Wrist-Worn Photoplethysmography Data: Opportunities and Challenges



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My General Background



New Horizons in Sensor Development

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ABSTRACT

INTILLE, S. S., J. LESTER, J. F. SALLIS, and G. DUNCAN. New Horizons in Sensor Development. *Med. Sci. Sports Exerc.*, Vol. 44, No. 1S, pp. S24–S31, 2012.

EMERGING TECHNOLOGY AND METHODS

We expect to see a gradual improvement in overall device performance rather than the development of fundamentally new types of sensors in physical activity measurement devices. Breakthroughs will likely result from using multimodal sensor fusion—combining data from several types of sensors, sometimes located on different parts of the body or in the environment—into a single system that is then used to infer precise, second-by-second detail about physical activity type, amount, and location. These changes, in turn, will create new opportunities in methods. New monitors should permit longer term, lower cost, higher compliance deployments enabling a broader spectrum of physical activity concepts to be simultaneously measured in real-life settings.

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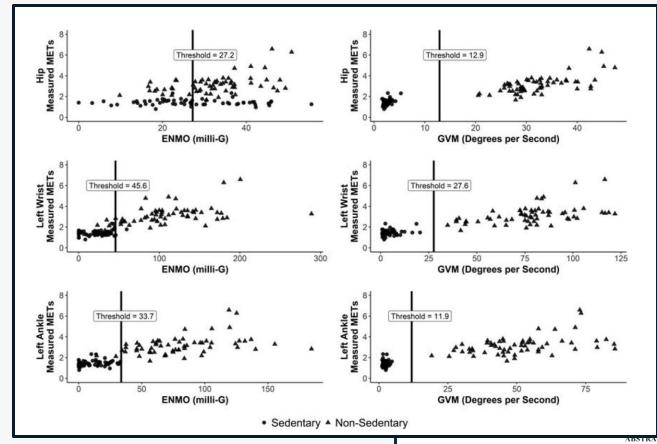
Estimating Energy Expenditure with ActiGraph GT9X Inertial Measurement Unit

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ABSTRACT

HIBBING, P. R., S. R. LAMUNION, A. S. KAPLAN, and S. E. CROUTER. Estimating Energy Expenditure with ActiGraph GT9X Inertial Measurement Unit. *Med. Sci. Sports Exerc.*, Vol. 50, No. 5, pp. 1093–1102, 2018.



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Redefining the Roles of Sensors in Objective Physical Activity Monitoring

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ABSTRACT

CHEN, K. Y., K. F. JANZ, W. ZHU, and R. J. BRYCHTA. Redefining the Roles of Sensors in Objective Physical Activity Monitoring. *Med. Sci. Sports Exerc.*, Vol. 44, No. 1S, pp. S13–S23, 2012.

Types of Sensors (Chen et al., 2012)

- Movement (inertial)
- Physiologic
- Contextual

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My Background with PPG



JOURNAL OF SPORTS SCIENCES, 2018 VOL. 36, NO. 15, 1734–1741 https://doi.org/10.1080/02640414.2017.1412235





Comparative evaluation of heart rate-based monitors: Apple Watch vs Fitbit Charge HR

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Introduction

The expanding market in wearable activity monitors for consumers has generated considerable interest within the physical activity (PA) research community (https://clinicaltrials.gov/ ct2/results?term=activity+tracker&pg=1). Most monitors continue to predict energy expenditure (EE) and step counts, while newer versions of many wrist-worn consumer monitors have also begun incorporating heart rate measures using a technology called photoplethysmography (https://support. apple.com/en-us/HT204666; https://www.fitbit.com/pure pulse). This involves flashing green LED light through the skin, to detect the expansion and contraction of wrist capillaries with each pulse (Maeda, Sekine, & Tamura, 2011). Algorithms are then applied to estimate heart rate continuously from that information (Ahmadi, Moradi, Malihi, Karimi, & Shamsollahi, 2015).





f heart rate-based monitors: Apple Watch vs Fitbit Charge HR antine Mantis of and Gregory J. Welk

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Fitness Trackers Can Be Fashionable and Functional

Paul R. Hibbing, M.S. Samuel R. LaMunion, M.S. Lindsay P. Toth, M.S.



