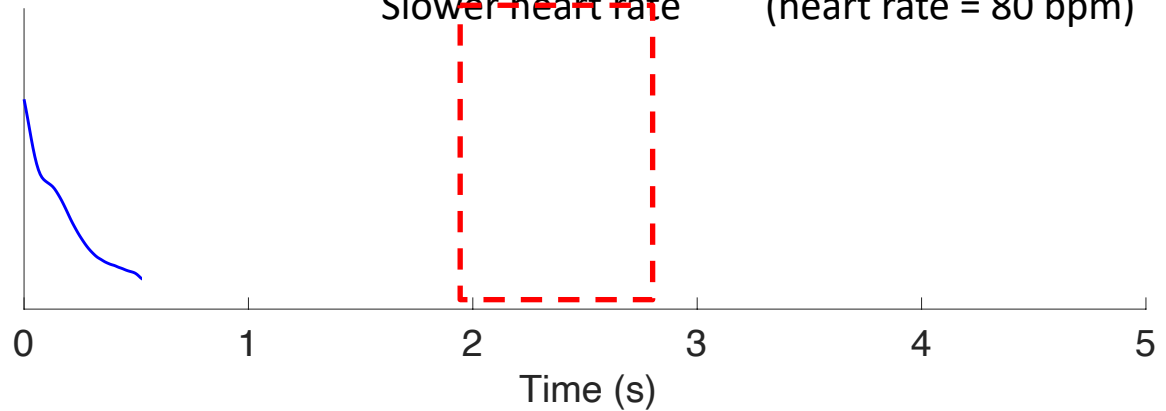
Photoplethysmogram

(heart rate = 100 bpm)



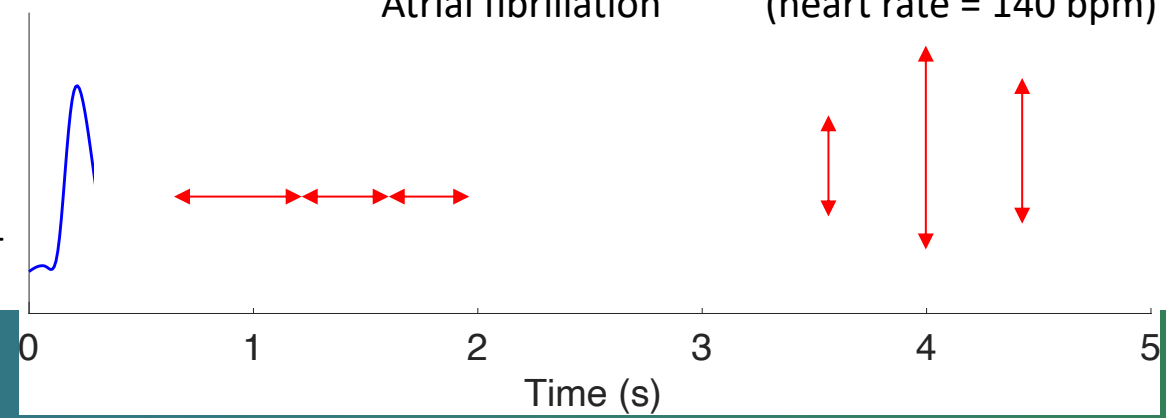
Slower heart rate

(heart rate = 80 bpm)



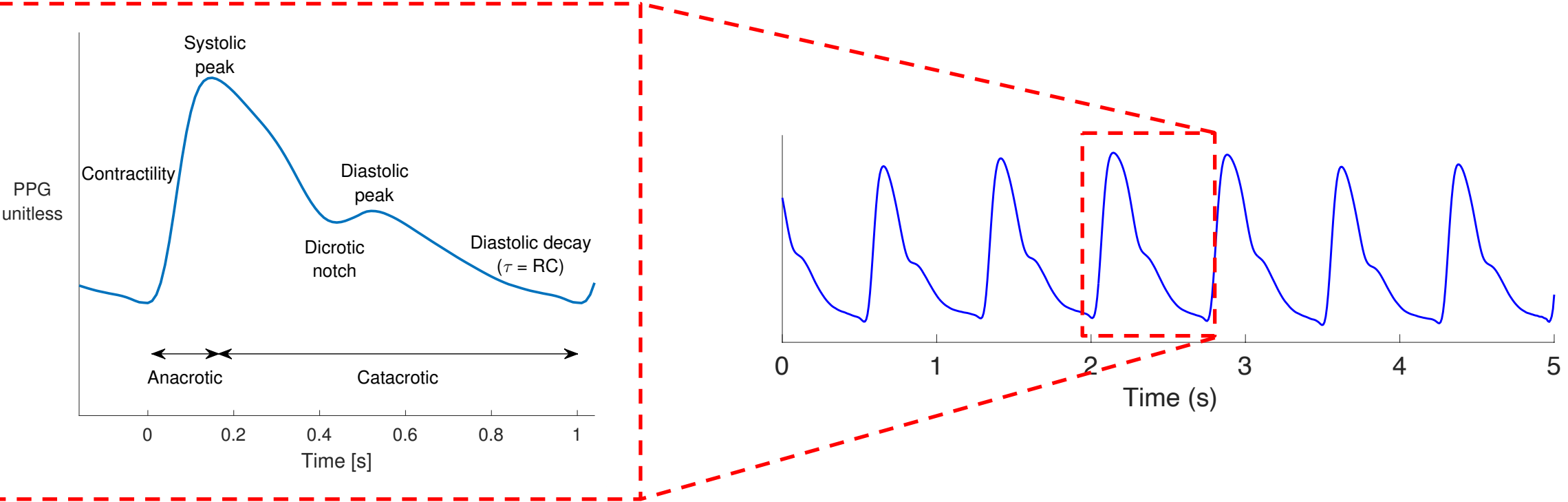
Atrial fibrillation

(heart rate = 140 bpm)

