

Energy Efficient Heart Rate Sensing using a Painted Electrode ECG Wearable

Christopher Beach

Authors: Sammy Krachunov, Christopher Beach, Alexander J. Casson, James Pope, Xenofon Fafoutis, Robert J. Piechocki and Ian Craddock

christopher.beach@manchester.ac.uk

www.eee.manchester.ac.uk/sisp

Objectives

Overview IoT in healthcare initiatives in the UK and how we align with these

Motivate the need for and benefit of wearable sensors for IoT in healthcare

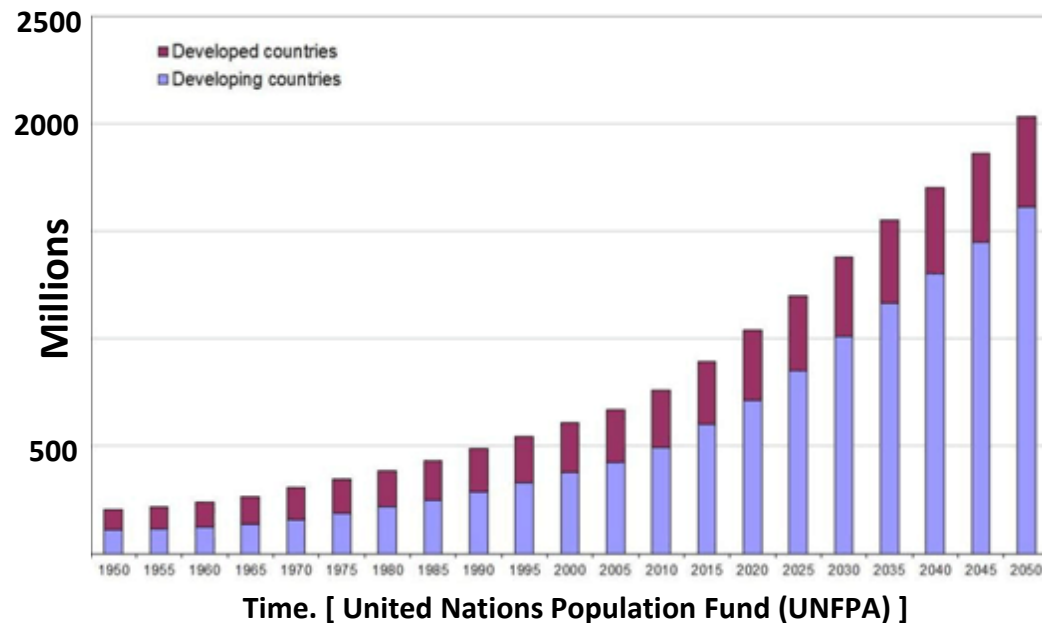
Demonstrate a wearable ECG unit with more than one month of battery life

Overarching motivation

Ageing of the global population is unprecedented and the existing medical system will soon not be able to meet this increasing demand.

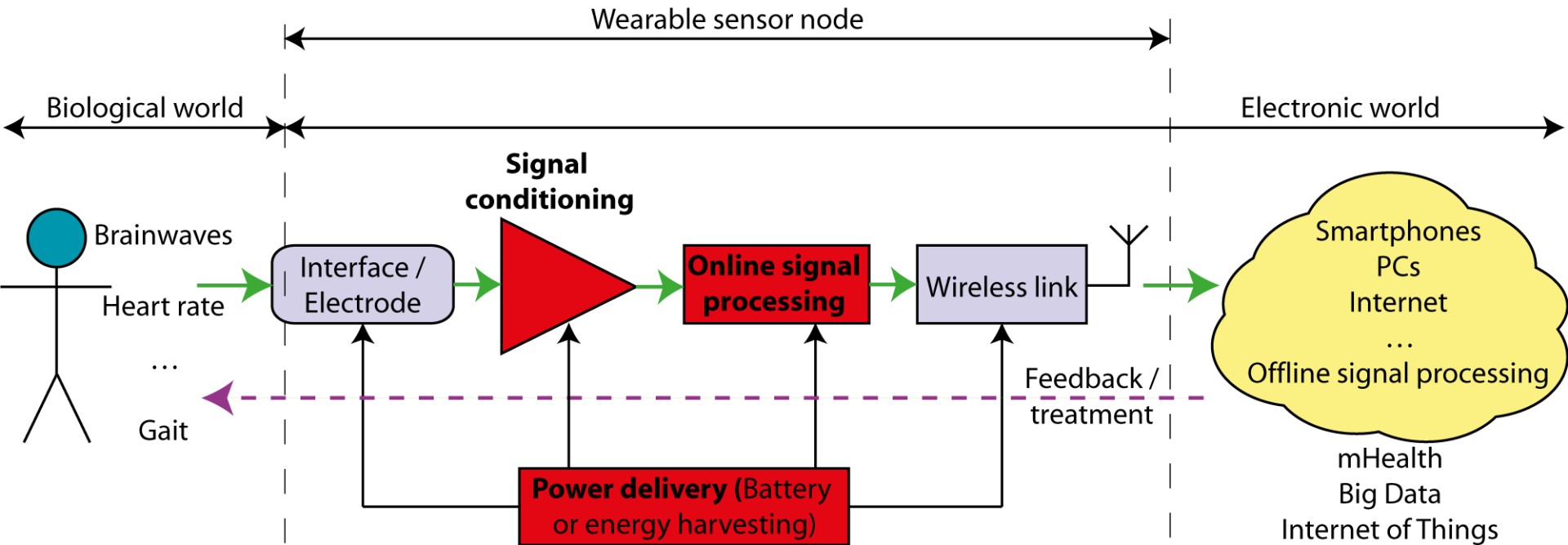
Therefore the use of emerging **digital technologies for healthcare monitoring** is essential.

Our aim is to impact a range of healthcare needs by employing data-fusion from a common platform of largely non-medical networked sensors in a home environment.