Halfpoint/iStock/Thinkstock

Pitness tracking has become a major industry, with new devices that uniquely combine form and function constantly entering the market. The products range from

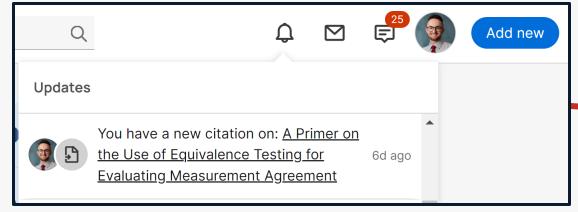
Volume 19/Number 3/October 2017

Some monitors measure heart rate using a new technology called photoplethysmography, which shines LEDs through the skin to capture the expansion and contraction of blood vessels as blood moves through.

(continued on page 4)

ACSM Fit Society, Page 3

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2017 / 2018	



Dramatization

ORIGINAL ARTICLE

Twenty-Four-Hour Ambulatory Blood Pressure Measurement Using a Novel Noninvasive, Cuffless, Wireless Device

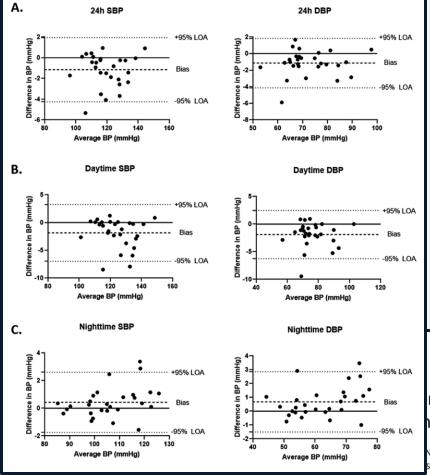
Dean Nachman,^{1,2,*} Adi Gilan,^{1,*,†} Nir Goldstein,³ Keren Constantini,³ Romi Littman,⁴ Arik Eisenkraft,^{1,4,o} Ehud Grossman,^{3,5} and Yftach Gepner³



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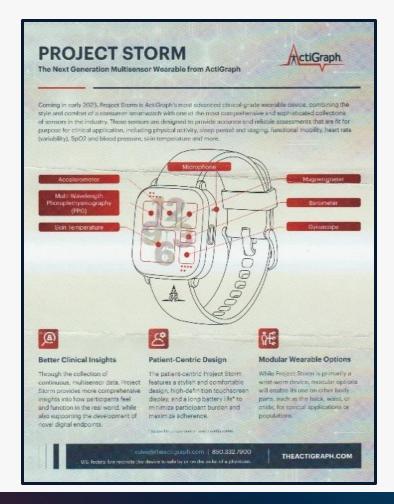
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The Quest for Raw Signals: A Quality Review of Publicly Available Photoplethysmography Datasets

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ADDS 2024 ActiGraph ActiGraph 18

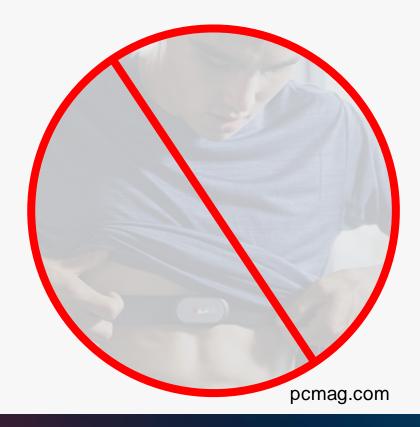
The Light-Tissue Interaction of Pulse Oximetry

Paul D. Mannheimer, PhD

The underlying science of pulse oximetry is based on a simple manipulation of the Lambert-Beer law, which describes the attenuation of light traveling through a mixture of absorbers. Signals from detected red and infrared light that has traveled through blood-perfused tissues are used to estimate the underlying arterial hemoglobin oxygen saturation. However, light scatters in tissue and influences some of the simplifications made in determining this relationship. Under most clinical circumstances, the empirical process that manufacturers use to calibrate the system during its design readily accommodates this and results in accurate readings. The same tissue light scattering properties allow sensors to be configured for use on opposing or adjacent surfaces, provided that the placement sites offer sufficient signal strength and are absent factors known to influence accuracy. In this paper I review the light-tissue interaction in pulse oximetry and describe some of the assumptions made and their implications. Certain deviations from the nominal conditions, whether clinical in nature or misuse of the product, can affect system performance. Consequently, users should be cautious in modifying sensors and/or using them on tissue sites not intended by the manufacturer (off-label use). While perhaps helpful for obtaining pulsatile signals or extending the lifetime of a sensor, some practices can disrupt the optical integrity of the measurement and negatively impact the oxygen saturation reading accuracy.