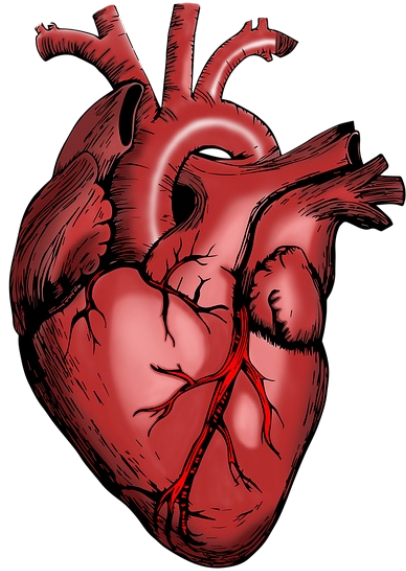




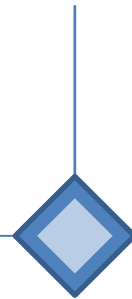
# The Photoplethysmogram



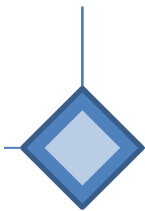
## Applications



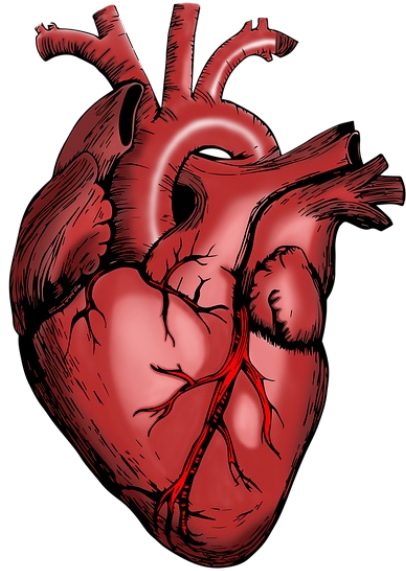
## Research and Development



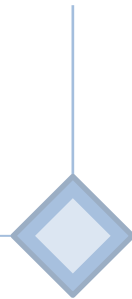
## Challenges



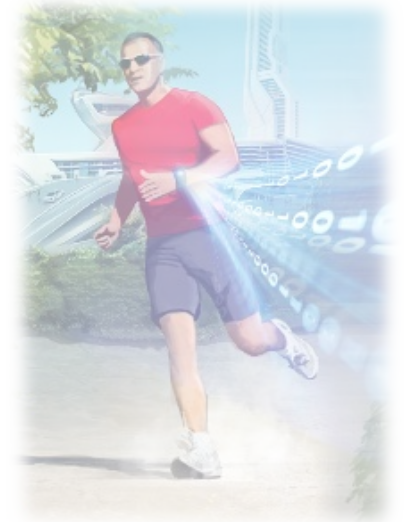
## Applications



## Research and Development



## Challenges





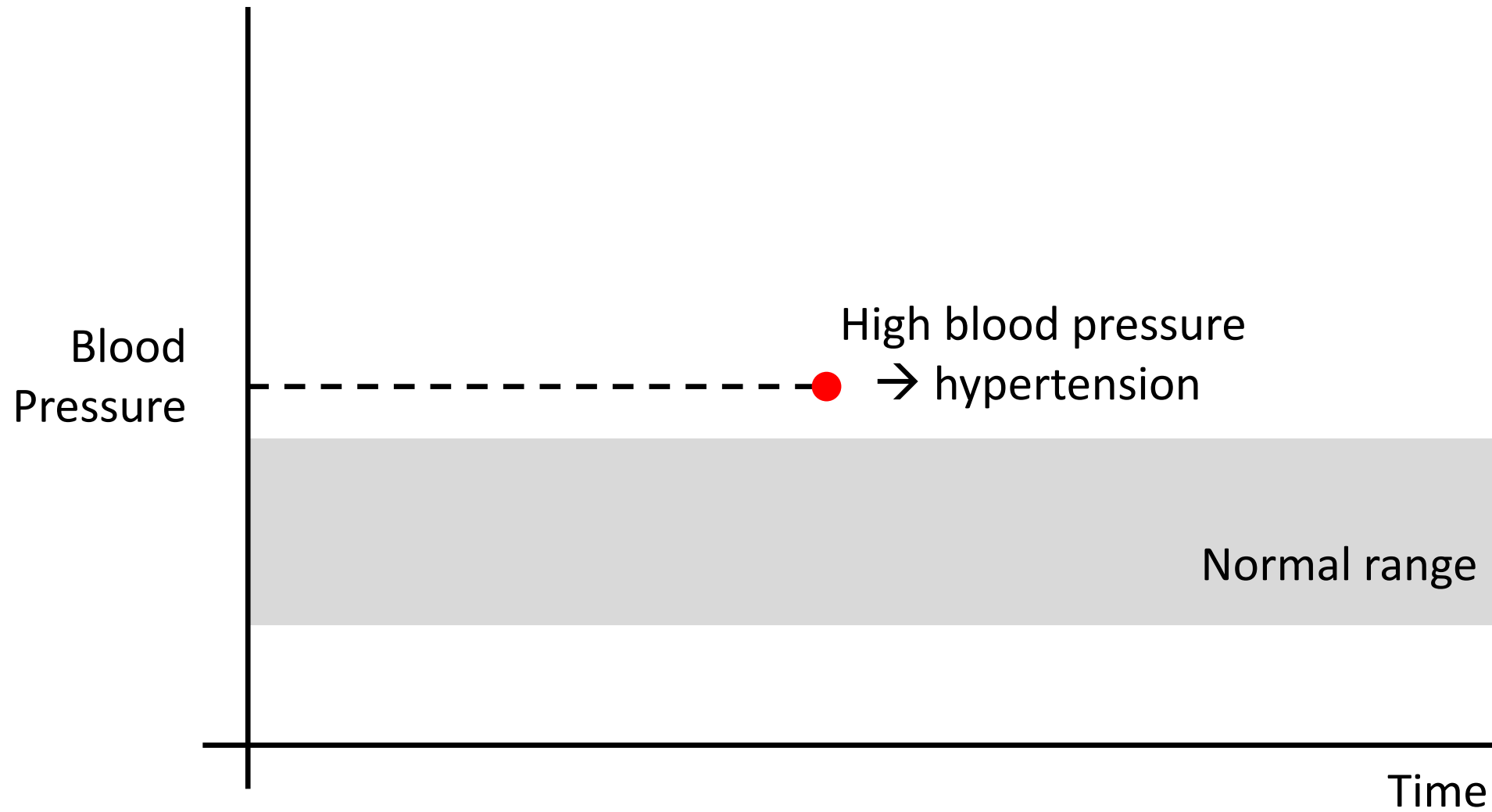
# cardiovascular monitoring



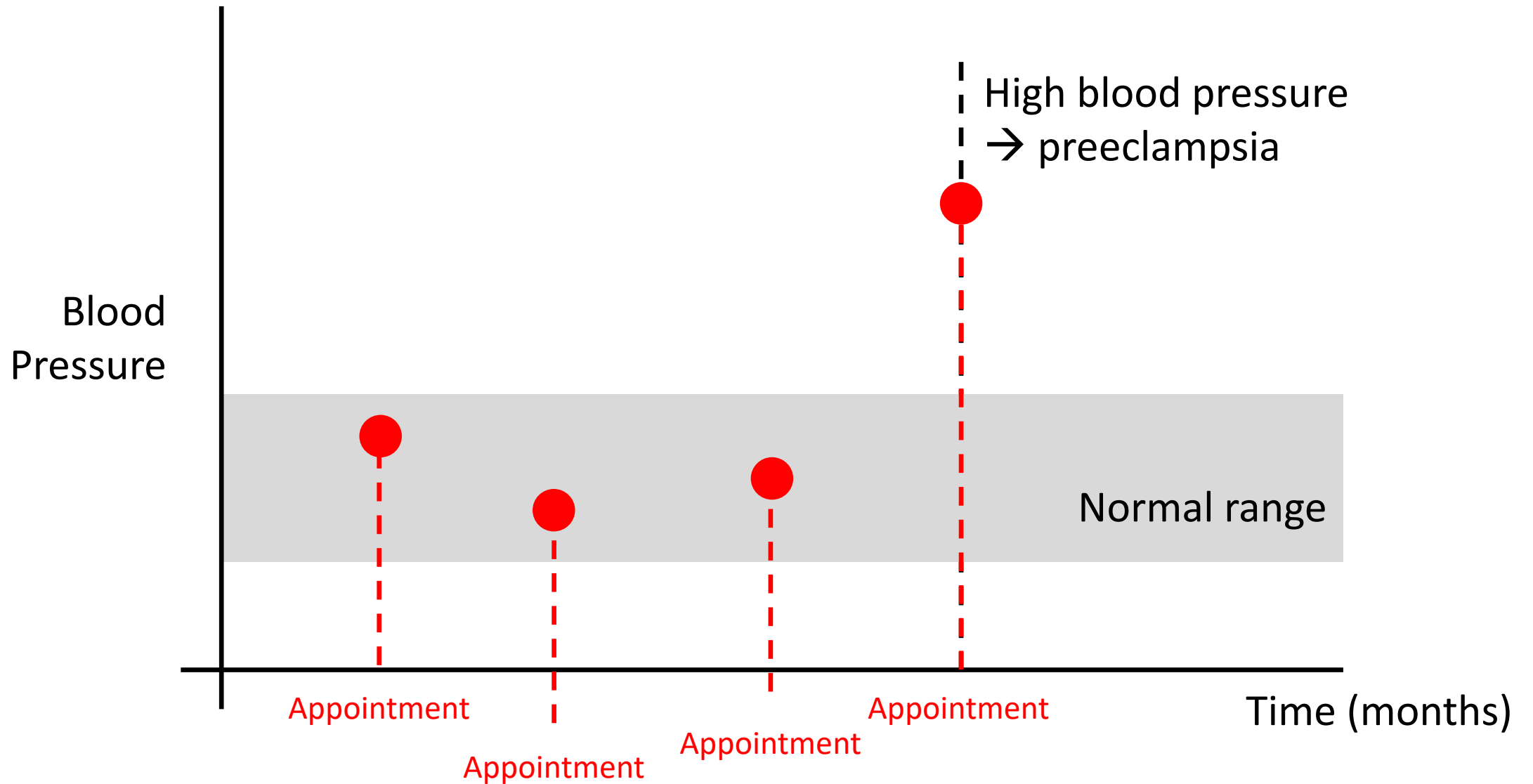
Blood  
Pressure

Time

# Identifying hypertension in a middle-aged man:



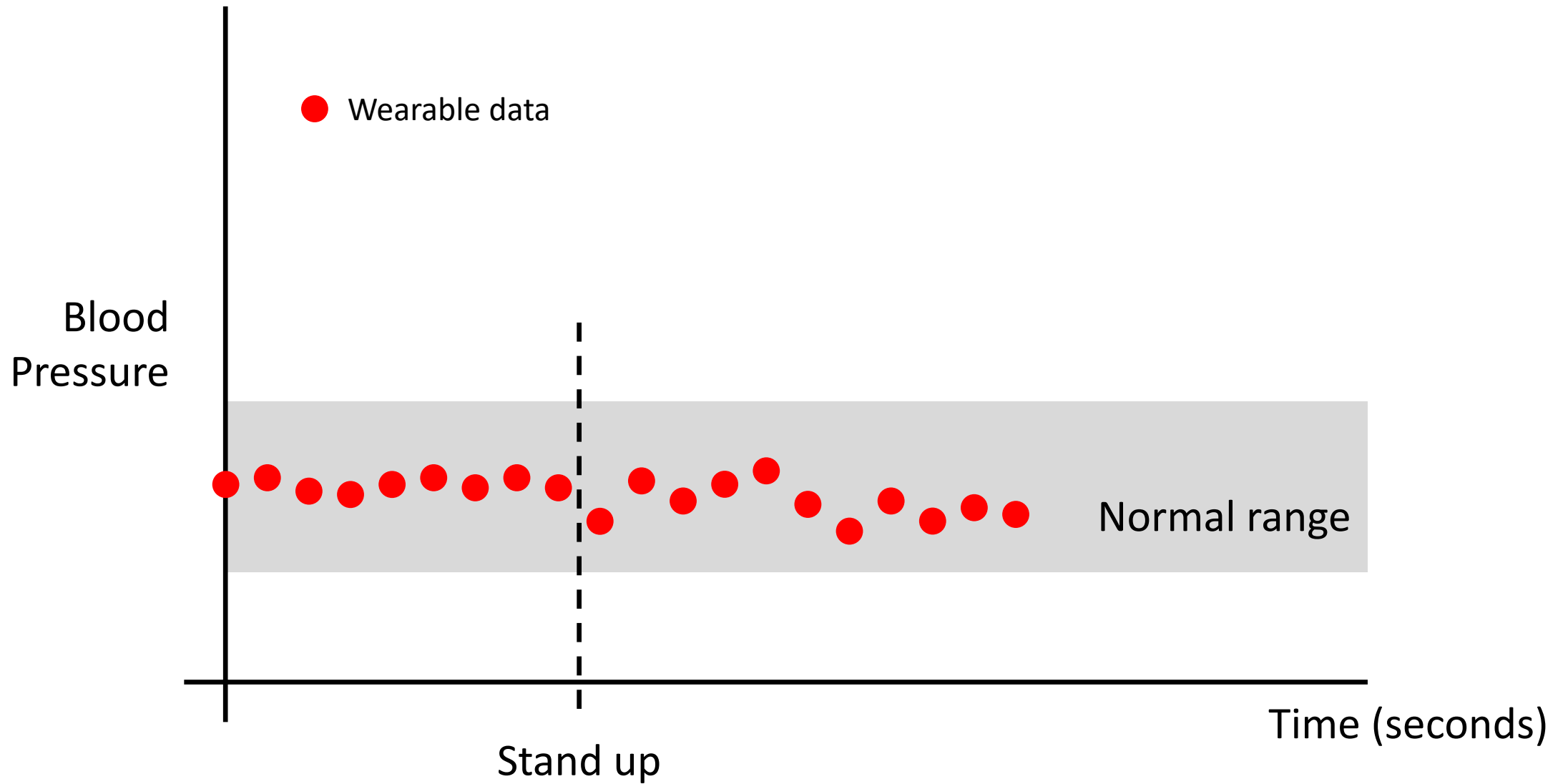
# Identifying early signs of preeclampsia in a pregnant woman:



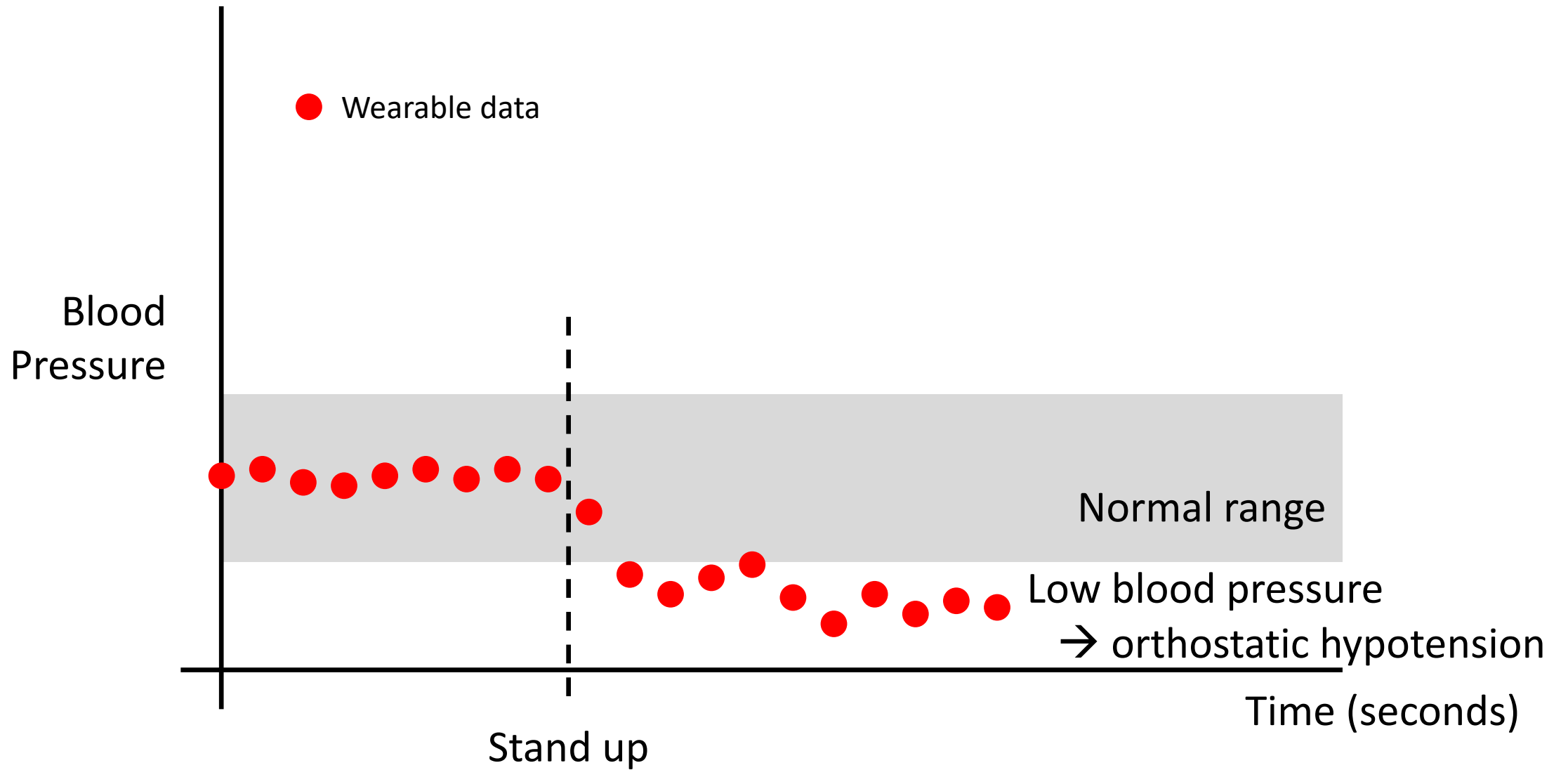




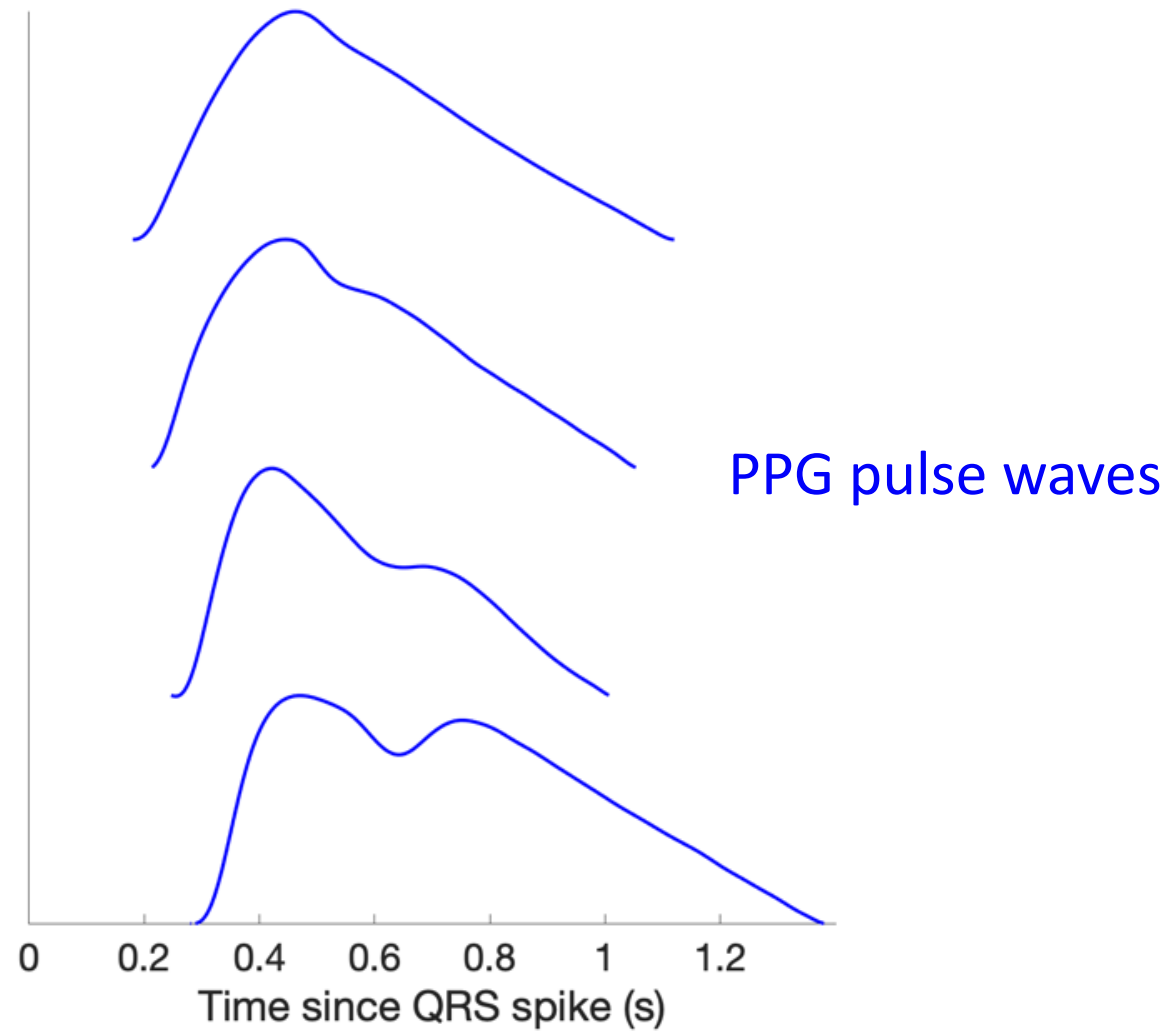
# Identifying orthostatic hypotension in an older adult, indicating risk of falls:



# Identifying orthostatic hypotension in an older adult, indicating risk of falls:

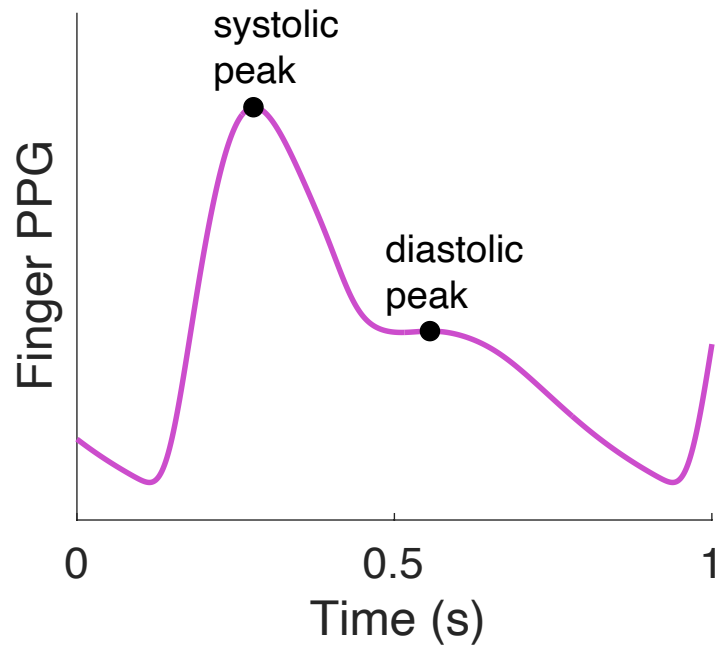


Pulse wave shape changes with blood pressure:

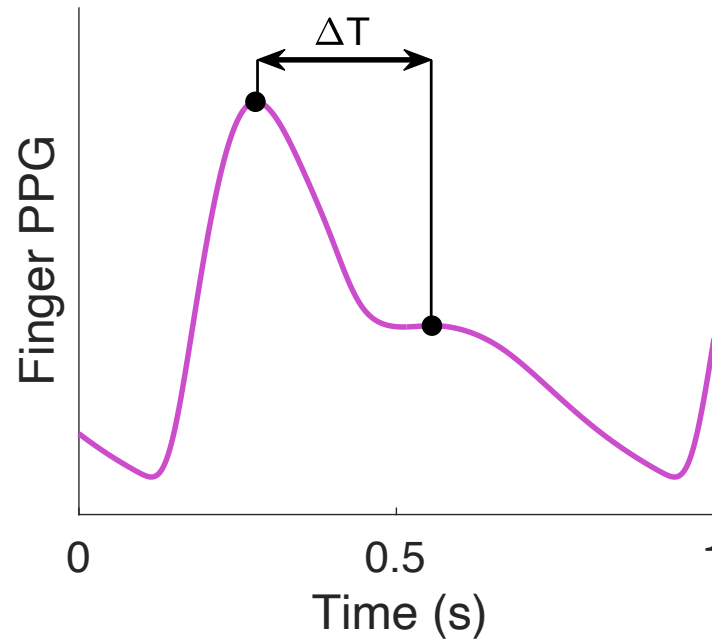


# Estimating blood pressure from a single PPG signal

i) Identify fiducial points



ii) Extract pulse wave feature



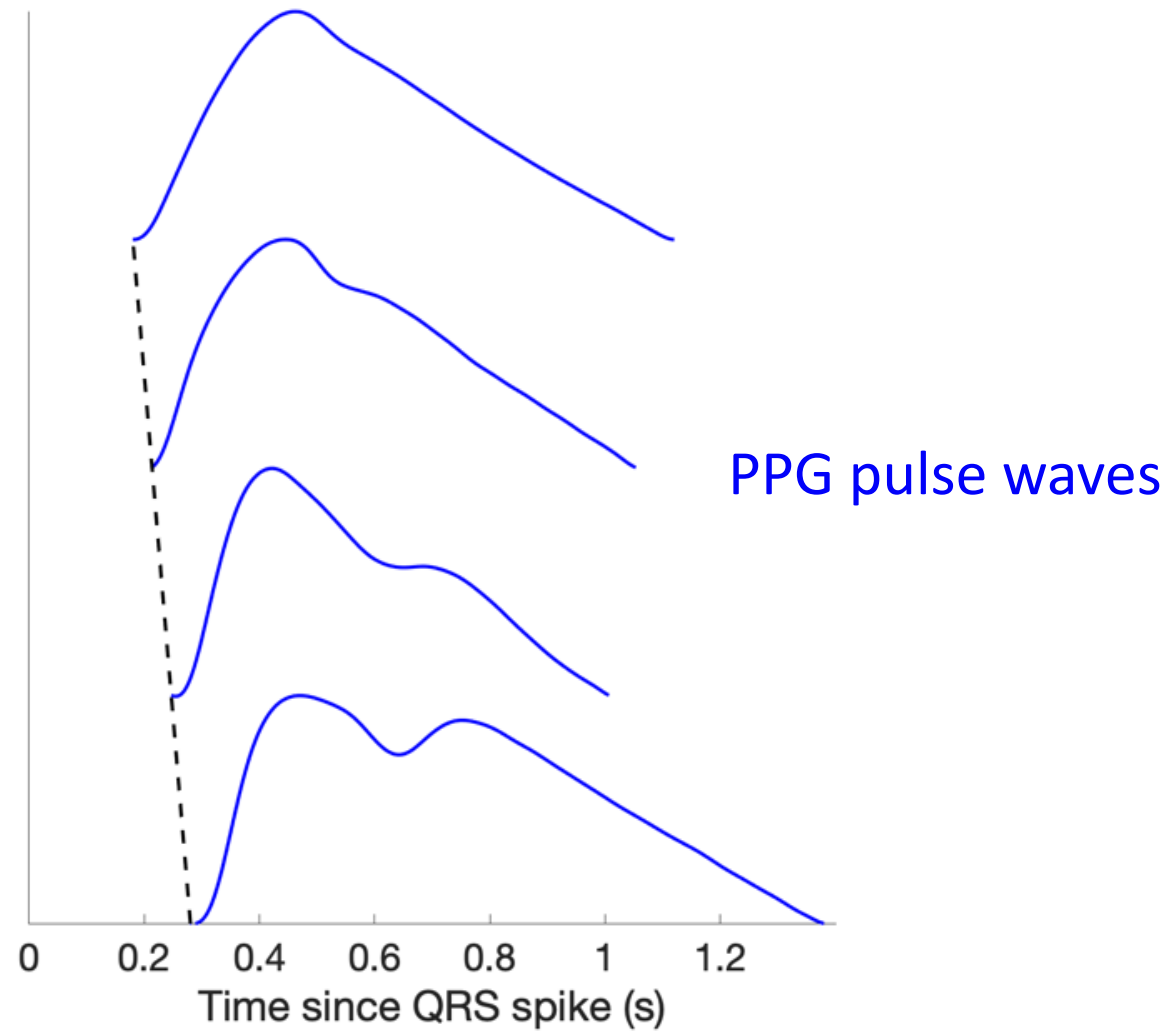
iii) Estimate BP

e.g. estimate systolic blood pressure,

$$SBP = \alpha \cdot \Delta T + \beta$$

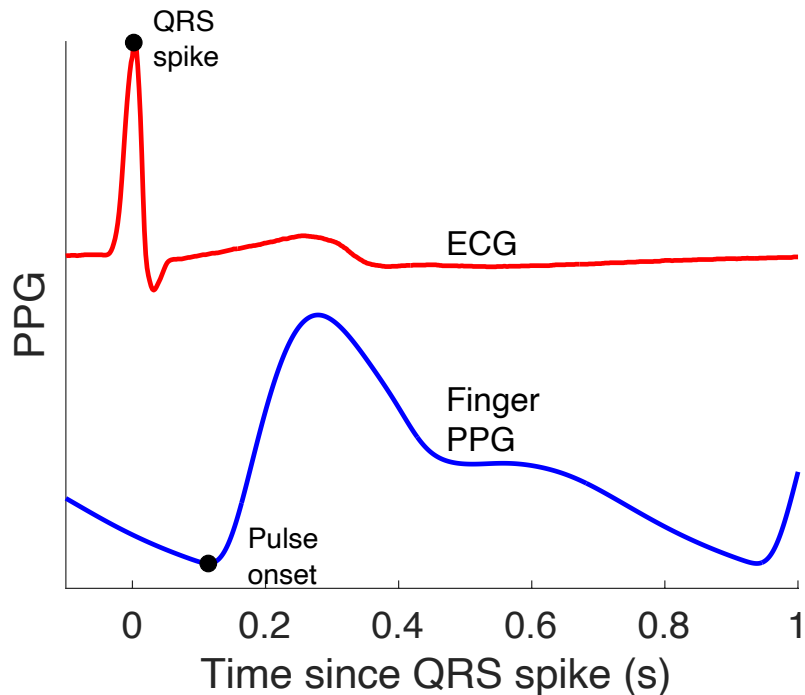
from the time delay ( $\Delta T$ )

Pulse wave timing changes with blood pressure:

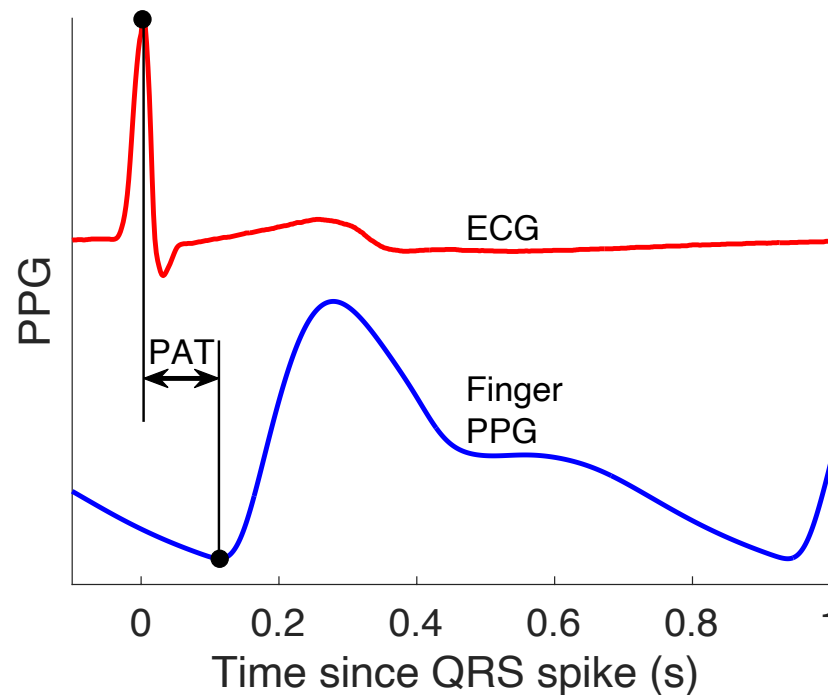


# Estimating blood pressure from PPG and electrocardiogram (ECG) signals

i) Identify QRS and pulse onset



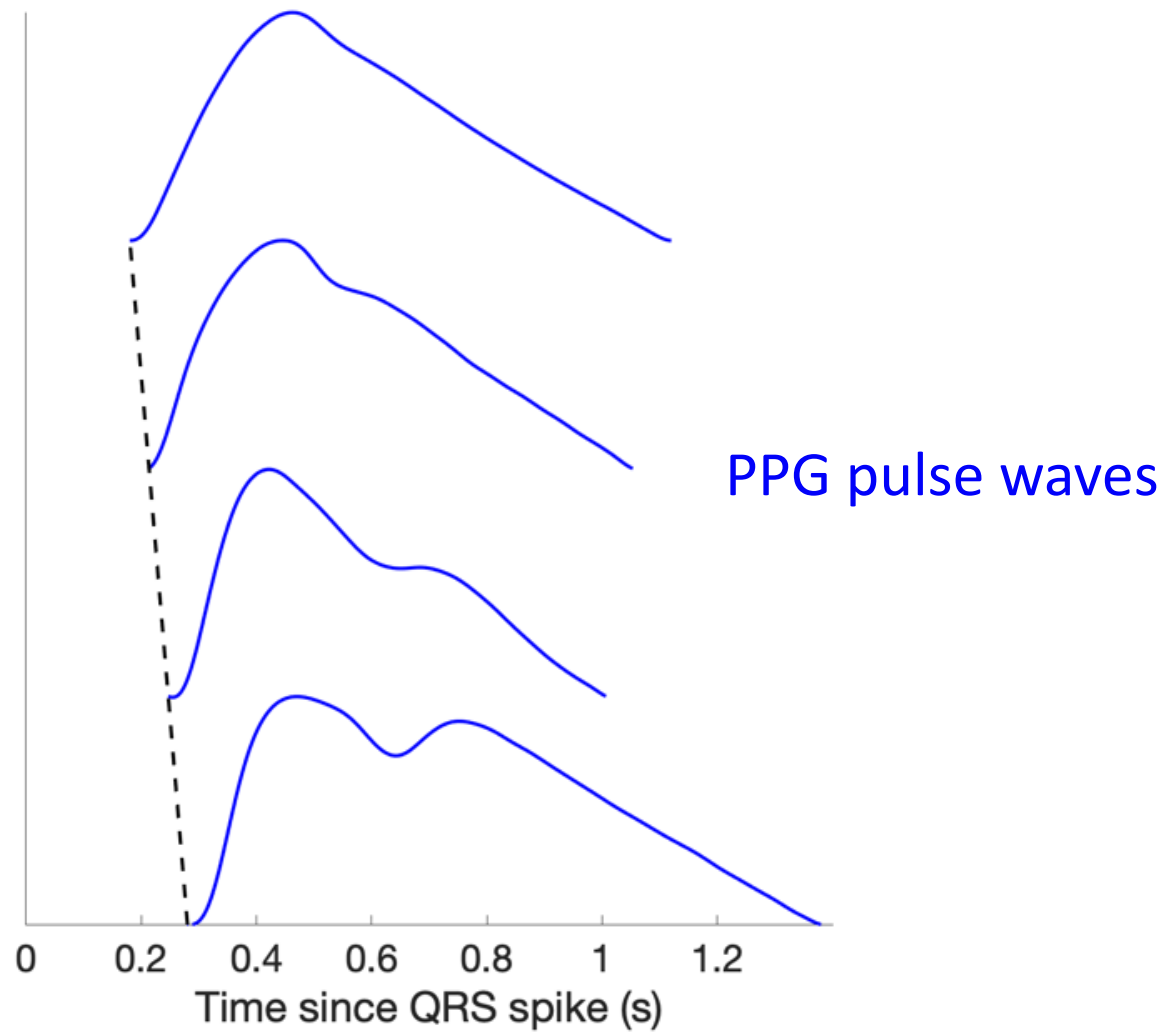
ii) Extract pulse arrival time



iii) Estimate BP

e.g. estimate systolic blood pressure,  
$$SBP = \alpha \cdot \Delta T + \beta$$
from the pulse arrival time (PAT)

Pulse wave shape and timing change with blood pressure:



## Aortic stiffness:

Ageing

~ 10% increase  
per decade

Disease

*e.g.* Diabetes  
Heart failure

Predictor

CV events  
Mortality

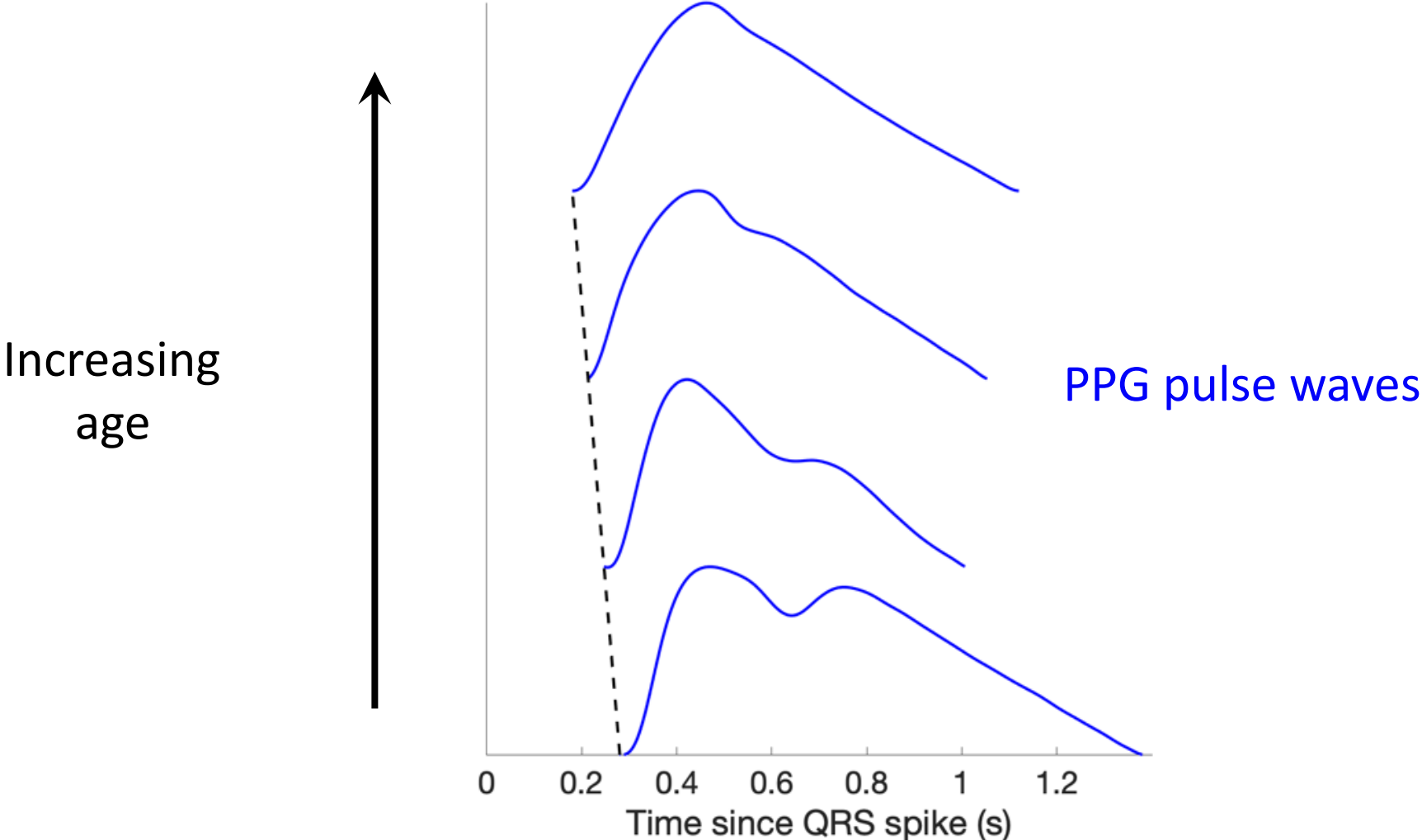
Vascular  
ageing

Benetos A, *Am J Hypertens*, 2002.  
Vlachopoulos C *et al.*, "*JACC*", 2010.

Qimono, *Pixabay* (Pixabay Licence)



# Changes in pulse waves with vascular ageing:



From a recent review:

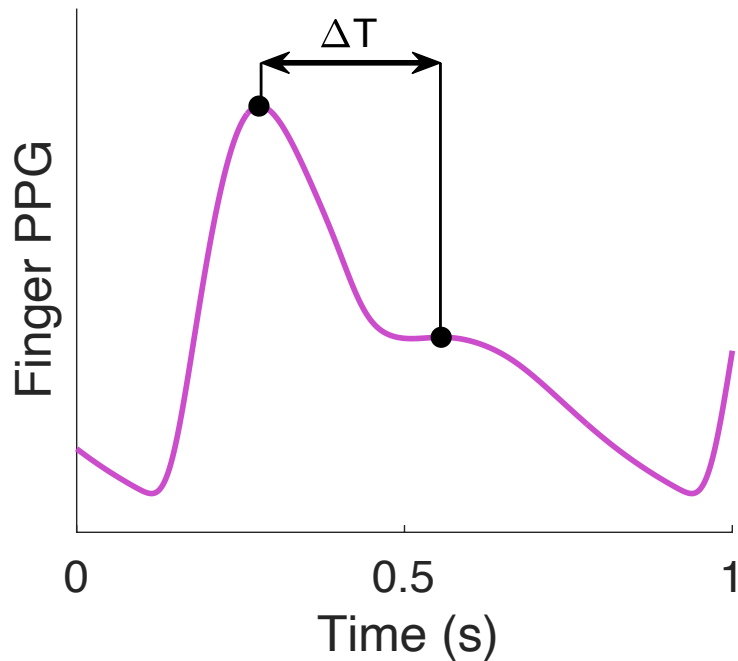
- The indicators of vascular ageing most commonly assessed from the PPG in research:

- Blood pressure (vast majority of articles)
- Arterial stiffness (few)
- Atherosclerosis (inc. peripheral arterial disease) (very few)

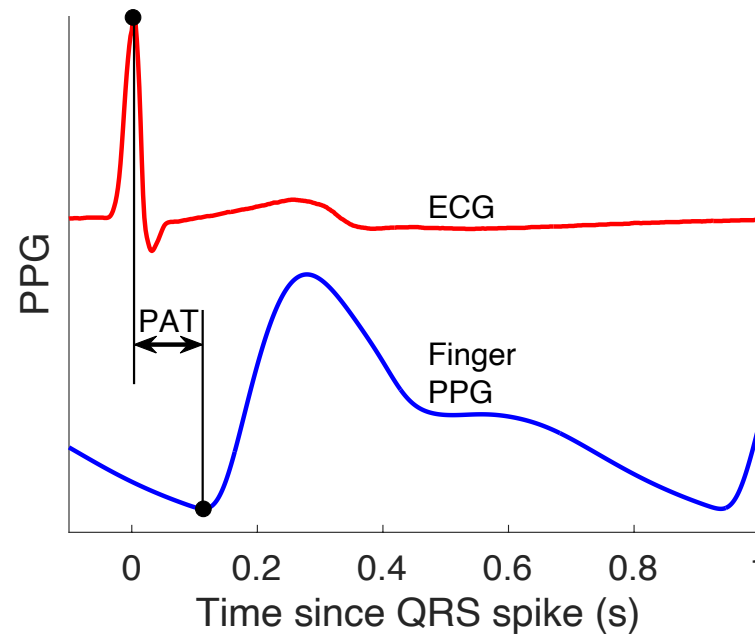
From a recent review:

- Three approaches to using the PPG to assess vascular ageing:

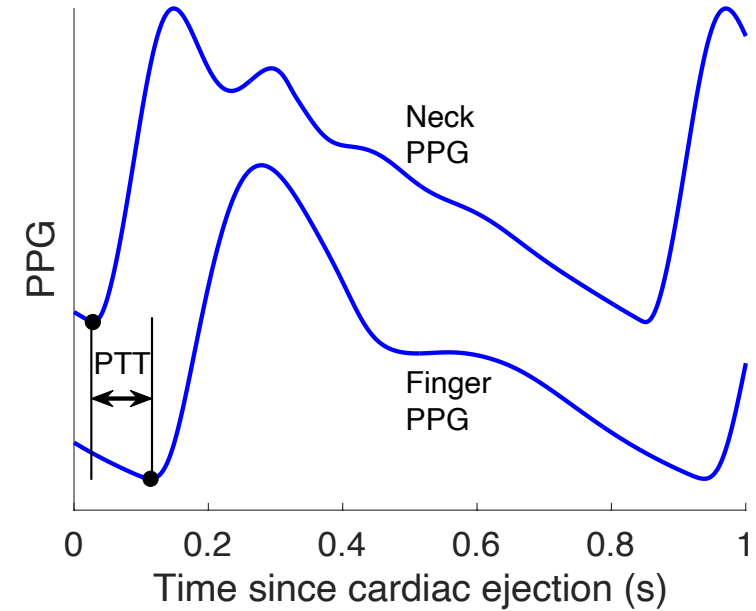
### 1. Single PPG



### 2. PPG and ECG

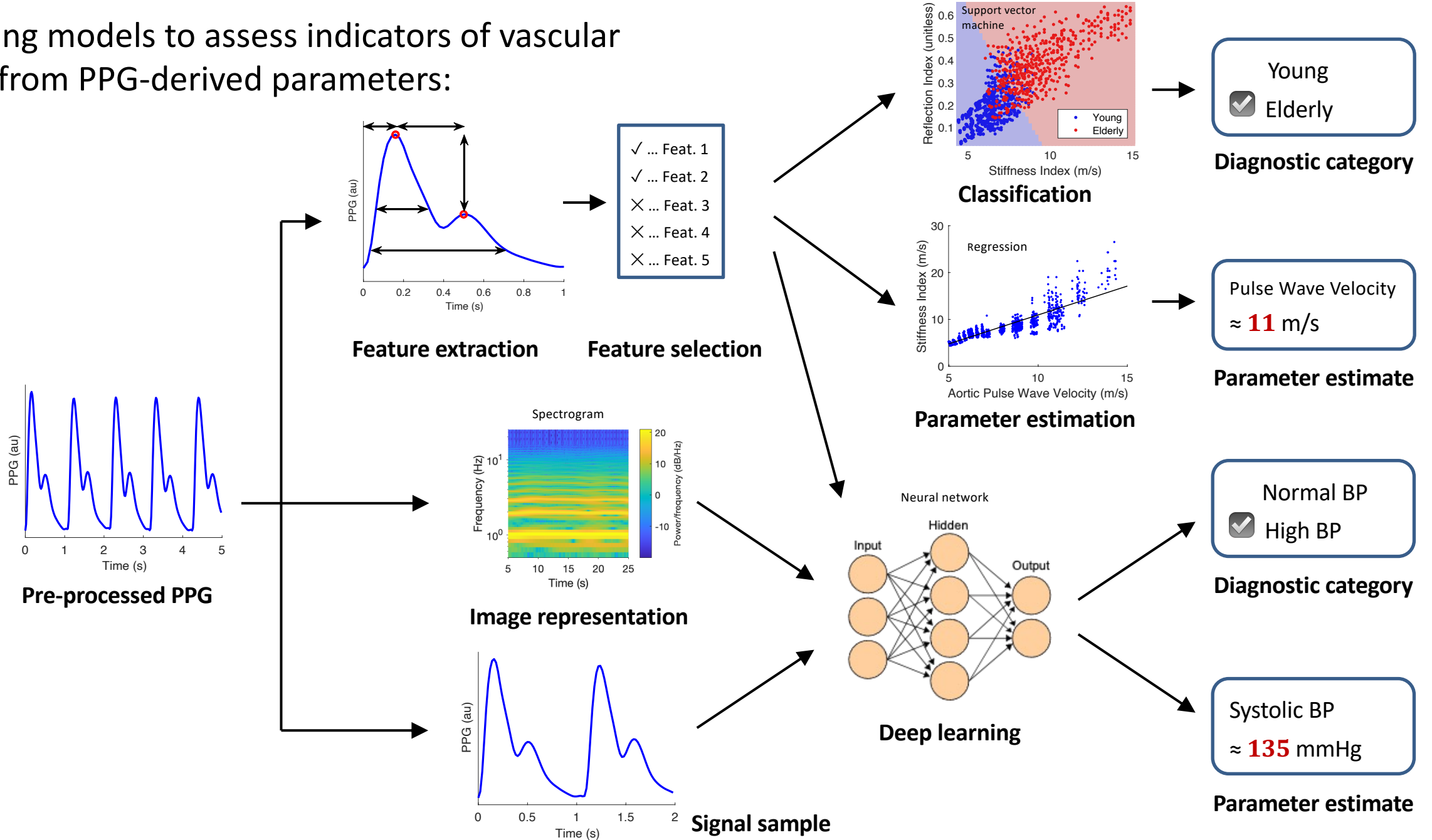


### 3. Multiple PPGs



*How do we estimate a parameter such as blood pressure from these measurements?*

- Using models to assess indicators of vascular age from PPG-derived parameters:



From a recent review:

- How has the performance of PPG-derived parameters of vascular age been assessed?

Studies:

- Mostly in healthy adults
- Small sample sizes
- Range of reference indicators of vascular age
- Range of statistics

From a recent review:

- How well do PPG-derived parameters of vascular age perform in comparison to reference indicators?

Studies:

- Pulse wave velocity: Moderate correlations between PPG-derived and reference pulse wave velocity (or pulse transit time)
- Blood pressure: Very few studies meet required precision
- Peripheral arterial disease: PPG-derived parameters useful

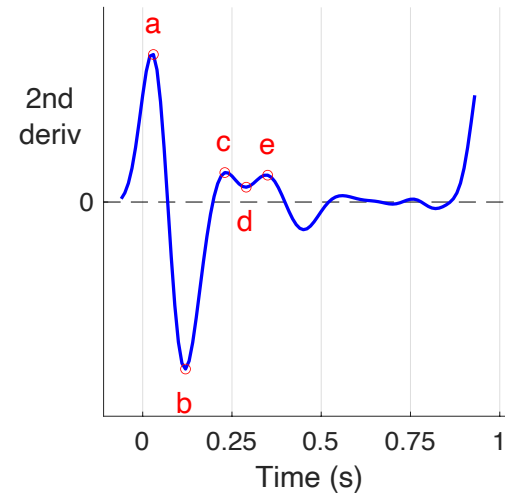
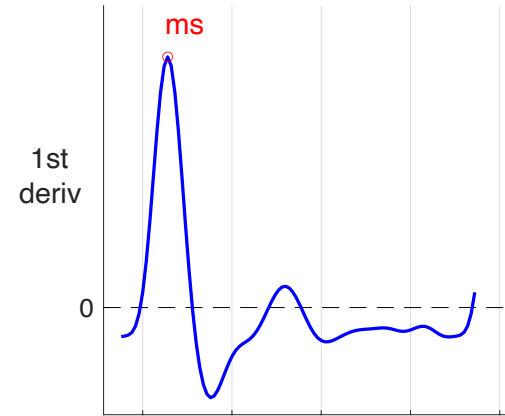
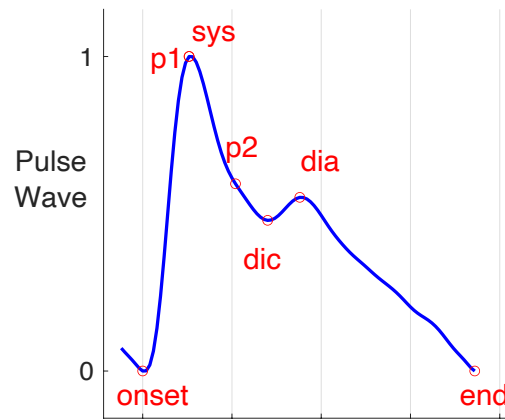
From a recent review:

- What is the potential clinical utility of PPG-derived parameters of vascular age?