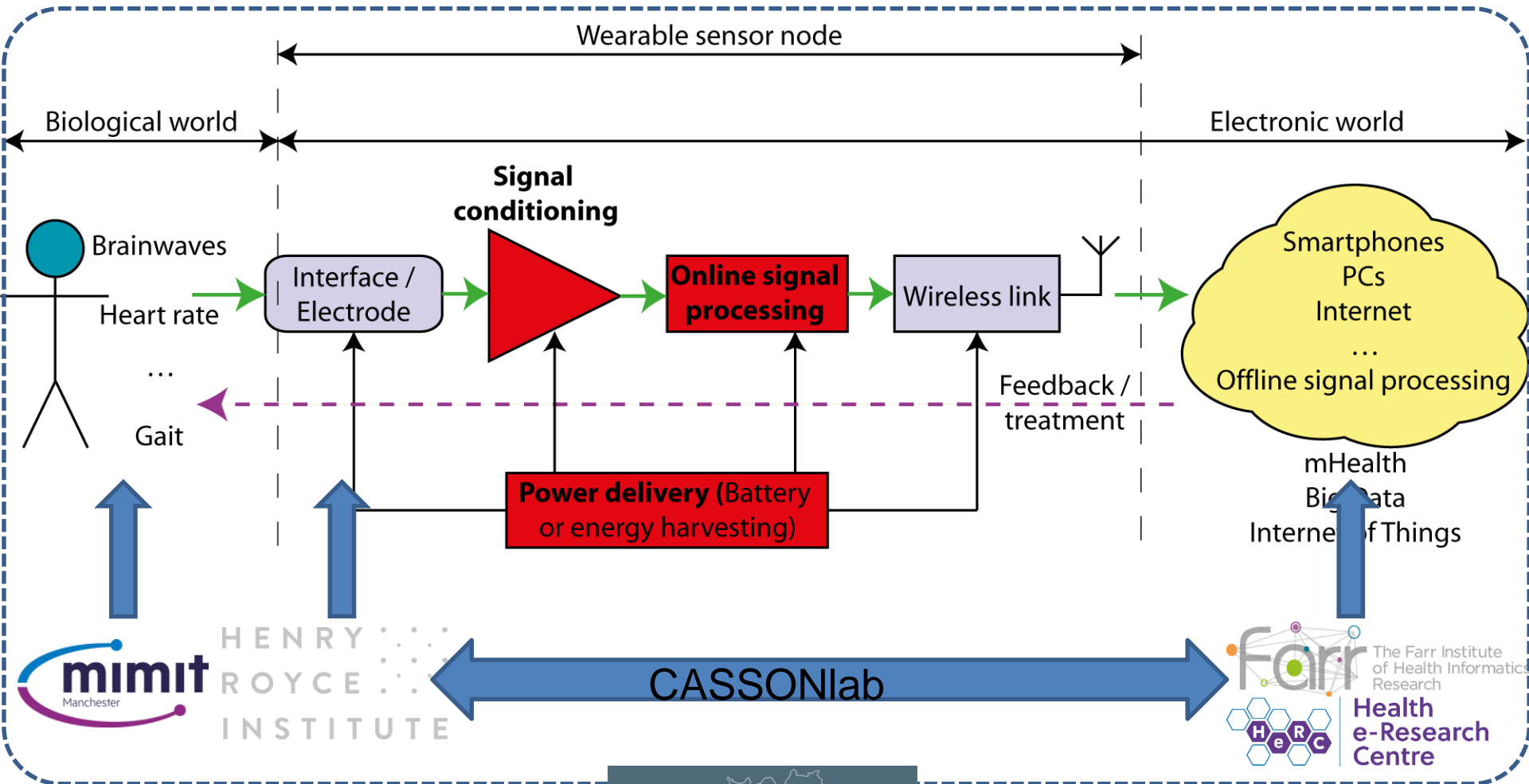


Number of people aged 60 or over is increasing exponentially

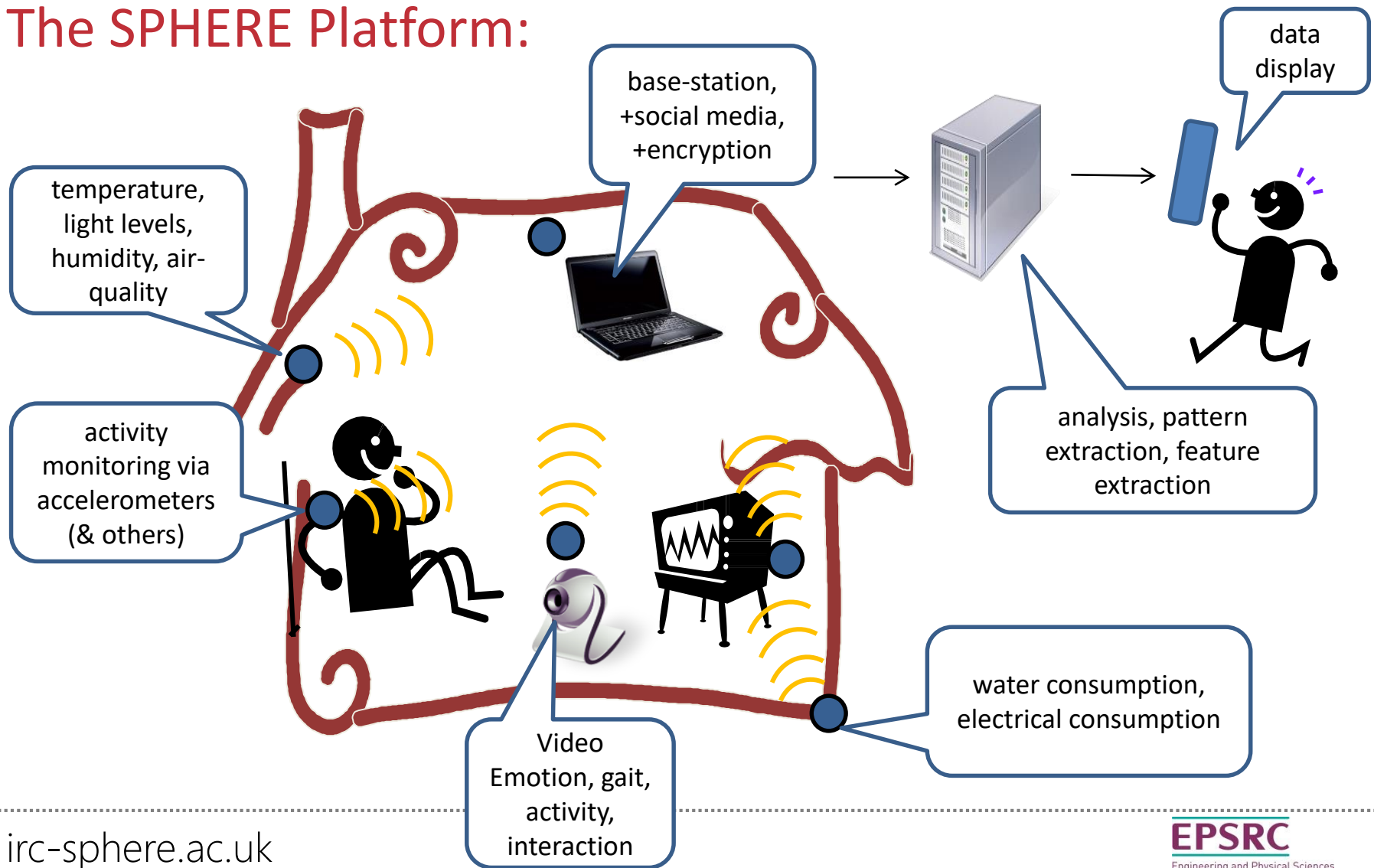
Our contribution: Smart sensors



Smart sensors in Manchester

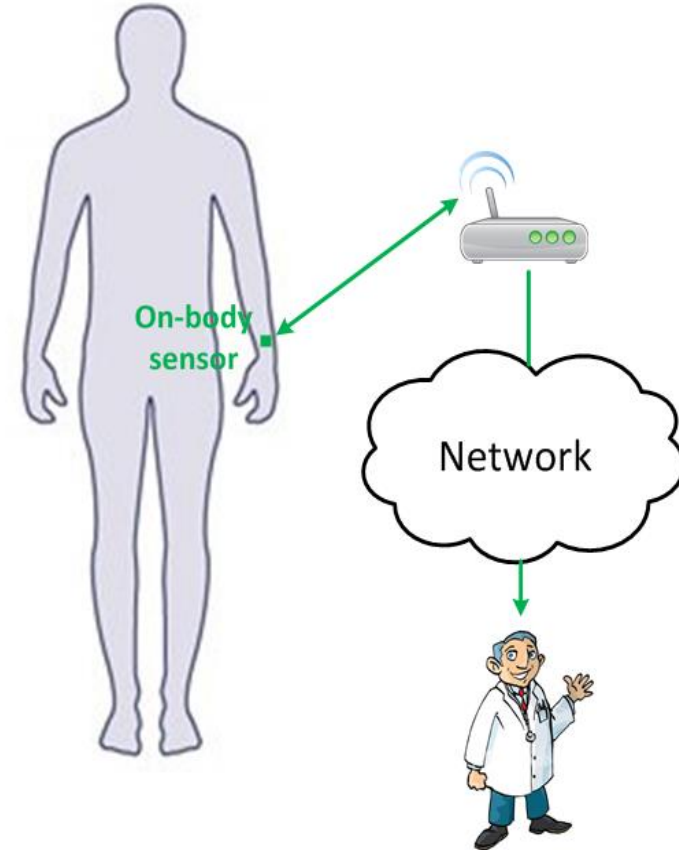


The SPHERE Platform:



On-body sensors

- Our vision is to develop a robust, energy efficient and user acceptable wireless communication system, consisting of:
 - one (or multiple) **on-body wearable sensor(s)** (e.g. accelerometer), and
 - an in-house **network of access points**.
- The system will provide real-time information regarding **physical activity** and **localisation**.
- The on-body node will ideally be powered using **energy harvesting** techniques.



Home Gateway

Intel NUC i5 running SPHERE software:
Data collection from sensor systems and real time classifier implementation.
Person-identification algorithm.
Raw data stored on encrypted SSD.



Summaries and diagnostics sent
via 3G/4G to SPHERE control
centre.

For liability reasons in this
context, raw data physically
collected on SSD.

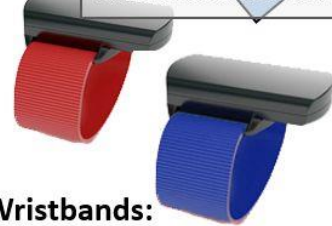
SPHERE Gateway

SPHERE hardware & firmware
BLE and 802.15.4 receivers.
ARM M3 processor.
Mains powered.



802.11ac network
- extracted features
- control signals

BLE or 802.15.4 network
Streamed sensor values



802.15.4 TSCH &
6LoWPAN/RPL mesh
network. raw sensor data



Wristbands:

SPHERE hardware:
TI CC2650 integrated circuit.
Dual accel. and gyro, ARM M3.
SPHERE firmware:
Implement BLE or TSCH MAC.
Battery power. Qi Wireless
charging.