(Anesth Analg 2007;105:S10-7)

Heart Rate and Heart Rate Variability (obviously)



Other Cardiovascular Parameters



Figure 2: Taking a Measurement with the Digital Crown and the Watch App

The ECG rhythm will be classified into one of the following categories:

- 1. Sinus rhythm
- 2. Atrial Fibrillation
- 3. Inconclusive

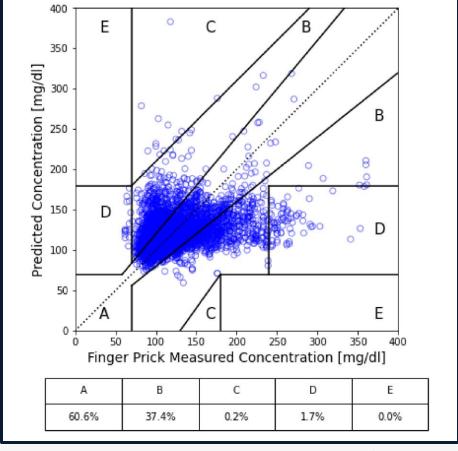
accessdata.fda.gov

Predicting Energy Expenditure and Activity Type



... It's stock alright? Are you reading this?

Glucose Monitoring



doi: 10.3390/s21237815

ARTICLE OPEN

Investigating sources of inaccuracy in wearable optical heart rate sensors

Brinnae Bent¹, Benjamin A. Goldstein², Warren A. Kibbe 6 and Jessilyn P. Dunn 6 and Jessilyn P. Dunn 6 and Jessilyn P. Dunn 7 and 7 an

Skin Type

- AgeSex
- Obesity status
- Skin tone

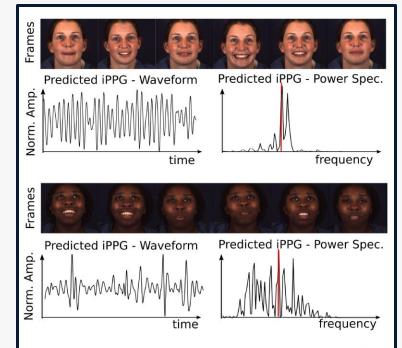
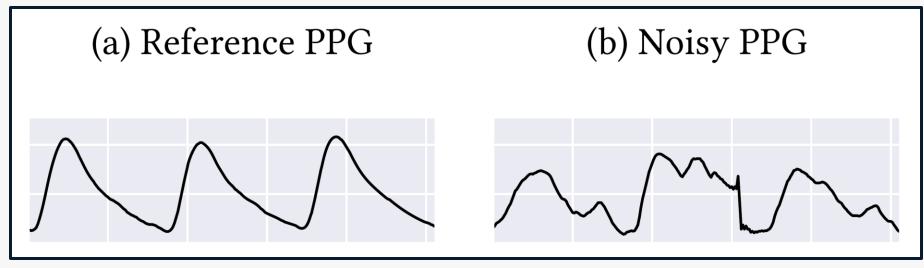


Figure 1. Characterizing the performance of computer vision algorithms is important to ensure that some demographic groups are not exposed to greater risk. We perform a meta-analysis of the impact of skin type on imaging photoplethysmography measurement. The examples shown illustrate that lighter skin types are associated with higher blood volume pulse signal-to-noise ratios. The red line indicates the heart rate measured via a contact sensor.

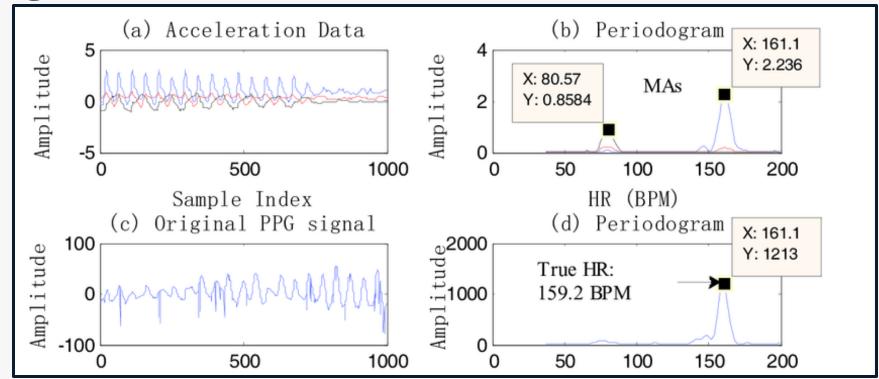
doi: 10.1109/CVPRW50498.2020.00150

Motion Artifact



doi: 10.1145/3563949

Signal Crossover



doi: 10.3390/s17112450

Thank You!

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