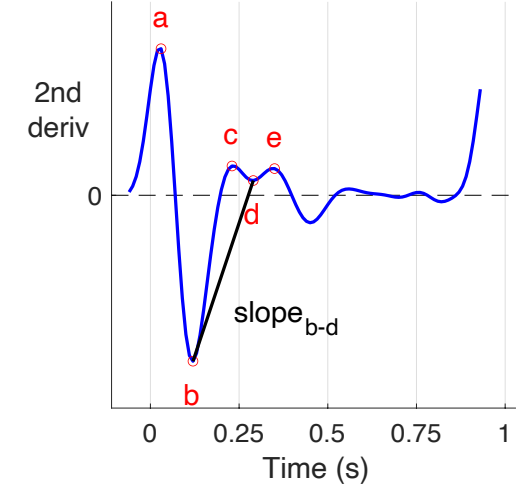
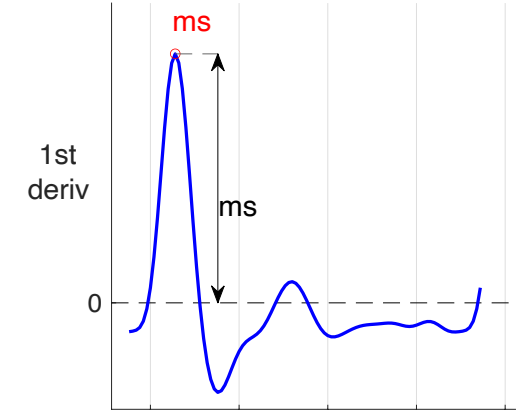
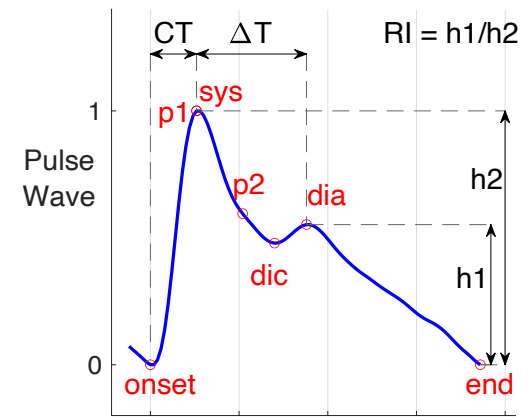
Much of the evidence relates to:

- Identifying peripheral arterial disease
- Identifying diabetes
- Cardiovascular risk prediction

Identify fiducial points:



Extract pulse wave features:



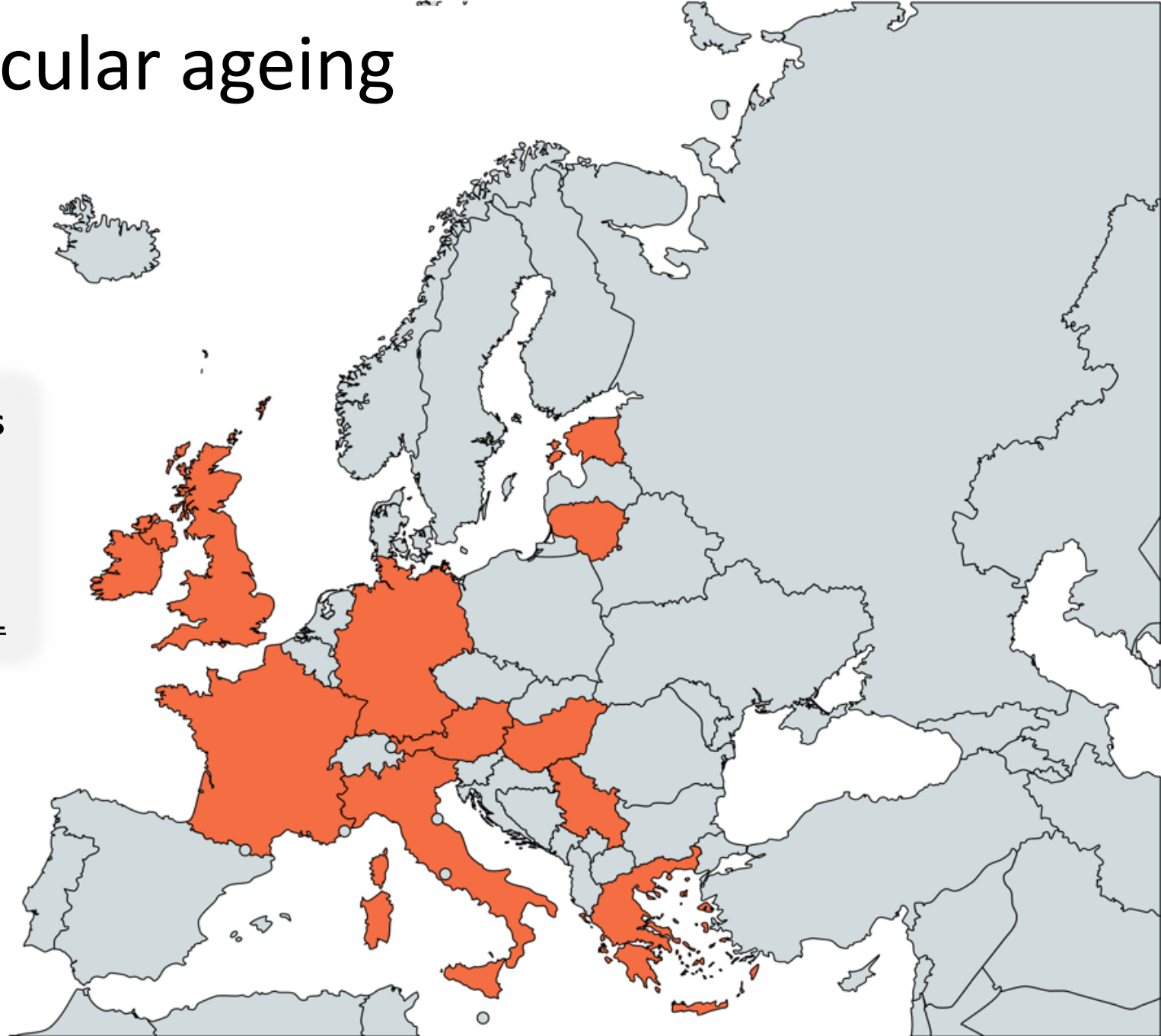
d/a found to be an independent predictor of cardiovascular mortality



# Blood pressure and vascular ageing

For further information see:

Charlton P.H. *et al.*, **Assessing hemodynamics from the photoplethysmogram to gain insights into vascular age: a review from VascAgeNet**, *AJP: Heart Circ.*, 2022.  
<https://doi.org/10.1152/ajpheart.00392.2021>



An anatomical illustration of the human respiratory system. The lungs are shown in a semi-transparent, light blue color, revealing the internal bronchial tree. The bronchial tree is highlighted in a vibrant red, showing the branching structure from the trachea down to the bronchioles. The background is a dark blue gradient with faint, glowing lines suggesting a digital or medical theme.

**respiratory  
monitoring**

# 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

24

20

12

9

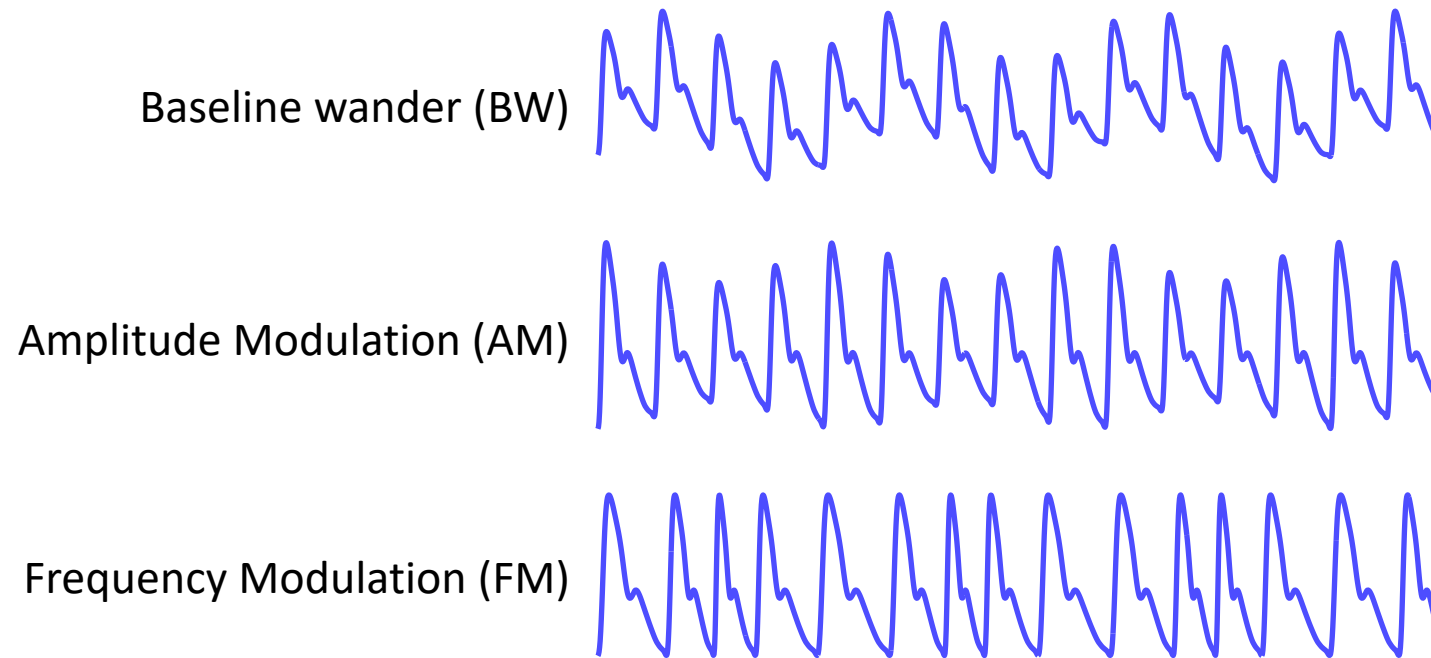
Abnormal

Normal

Abnormal

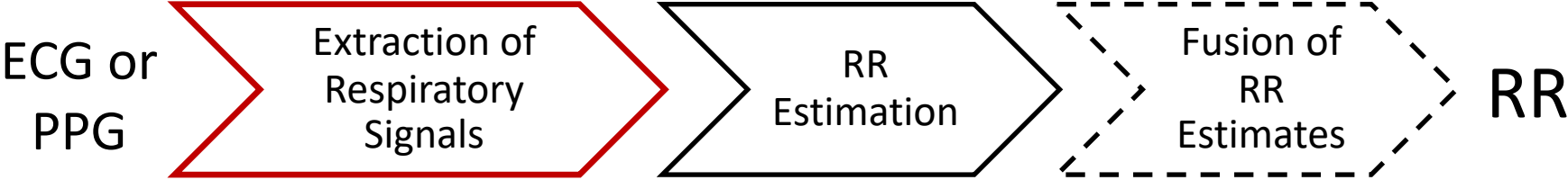
# Estimating respiratory rate

Influence of breathing:

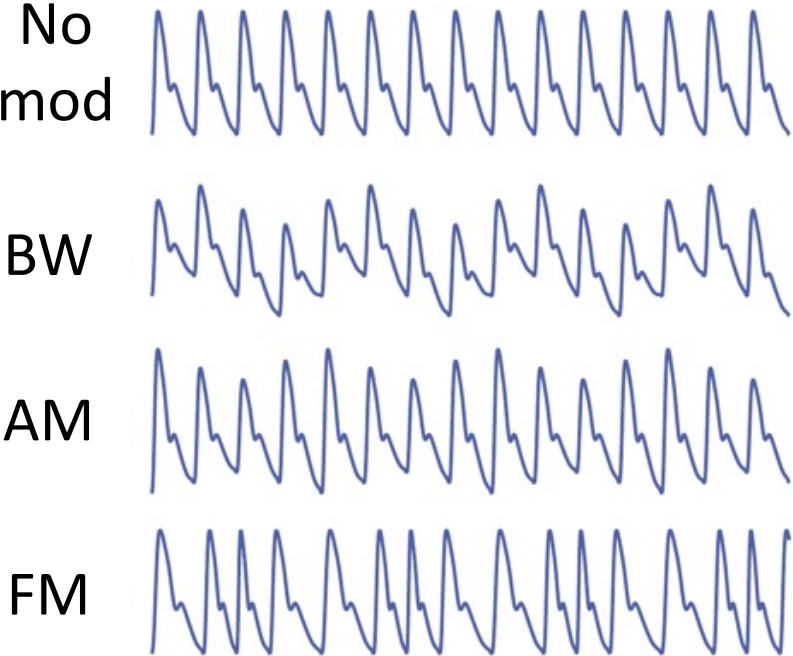


# Estimating respiratory rate

Respiratory rate is estimated by analysing the respiratory modulations in the PPG:

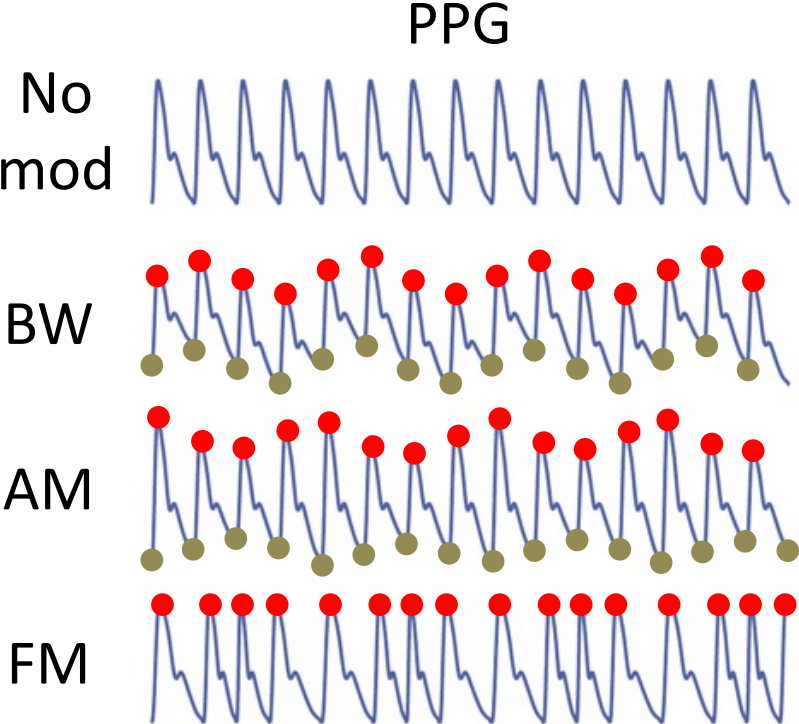


PPG



# Estimating respiratory rate

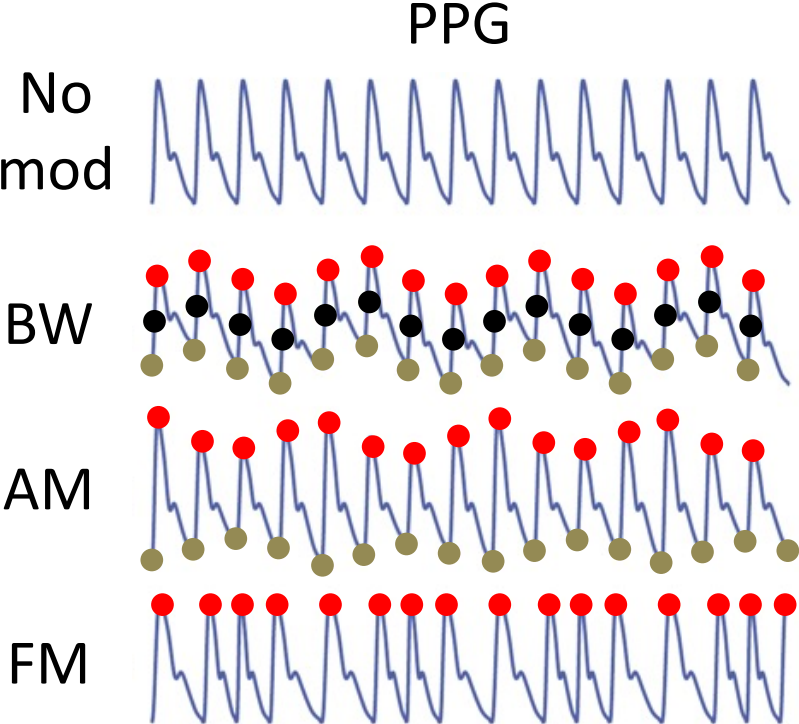
Respiratory rate is estimated by analysing the respiratory modulations in the PPG:



Identify fiducial points

# Estimating respiratory rate

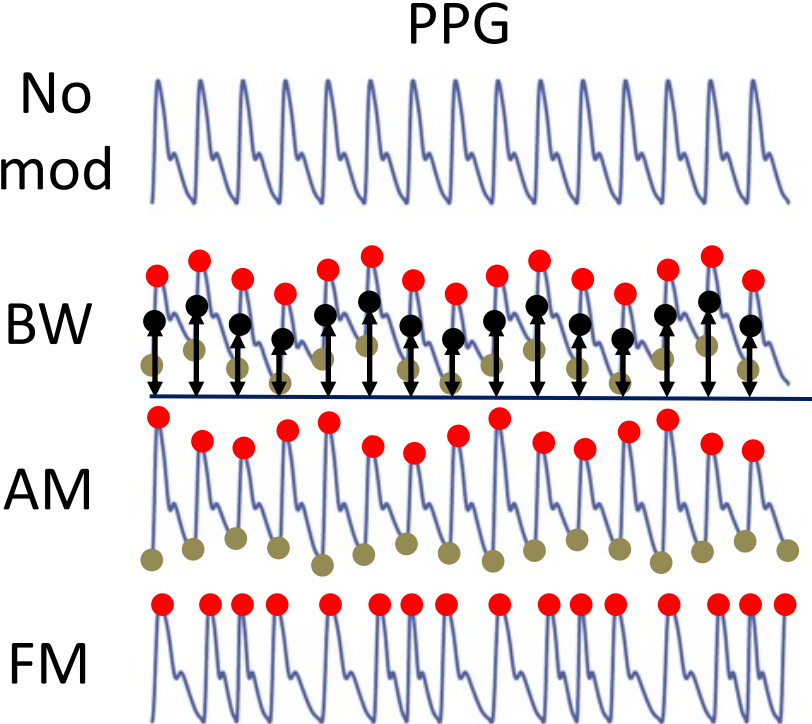
Respiratory rate is estimated by analysing the respiratory modulations in the PPG:



Find baseline

# Estimating respiratory rate

Respiratory rate is estimated by analysing the respiratory modulations in the PPG:

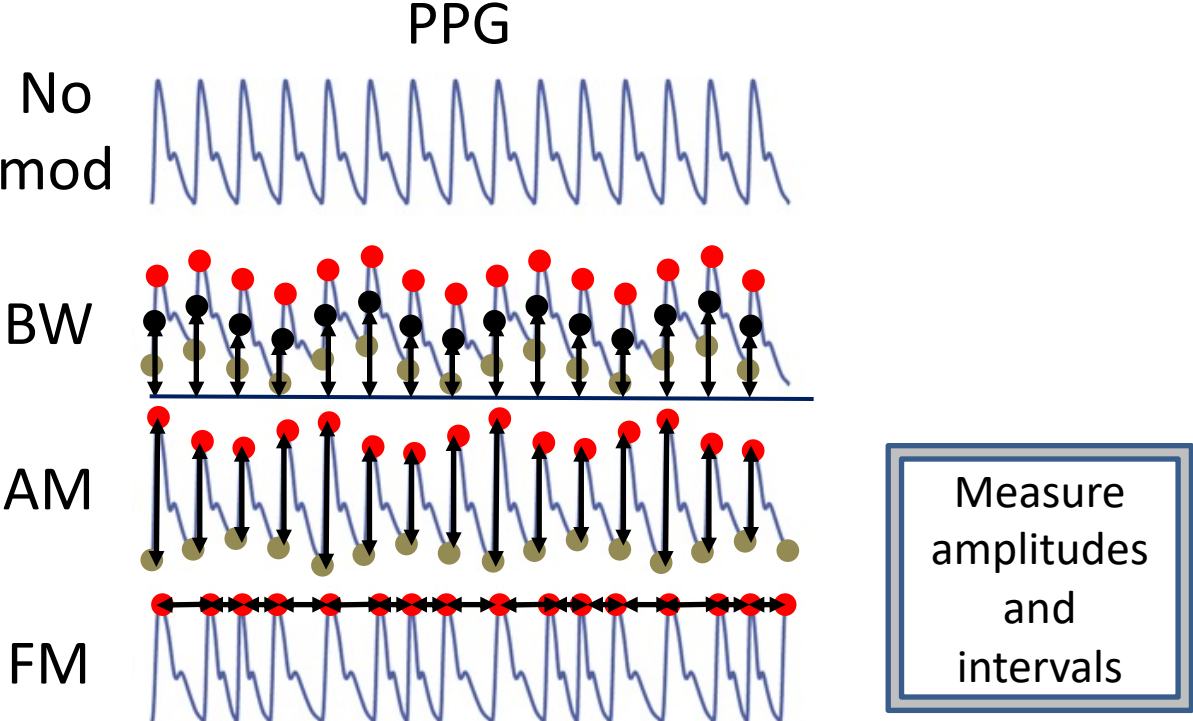


Find baseline



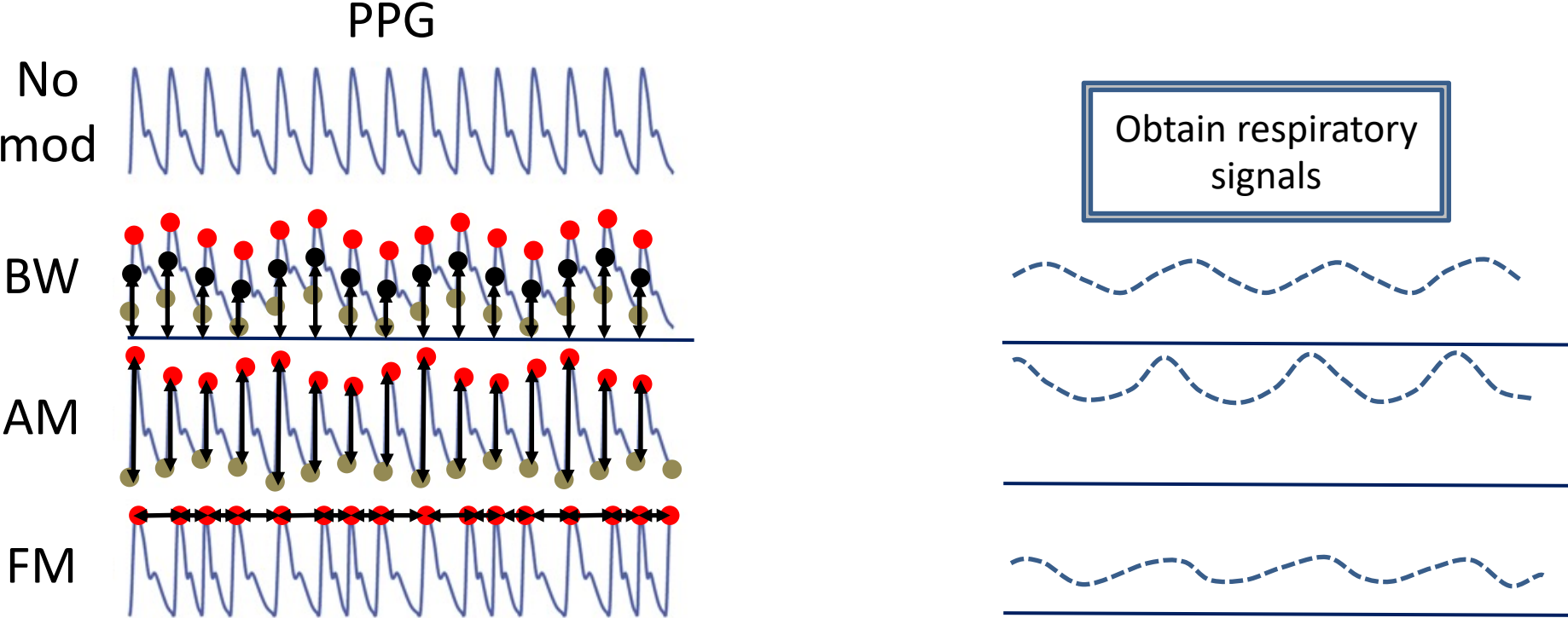
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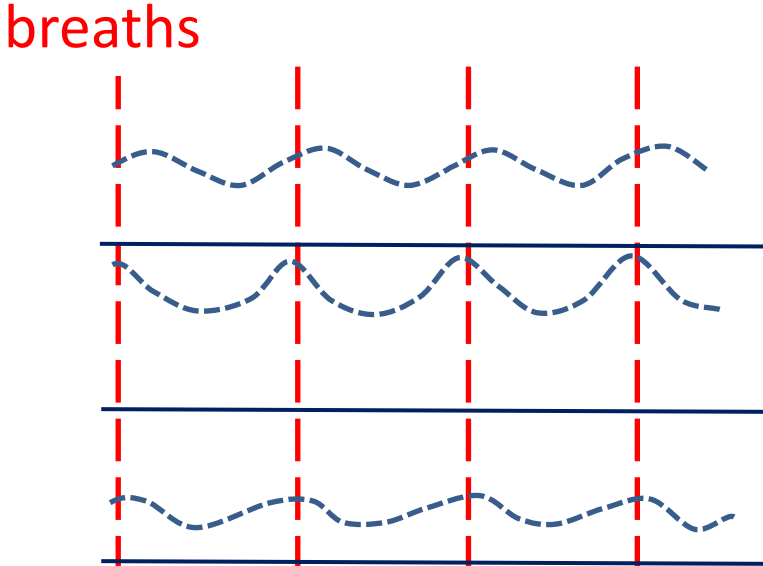
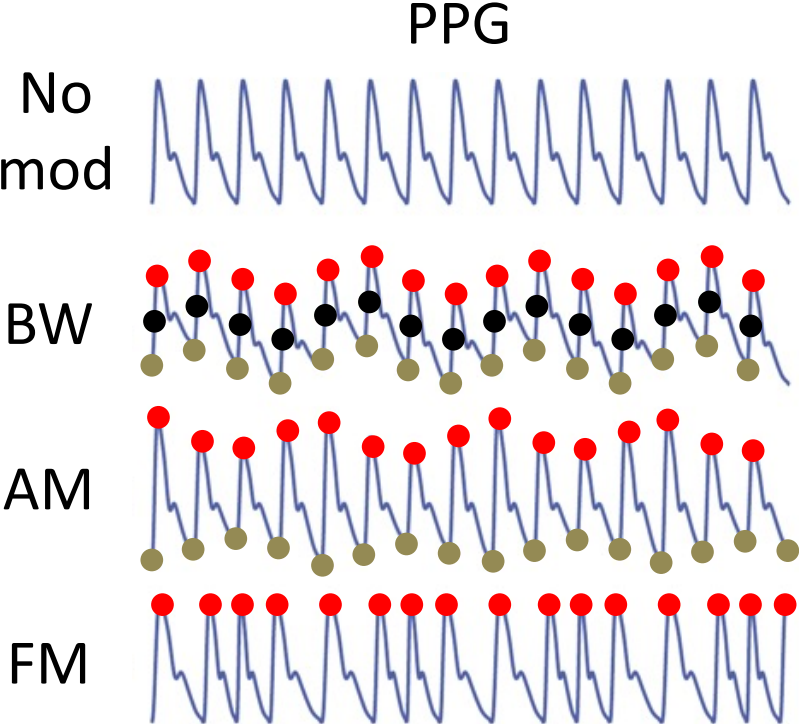
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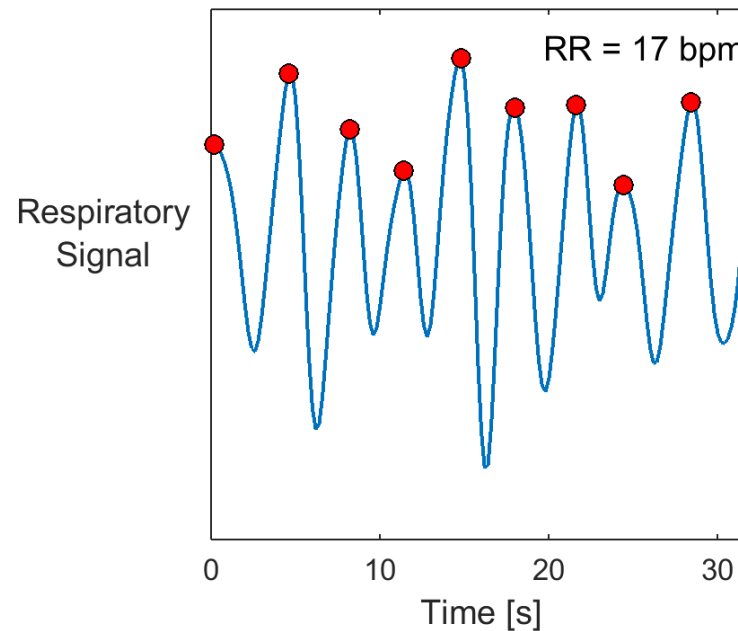
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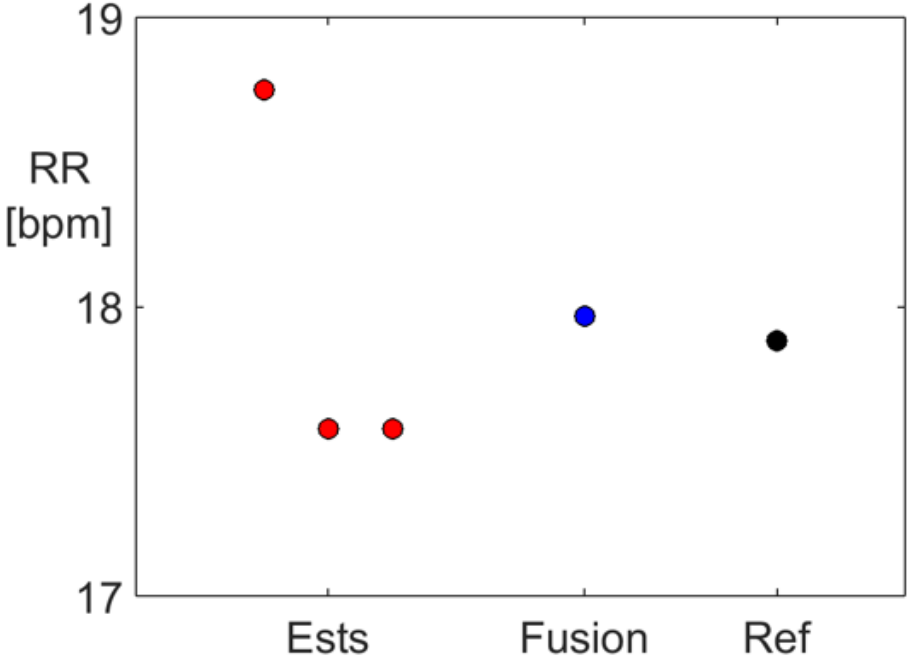
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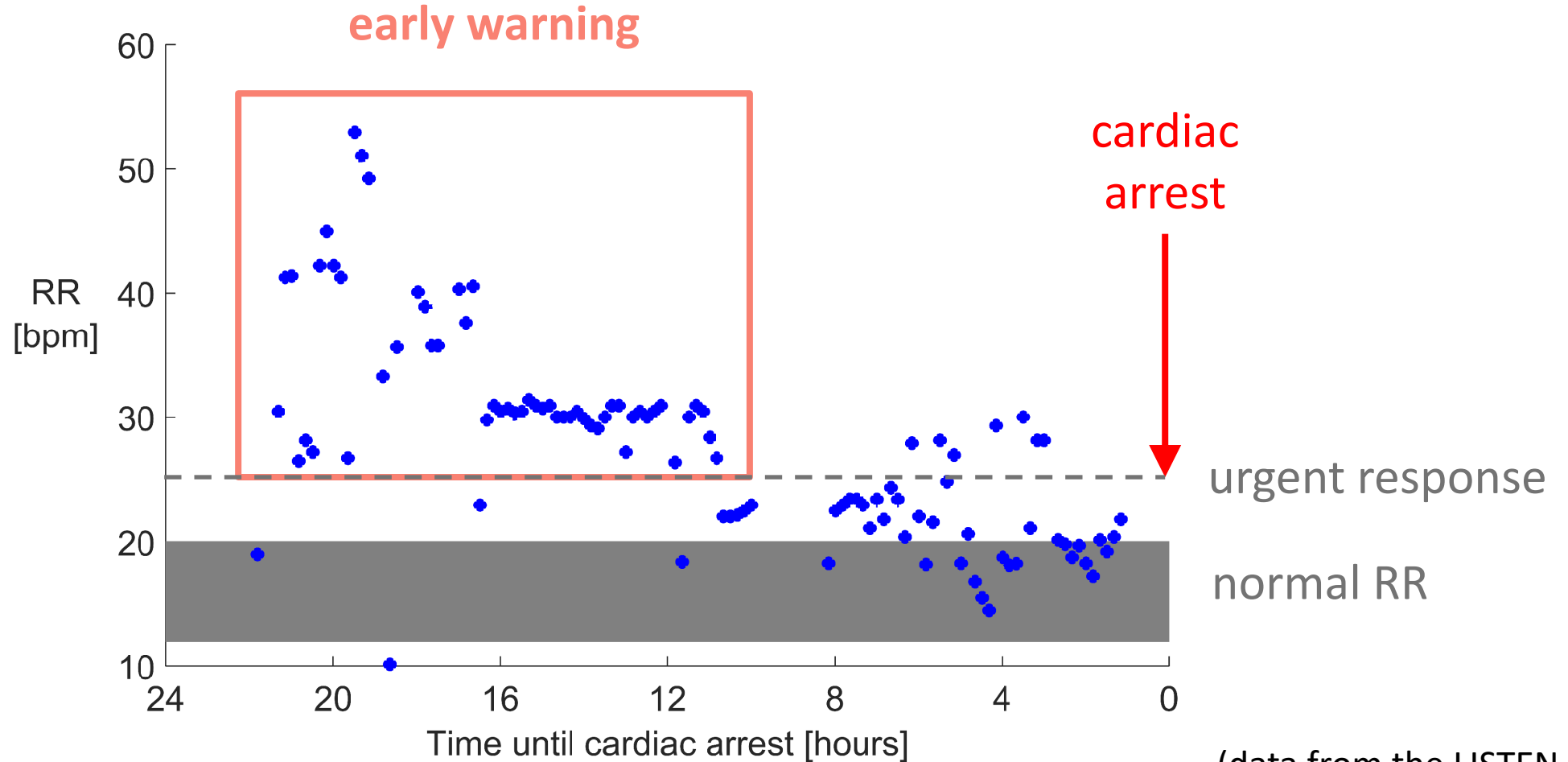
# Estimating respiratory rate