Environmental Sensors

SPHERE hardware:
TI CC2650 integrated circuit.
PIR, temperature, light, humidity
SPHERE firmware:
Implementing 802.15.4 TSCH &
RPL mesh network.
Battery power (1 year life)

ASUS Xtion depth cameras

connected to Intel NUC i5
running SPHERE software
Implementing real time tracking,
bounding box, silhouette,
appearance and biometric
feature calculation



SPHERE GUI

For homeowner
On/Off
Delete Data
Data visualisation

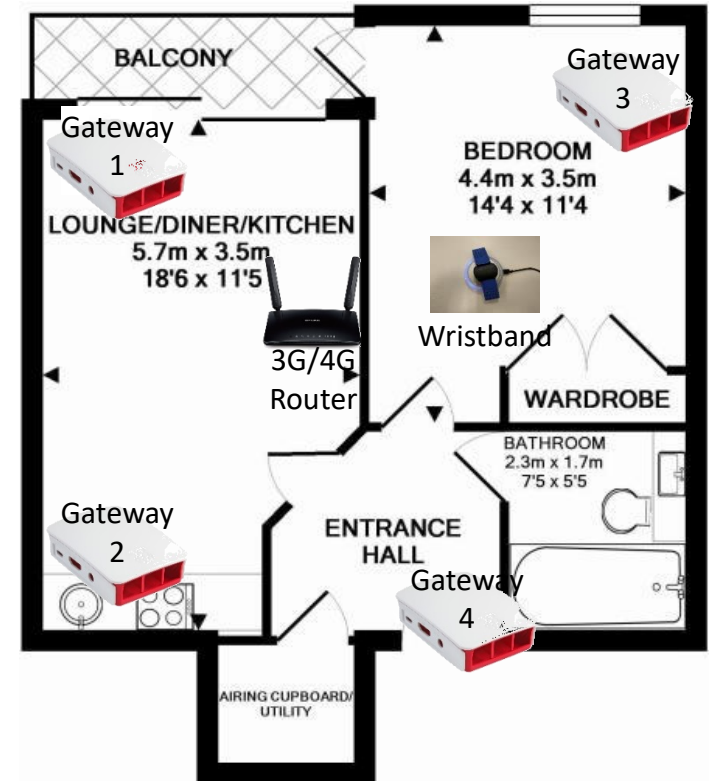


Off the Shelf CurrentCost Appliance Monitoring

Off the Shelf water sensor
connected to SPHERE
wireless solution

Example of SHERE IoT enabled house

- Items per system
 - 1 x **Router** 4G / 3G, WiFi
 - 4 x **Gateway** Raspberry Pi 3 (built in WiFi and BLE)
 - 1 x **Wristband** SPHERE (BLE, ADXL362, 4Gb FLASH)
- The Router and Gateways are fixed and the Wristband is worn by the participant
 - Charge when showering



TOTAL APPROX. FLOOR AREA 44.9 SQ.M. (484 SQ.FT.)
Made with Metropix ©2011

Motivation for this work

Want to integrate heart monitoring into this platform

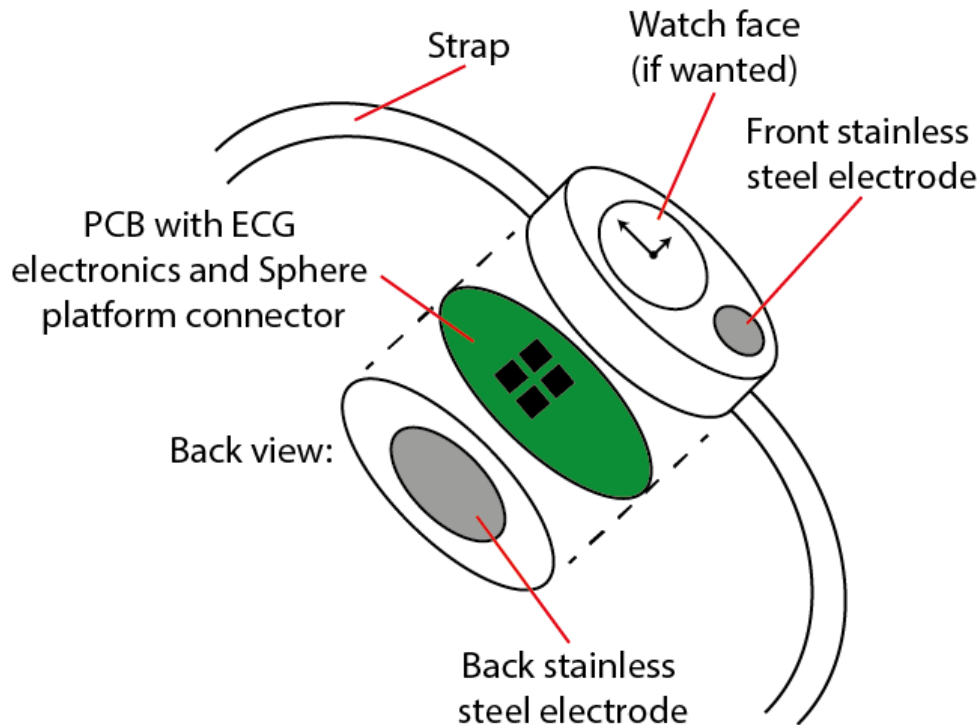
PPG is common in smart watches, and great, but:



- × High power due to light source.
- × Difficult to extract heart rate variability.
- × Less clinically relevant information than the ECG.
- × Not integrated with IoT infrastructure.

Our aim

We aim to make a wrist worn ECG device



- ✓ Lower power as no light source.
- ✓ Very light weight.
- ✓ ECG waveform allowing heart rate, heart variability, and other measures to be extracted.
- ✓ More healthcare relevant information than just PPG.
- ✓ Integrate with SPHERE IoT infrastructure.

