





Walkbotr: Extracting ‘Walk Bouts’ from GPS and Accelerometry Data for Physical Activity Research

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Walking is a common form of exercise and a popular intervention of interest for population health researchers. GPS devices and accelerometers, which can be worn by study participants to provide accurate information on walking duration, speed, and intensity, are an efficient and cost-effective method for measuring walking characteristics (Feng & Timmermans, 2013, Troped et al. (2008)). The outcome of research on walking can inform policy related to health communication, medical interventions, and built environments. However, data collected by these devices have two main drawbacks: they require intensive data preprocessing to identify walking periods, and they may contain identifying personal information, limiting storage and shareability options. To address these challenges, we have developed an R package, walkbotr, to identify walking periods within monitoring data and to create de-identified summaries of walking characteristics that can be used to answer research questions related to walking.

walkbotr defines periods of walking (or other activity) based on pre-specified, but adjustable, parameters from GPS and accelerometry data. The main data components used to identify walking are speed, activity level, and distance traveled measured by accelerometers and GPS devices. walkbotr splits raw accelerometry data into small discrete time chunks or epochs (defaults to 30 seconds), and then aggregates these chunks into physical activity bouts when movement rises above a minimum activity threshold for a minimum number of sequential epochs. The algorithm defines walking as periods of activity where the counts per epoch, or level of activity, fall within a pre-specified range consistent with walking (Figure 1) (Kang et al., 2013).

Once the algorithm identifies a sequence as potentially walking, it then defines the period as a walk bout if the GPS trace indicates travel outside a specified radius. This allows the researcher to distinguish periods of walking from running or other high-

THE DOWNLOAD



Citation:

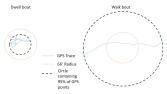
Wilner, L. B., Zhou, W., Hurvitz, P. M., Moudon†, A. V., Kang, B., Saelens, B. E., Phuong, J., Dekker, M., & Mooney, S. J. (2026). Walkbotr: Extracting ‘Walk Bouts’ from GPS and Accelerometry Data for Physical Activity Research. <https://doi.org/DOI pending>

Published: 2026-03-16

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Related Denominator



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Walking is the most common form of physical activity and a behavior of key interest for urban planners, health promotion researchers, and rehabilitation medicine practitioners. Data collected from monitoring devices, such as Global Positioning System (GPS) trackers and accelerometers, hold considerable public health research potential. By analyzing patterns in individual energy expenditure and movement... [read more](#)

intensity exercise, as well as from periods where a person may be moving (e.g., doing chores around the house) but not specifically walking. One major benefit of walkbotr is that parameters defining walk bouts are set at reasonable defaults, allowing for comparison across studies, but they can be changed by the user. This grants researchers flexibility in answering specific questions about walking characteristics as well as an easy way to conduct sensitivity analyses on walking definitions.

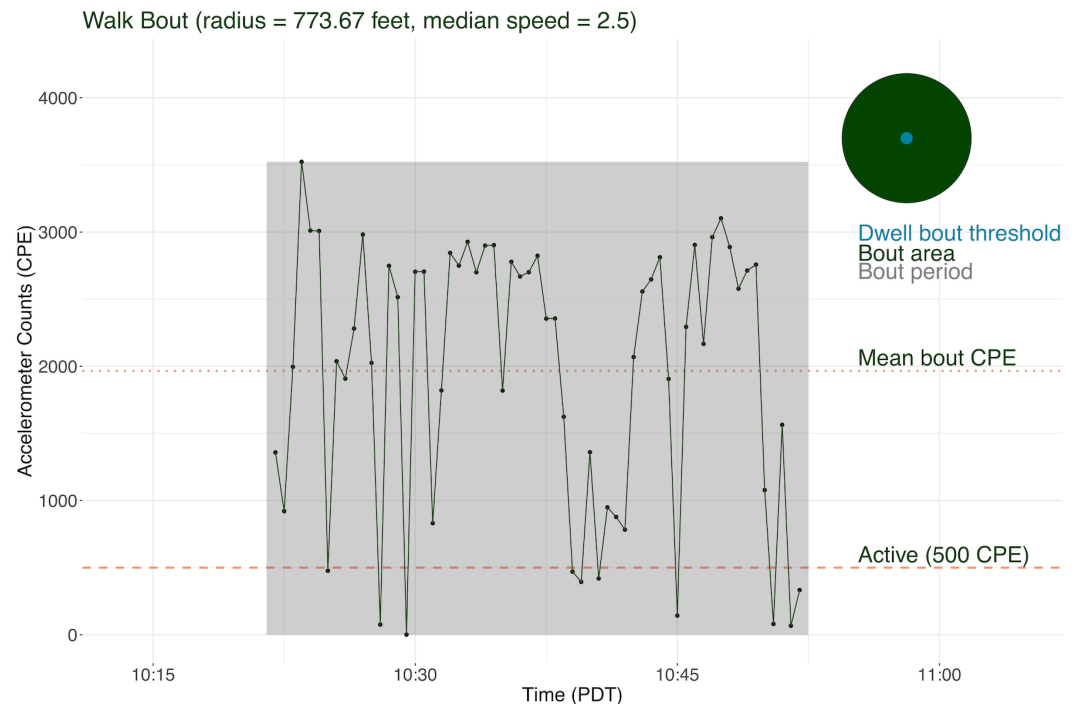


Figure 1: Example of a walk bout. The physical activity bout in the gray box is classified as a walk bout, because (1) the activity CPE is above the active threshold and are consistent with that of walking and (2) the bout area is larger than that of the dwell bout. The circles show the size of the walk bout (green) in relation to the dwell bout max circumference (blue). These two measures define walks and the application allows researchers to set the activity and bout area thresholds to define active walking for reproducibility and comparison to other research sites.

Once the data are processed in walkbotr – and walking periods, or walk bouts, are defined – a full dataset is created that includes the raw data with the newly created bout definitions attached to each observation. In addition, a second summary dataset of walk bouts is created. These data are deidentified (i.e., time and location are removed) but include relevant walking information, such as distance traveled and time spent walking, that can be used for analyses and shared with other researchers without worrying about compromising study participant privacy. walkbotr aims to make the measurement of walking easier and more systematic by removing barriers to

researchers interested in collecting, using, and sharing individual data on walking. For those interested in additional details, please refer to the related Denominator article.

Computation & Reproducibility

The original repository maintained by Lauren Wilner can be found here: <https://github.com/rwalkbout/walkbotr>. Note: this repository is maintained by Lauren Wilner and may differ from that originally used to produce the results in this publication.

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