Respiratory rate is estimated by analysing the respiratory modulations in the PPG:



# Assessing clinical utility

ECG-derived RRs every 10 mins on hospital ward



(data from the LISTEN study)

# Identifying Obstructive Sleep Apnea



## Key publications:

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

# Infectious Disease Surveillance



*Fitness trackers of the future ...*

## Influenza-like illness

Resting  
heart rate

Sleep  
duration

## COVID-19

Heart  
rate

Sleep

Activity

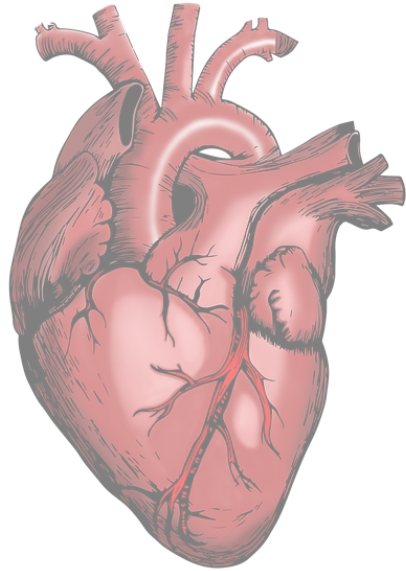
Radin *et al.* [Harnessing wearable device data to improve state-level real-time surveillance of influenza-like illness in the USA: a population-based study](#). *Lancet Digit. Heal.* 2020

Quer *et al.*, [Wearable sensor data and self-reported symptoms for COVID-19 detection](#). *Nat Med*, 2020

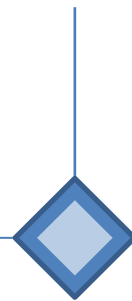
Gadaleta *et al.* [Passive detection of COVID-19 with wearable sensors and explainable machine learning algorithms](#). *npj Digit Med*, 2021



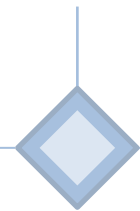
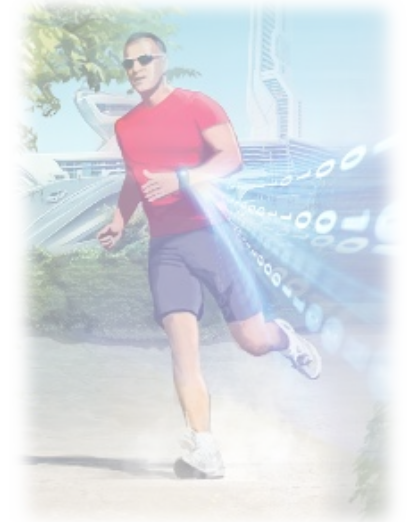
## Applications



## Research and Development



## Challenges

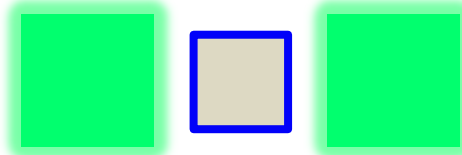


# Optimising sensor design

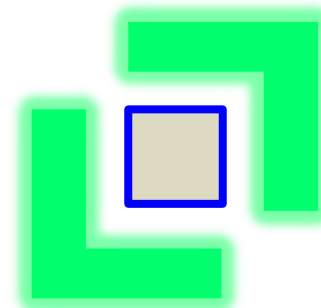
*What is the optimal sensor design for obtaining photoplethysmography signals?*



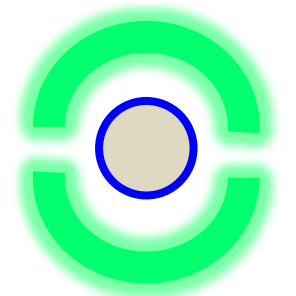
- Sensor geometry



Rectangular



Bracket



Circular

*Affects the amplitude of the signal:*

*See Khan et al. (2019): DOI: [10.1109/ACCESS.2019.2939798](https://doi.org/10.1109/ACCESS.2019.2939798) (CC BY 4.0)*

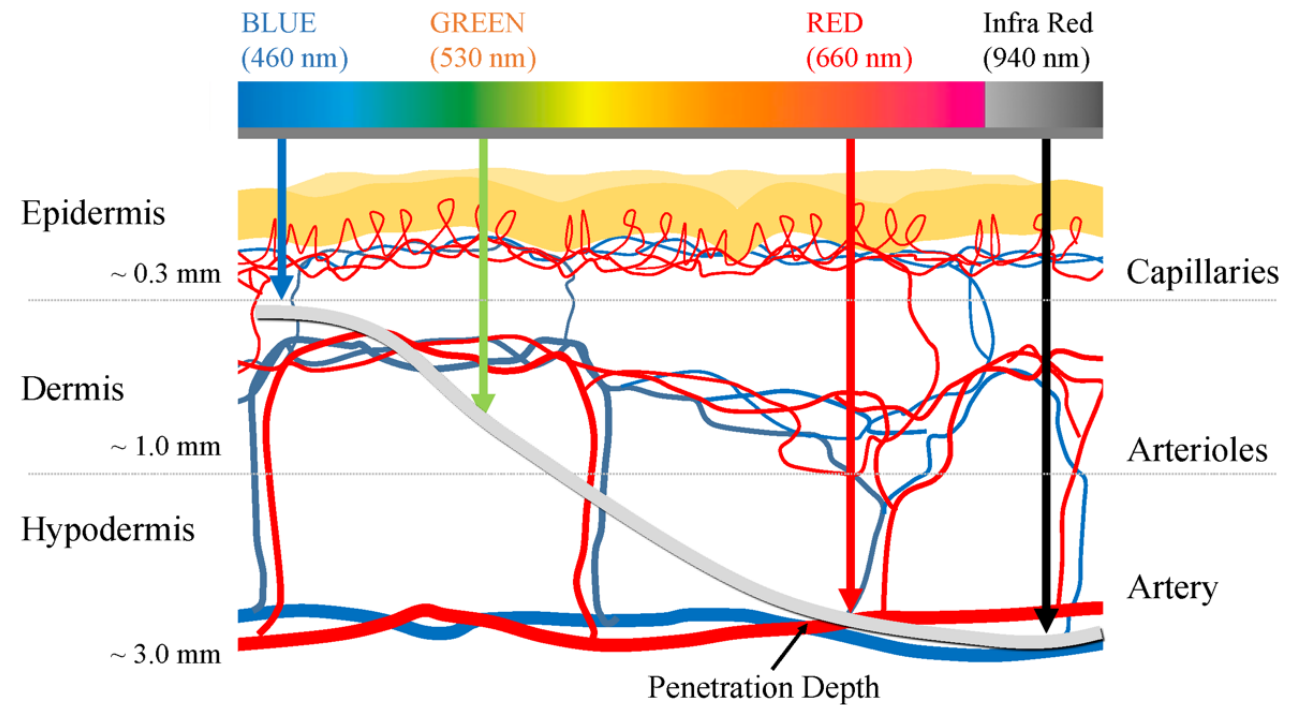
*(from which the images were adapted)*

# Optimising sensor design

*What is the optimal sensor design for obtaining photoplethysmography signals?*



- Wavelength of light



*Affects the penetration depth:*

*See Han et al. (2019): DOI: [10.3390/s19245441](https://doi.org/10.3390/s19245441) (CC BY 4.0), from which the image was sourced*

# Optimising sensor design

*What is the optimal sensor design for obtaining photoplethysmography signals?*



- Sensor attachment:
  - contact pressure
  - vulnerability to motion artifact
- Simultaneous signals:
  - Accelerometry for motion detection and cancellation
  - Electrocardiography for pulse arrival time
- Alternative pulse wave sensing technologies

# Algorithm development

*What are the optimal algorithms for estimating physiological parameters?*



- Understanding prior work - many algorithms proposed:
  - >100 for estimating respiratory rate from PPG or ECG
  - >39 different pulse wave features used to assess vascular ageing from the PPG
- Interpreting prior work:
  - Performance differs greatly between algorithms
  - Performance differs on different datasets
- Leaderboards: (e.g. [paperswithcode.com](https://paperswithcode.com))
  - Open-source algorithms
  - Well-labelled public datasets
- Incorporating uncertainty assessment

# Integration into clinical pathways

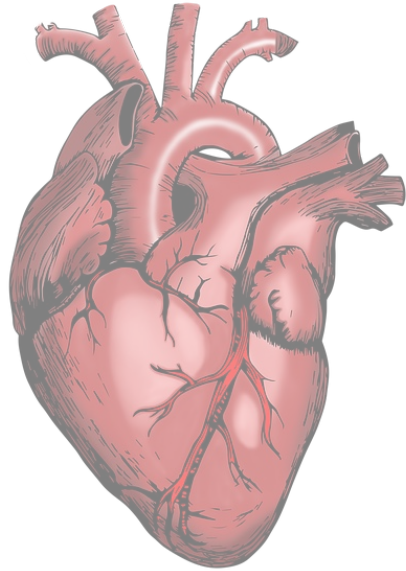
*How should wearable data be integrated into clinical decision making?*



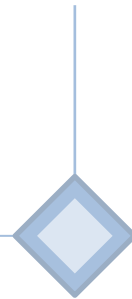
- Screening programmes
- Patient-led measurements to prompt clinical assessment
- Supplementary clinical use in specific settings
- Self-directed health monitoring
- Population-level surveillance