Design requirements

Power:

1 month battery life

10 μW power consumption, 1.8 V supply

Data:

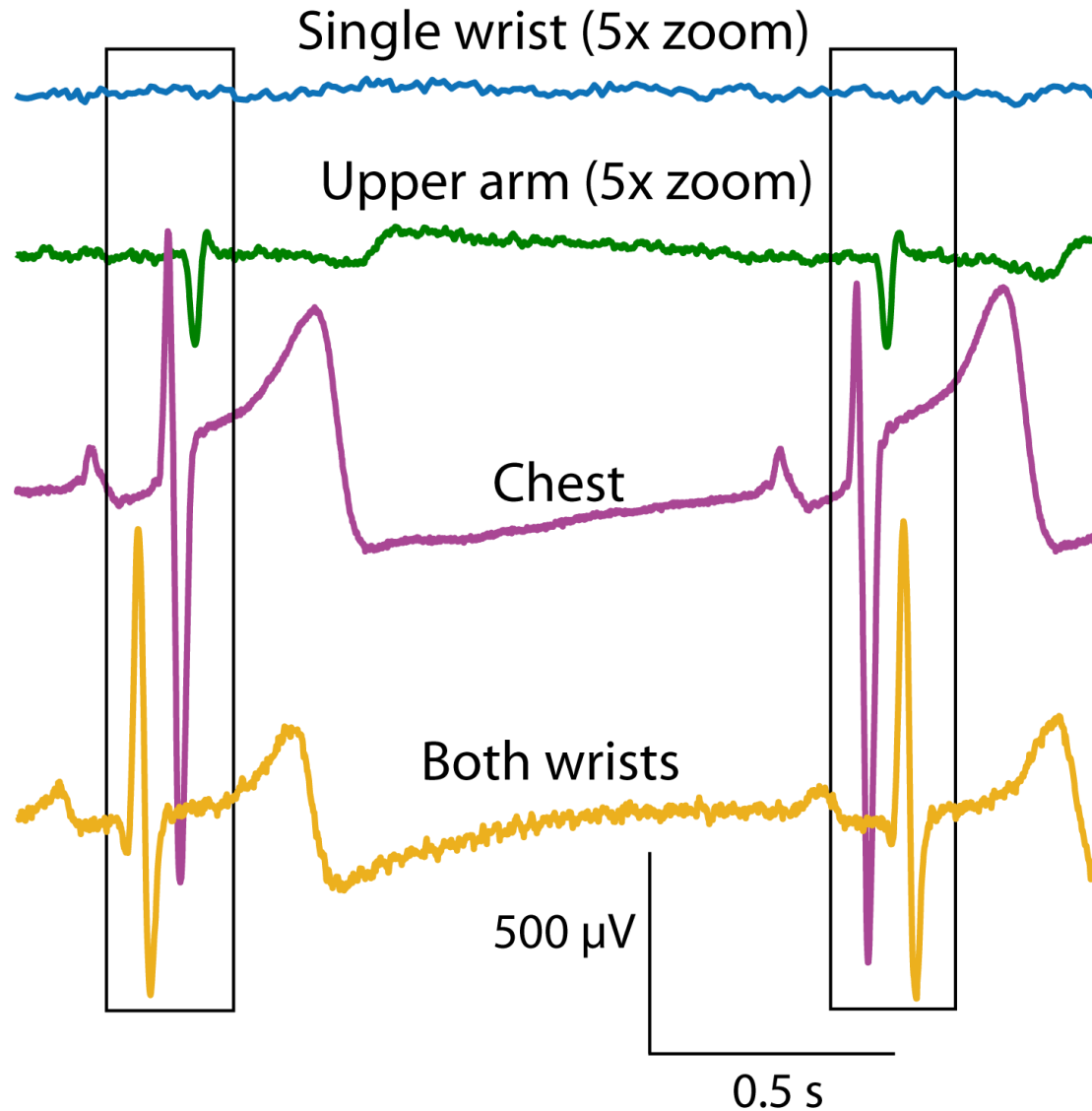
Ability to extract HR and HRV

Accuracy of a few BPM

Safety:

18 μA max fault current into user

Challenge of wearable ECG



Final system

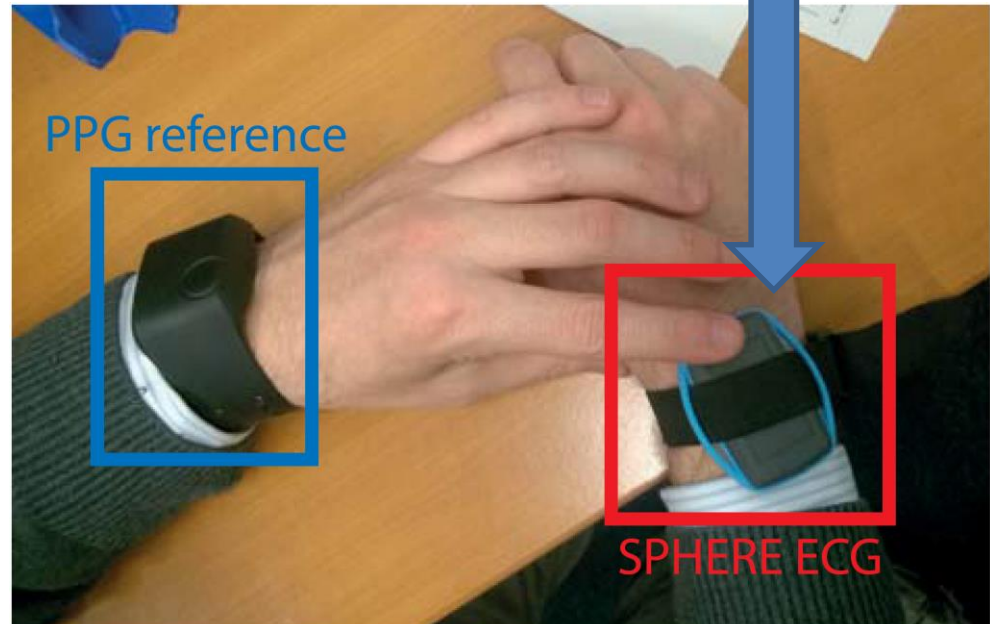
Painted electrode



Electrodes on both sides of body to get large signal size



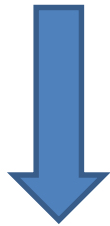
SPHERE board ECG board



Electrodes

Medical grade Silver/Silver Chloride electrodes painted on

Allows size and shape personalisation



Painted electrode



3D print for more complicated shapes and personalisation



Bumps for penetrating hair



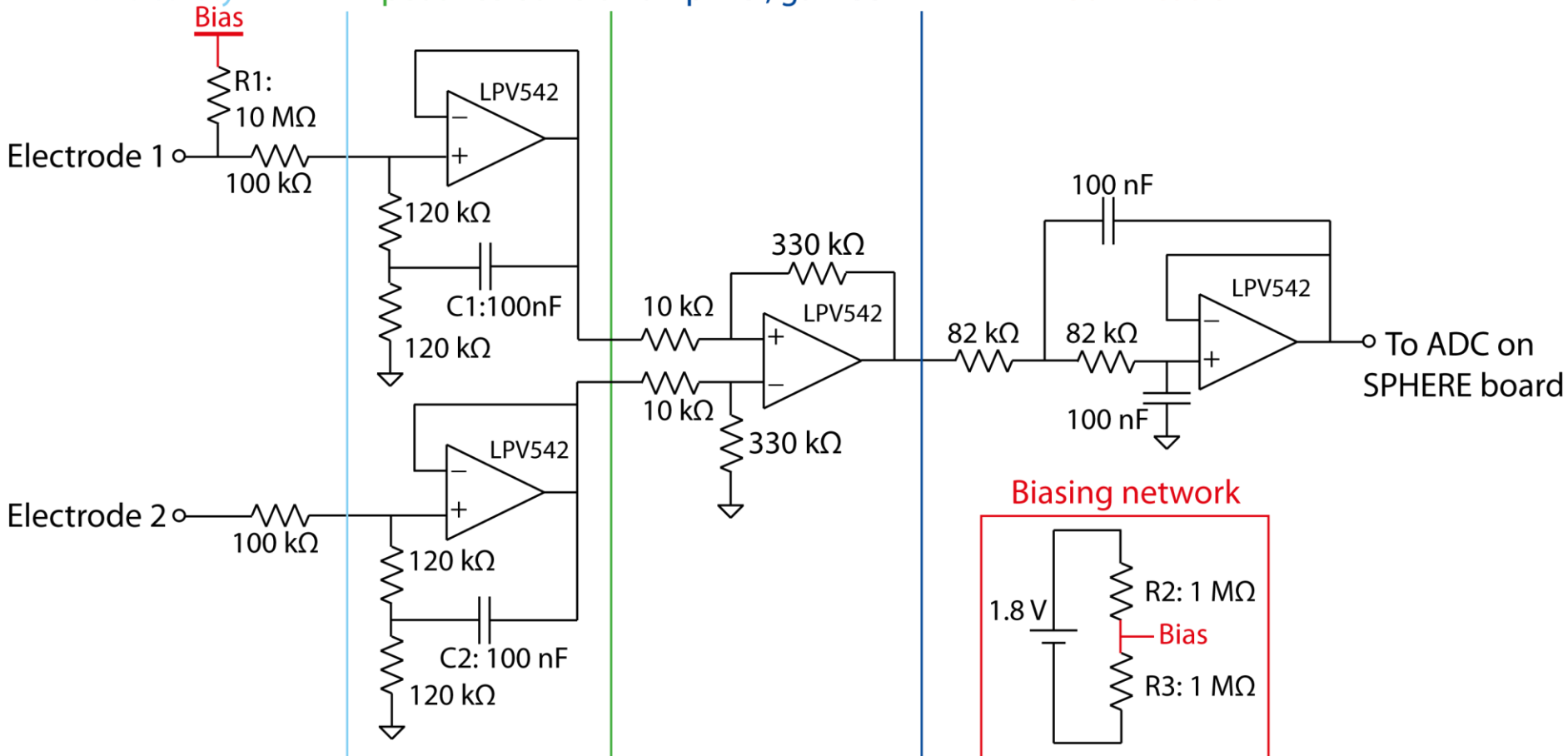
Front-end circuit design

User connection
and safety

High input
impedance buffers

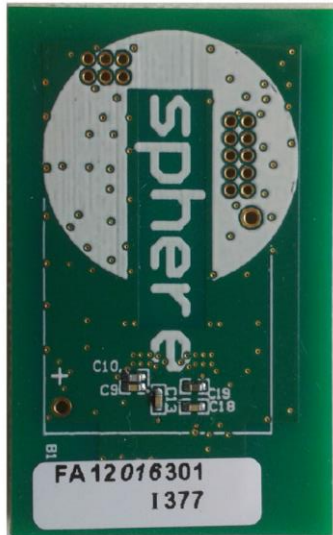
Difference
amplifier, gain 33

Second order low pass filter,
30 Hz cut-off



SPHERE wearable

4 cm



- ✓ Based on TI CC2640
- ✓ Bluetooth low energy wireless transmission
- ✓ Local SD card memory
- ✓ 100 mAh rechargeable Li-ion battery
- ✓ Qi wireless charging
- ✓ ADXL 3-axis accelerometer

- ✓ We use the TI **sensor controller** for low power ADC and data storage without waking main MCU.

Software

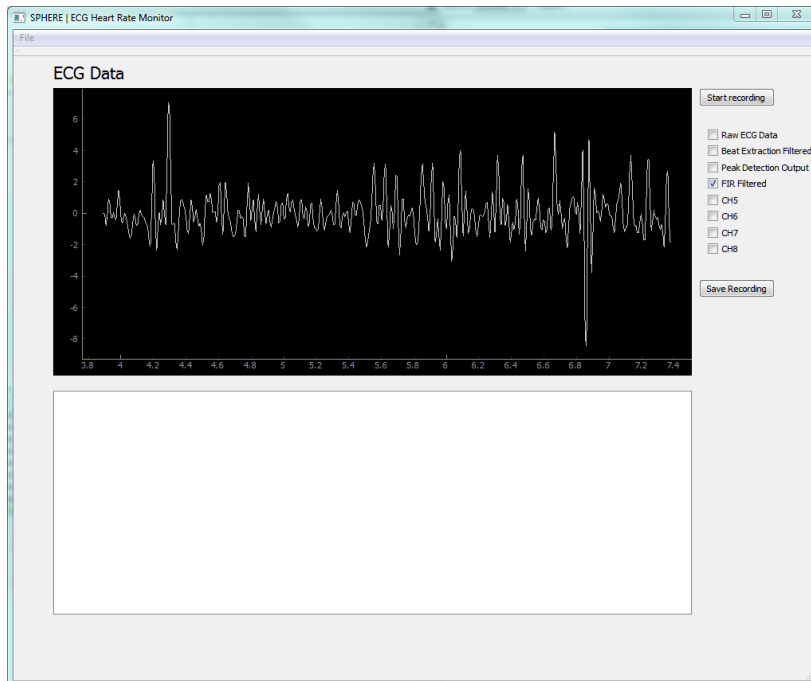
Two options for communication

‘Connected’ state

- ✓ All data transmitted over BLE.
- ✓ Allows online and offline signal analysis in basestation.

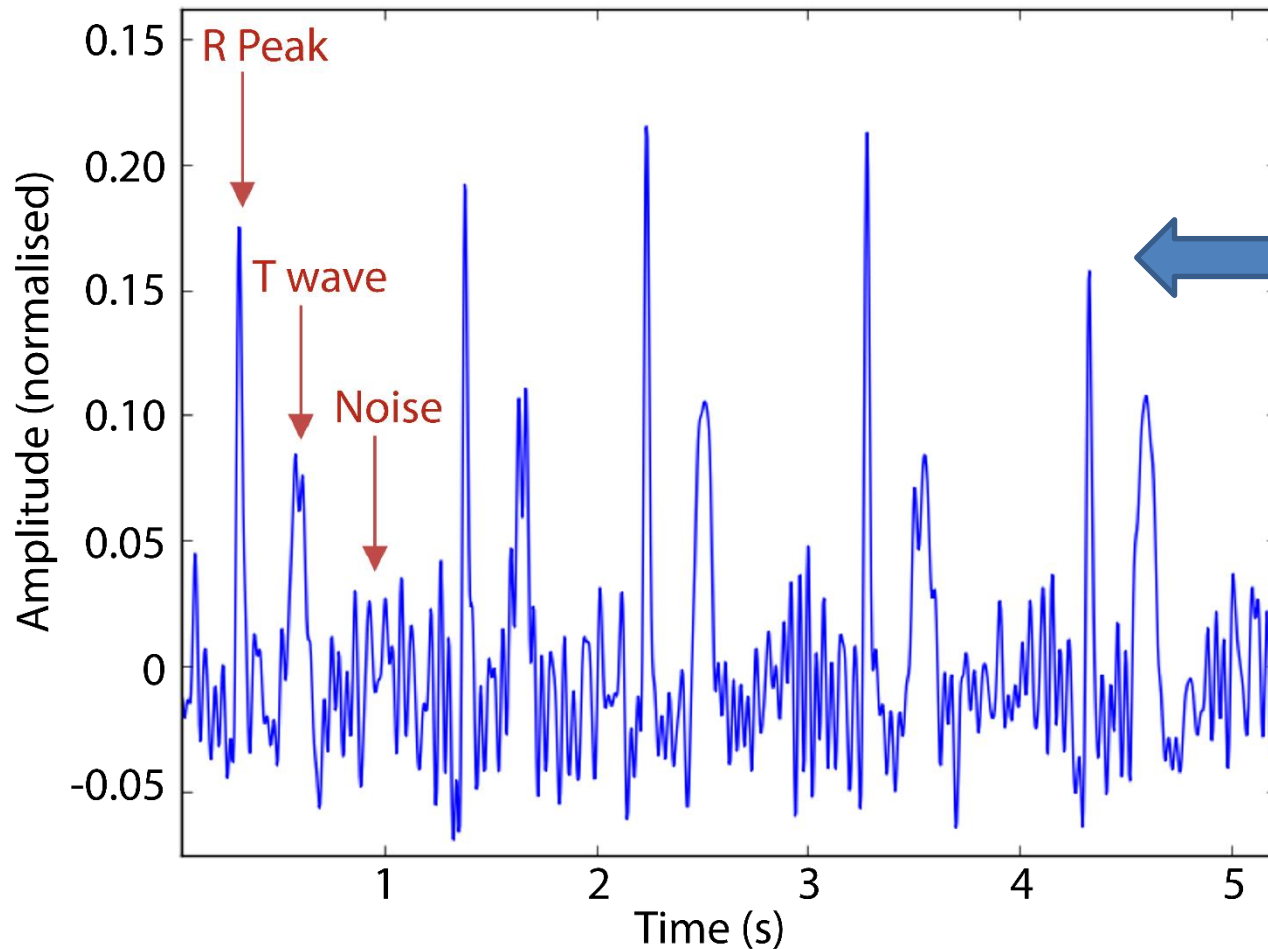
‘Not-connected’ state

- ✓ Heart rate calculated on the wearable.
- ✓ Just heart rate information transmitted.
- ✓ Low data rate means data can be included in the BLE advertising packets.
- ✓ Reduces power significantly if full ECG trace is not required.