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

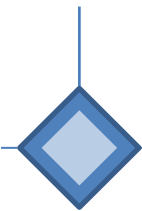
## Applications



## Research and Development

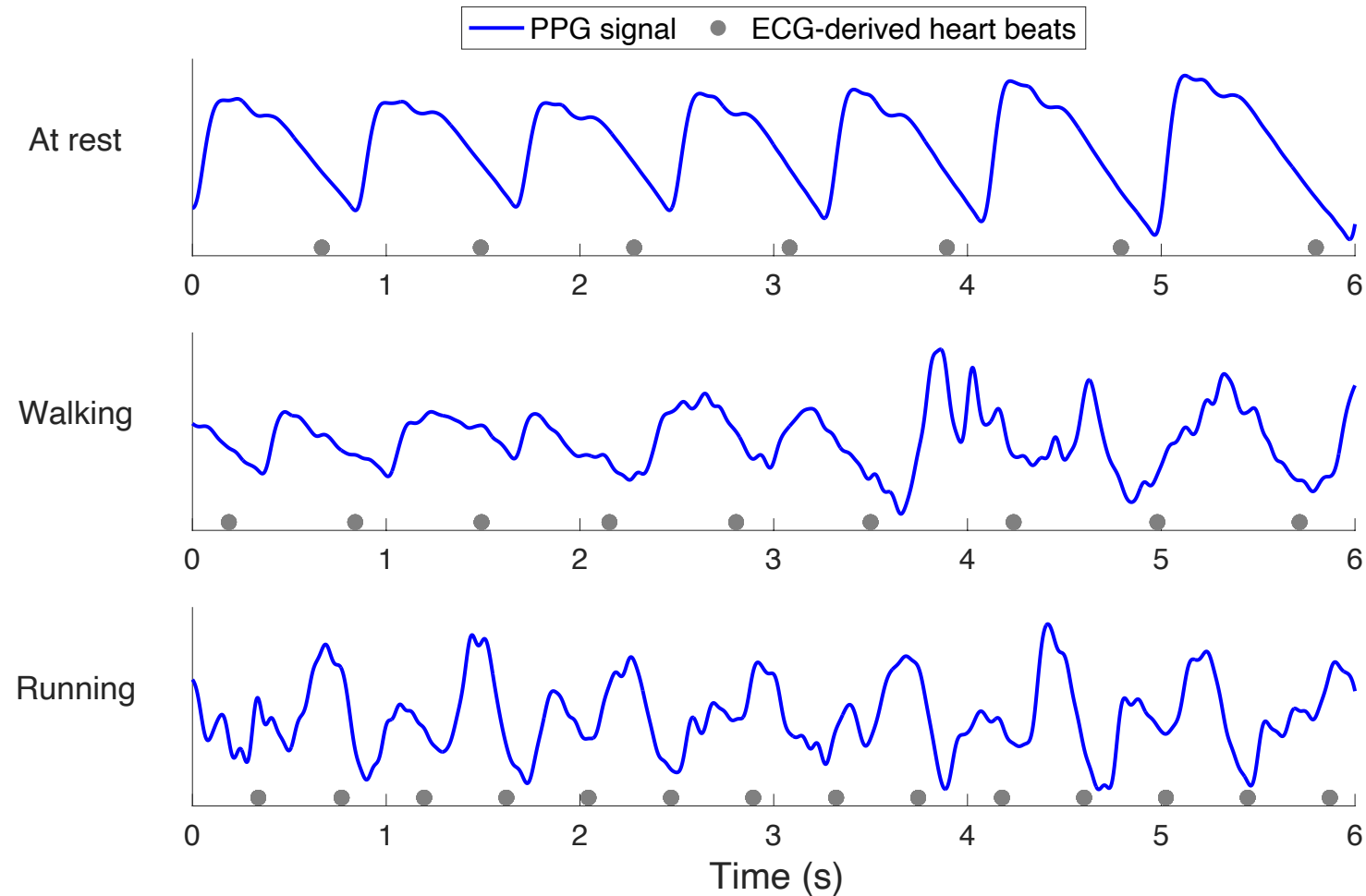


## Challenges



# Optimising sensor design

Motion artifact:



Potential strategies:

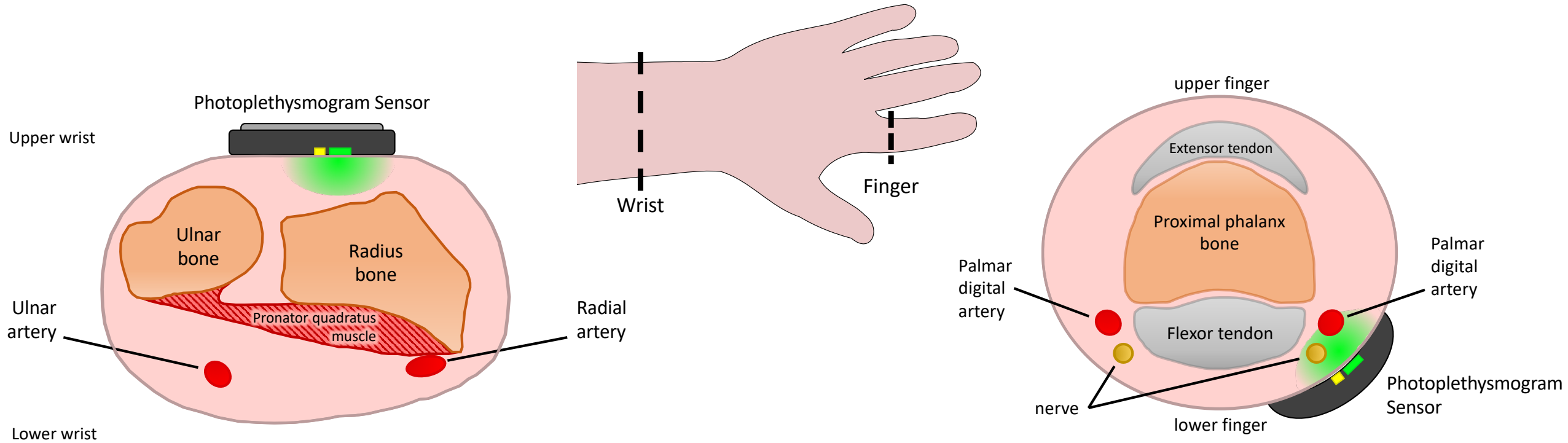
- Exclude artifactual data
- Motion artifact cancellation



# Optimising sensor design

Unobtrusiveness vs. utility:

- Measurement site



# Optimising sensor design

Unobtrusiveness vs. utility:

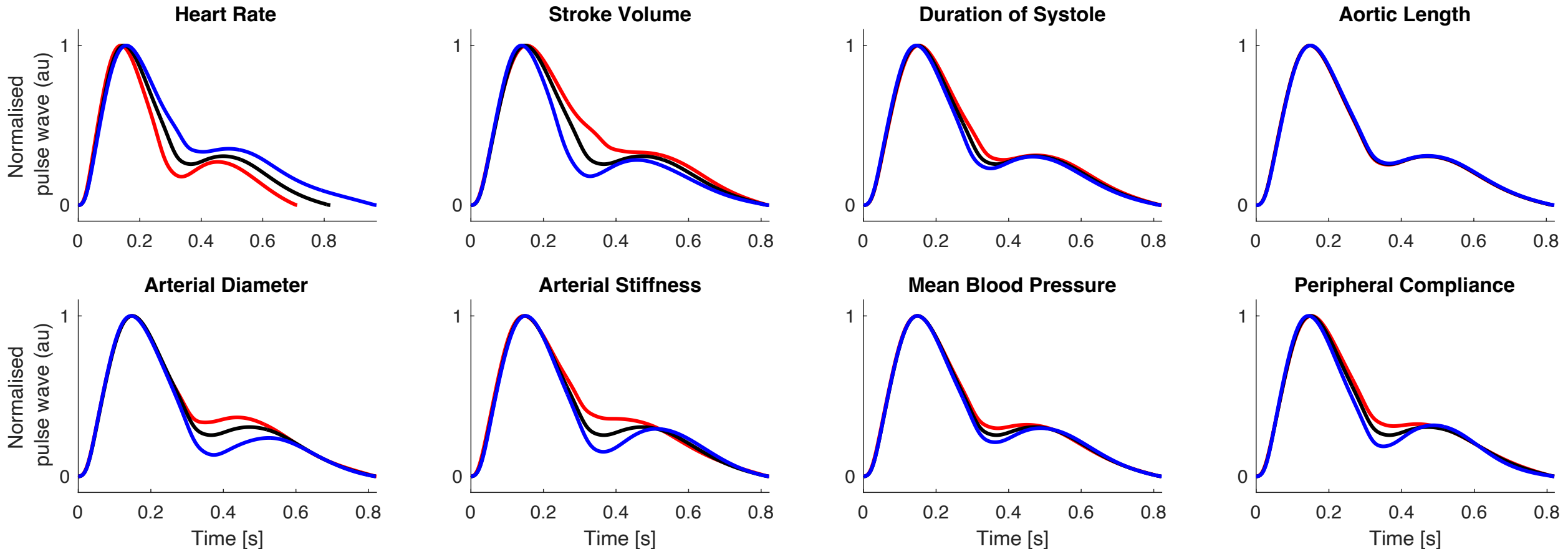
- Measurement site
- Pulse arrival time



# Algorithm development

*What are the physiological origins of the PPG signal?*

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



# Algorithm development

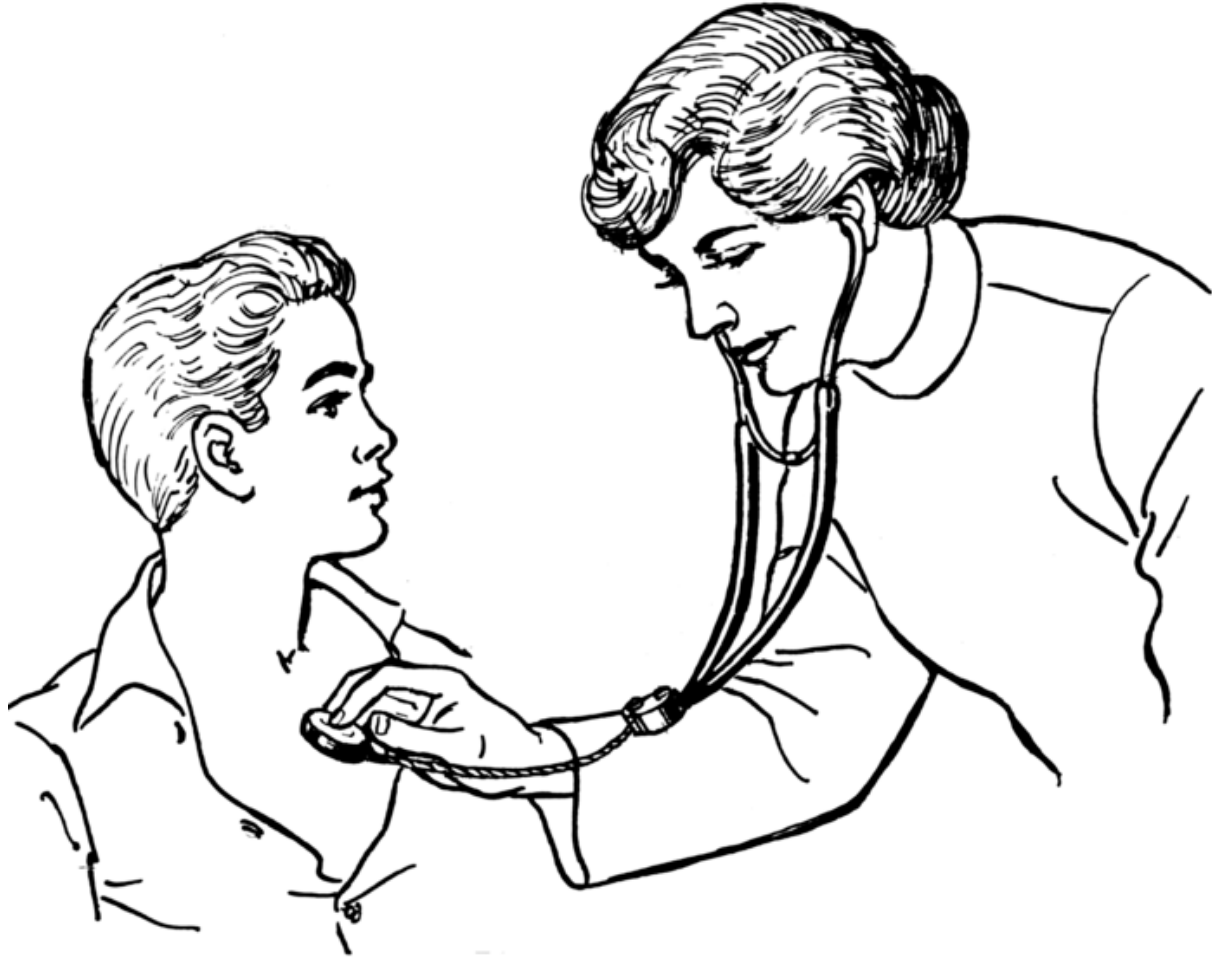
*How should algorithms be developed, trained, and validated?*

- Simulated data and controlled experiments can be used to develop initial algorithms
- Further development and training should be performed with real-world data
- Validation should be performed using robust, agreed protocols (see [Mukkamala 2021, Hypertension](#))
- Performance monitoring could be performed using device data

Transformative initiatives:


- Access to raw sensor data (PPG signals)
- Linking device data with participant outcomes via opt-in studies

# Integration into clinical pathways



With grateful thanks to:

- The British Heart Foundation
- The many colleagues and patients who have contributed to our research.



Wearable photoplethysmography devices have many potential applications, including cardiovascular and respiratory monitoring.

Key steps required to make these a reality include optimising sensor design, algorithm development, and integration into clinical pathways.

Despite the challenges, when used clinically, photoplethysmography devices should be like a climbing rope:

- highly reliable
- validated
- for specific purposes

Further reading:

Charlton P.H. *et al.*, **Wearable photoplethysmography for cardiovascular monitoring**, *Proceedings of the IEEE*, 2022, <https://doi.org/10.1109/JPROC.2022.3149785> (CC BY 4.0)

Charlton PH and Marozas V, **Wearable photoplethysmography devices**, in *Photoplethysmography*, Elsevier, 2022. [http://peterhcharlton.github.io/publication/wearable\\_ppg\\_chapter/](http://peterhcharlton.github.io/publication/wearable_ppg_chapter/)

Charlton P.H. *et al.*, **Breathing rate estimation from the electrocardiogram and photoplethysmogram: a review**, *IEEE RBME*, 2018, <https://doi.org/10.1109/RBME.2017.2763681> (CC BY 3.0)

Charlton P.H. *et al.*, **Establishing best practices in photoplethysmography signal acquisition and processing**, *Physiological Measurement*, 2022, <https://doi.org/10.1088/1361-6579/ac6cc4> (CC BY 4.0)

Charlton P.H. *et al.*, **Assessing hemodynamics from the photoplethysmogram to gain insights into vascular age: a review from VascAgeNet**, *AJP Heart Circ*, 2022, <https://doi.org/10.1152/ajpheart.00392.2021> (CC BY 4.0)

# Wearable photoplethysmography devices for health monitoring

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