System performance

Raw data shows clear heart beats



This is before
standard ECG
processing for
noise removal

Deliberately high
noise floor to
keep power low

Comparison to a reference PPG

Methods for offline comparison of data

1. Raw ECG data filtered: Butterworth first order low pass, highpass and notch
2. Baseline wander removed with Discrete Wavelet Transform
3. Noise removed from ECG using extended Kalman filter
4. Heart beats detected using standard Pan-Tompkins algorithm
5. Heart rate calculated from beat timings
6. Kalman tracking filter applies with zero-order-hold model to smooth heart rate when incorrect (missing and/or additional) beats detected

Compare heart rate to that from a commercial PPG device

System performance

Accurate to within a few beats to minute

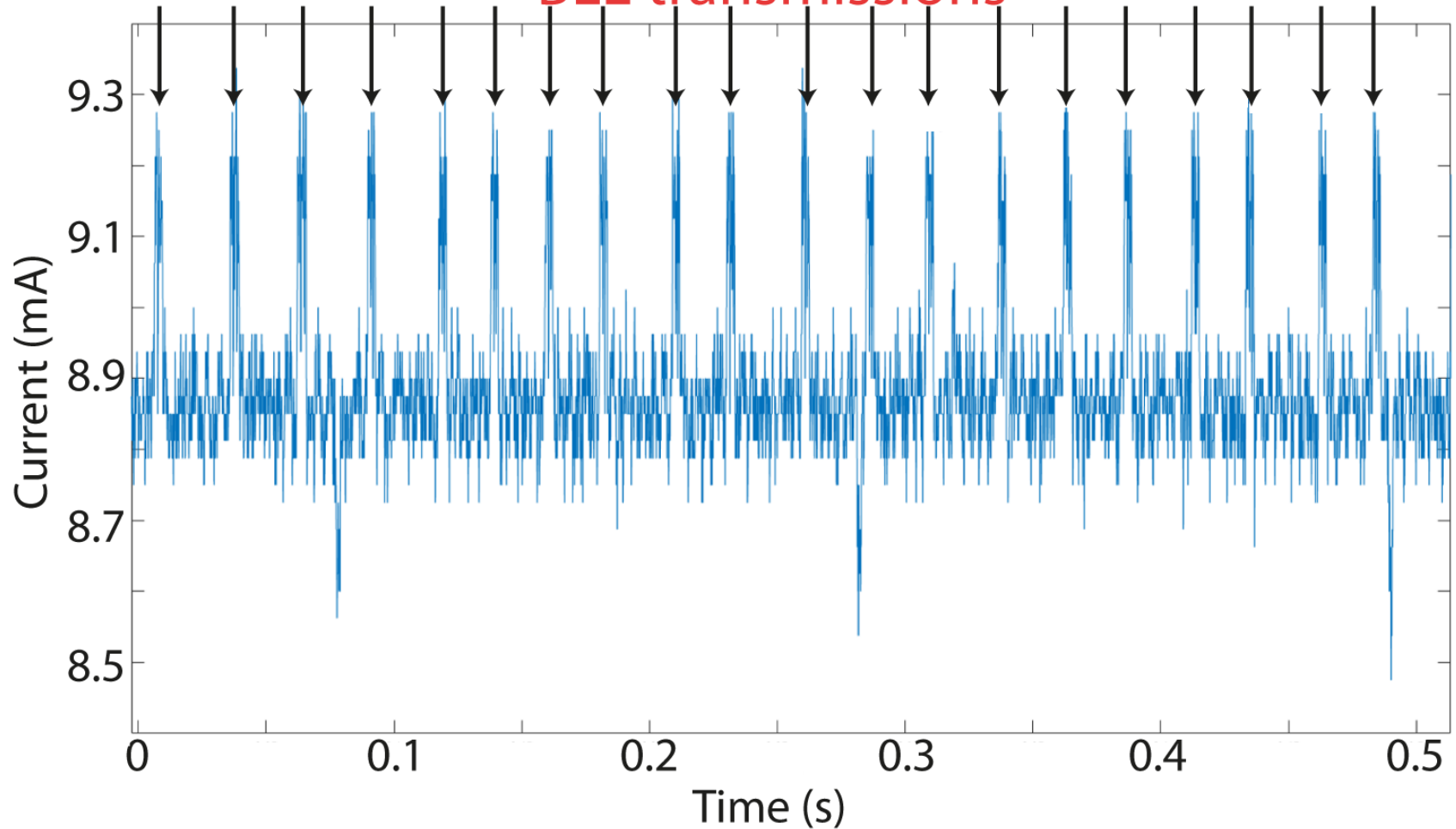
| Record | Mean difference (BPM) | Standard deviation (BPM) |
|-------------|-----------------------|--------------------------|
| 1 | 5.61 | 4.02 |
| 2 | 2.87 | 1.80 |
| 3 | 6.77 | 6.08 |
| 4 | 11.3 | 2.59 |
| 5 | 2.66 | 2.87 |
| 6 | 5.96 | 6.17 |
| 7 | 1.83 | 1.85 |
| 8 | 2.19 | 2.29 |
| 9 | 1.88 | 1.36 |
| Mean | 4.56 | 3.23 |

Note: We estimate the reference PPG device is accurate to 2 bpm

System power consumption

Connected mode:

BLE transmissions



Quiescent: 11.3 μ A

System power consumption

If measure heart rate for 30 s every hour this gives an average current draw of 86 μ W

Gives 48 days battery life from a miniature 100 mAh battery

Step change for a wrist worn ECG IoT device weighing only 29 g

Future steps

- ✓ Improve signal quality further
- ✓ Use electrodes to provide touch input to the wearable
- ✓ Add 3 electrode design
- ✓ Use capacitive driven right leg

Summary

30 g ultra low power ECG unit

Integrates with UK IoT smart home demonstrator

Battery life >1 month

Personalisable electrodes

christopher.beach@manchester.ac.uk

www.eee.manchester.ac.uk/sisp