








# 1 LightLogR: Reproducible analysis of personal light 2 exposure data

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## Software

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## 11 Summary

12 Light plays an important role in human health and well-being, which necessitates the study of  
13 the effects of personal light exposure in real-world settings, measured by means of wearable  
14 devices. A growing number of studies incorporate these kinds of data to assess associations  
15 between light and health outcomes. Yet with few or missing standards, guidelines, and  
16 frameworks, setting up measurements, analysing the data, and comparing outcomes between  
17 studies is challenging, especially considering the significantly more complex time series data  
18 from wearable light loggers compared to controlled stimuli used in laboratory studies. In this  
19 paper, we introduce LightLogR, a novel resource to facilitate these research efforts in the form  
20 of an open-source, GPL-3.0-licenced software package for the statistical software R. As part of  
21 a developing software ecosystem, LightLogR is built with common challenges of current and  
22 future datasets in mind. The package standardizes many tasks for importing and processing  
23 personal light exposure data, provides quick as well as detailed insights into the datasets  
24 through summary and visualization tools, and incorporates major metrics commonly used in  
25 the field (61 metrics across 17 metric families), while embracing an inherently hierarchical,  
26 participant-based data structure.



Figure 1: LightLogR logo

## 27 Statement of need

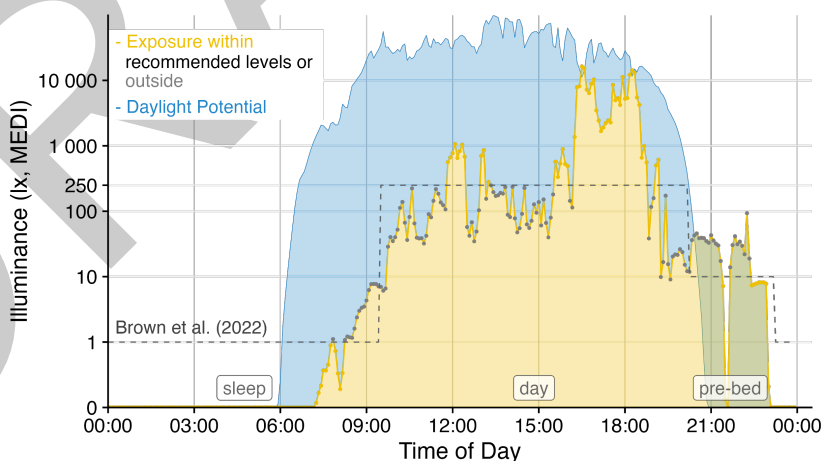
28 Personalized luminous exposure data is progressively gaining importance across various domains,  
29 including research, occupational affairs, and lifestyle tracking. Data are collected through  
30 a proliferating selection of wearable light loggers and dosimeters, varying in size, shape,

31 functionality, and output format (Hartmeyer et al., 2023). Despite or potentially because of  
 32 numerous use cases, the field still lacks a unified framework for collecting, validating, and  
 33 analyzing the accumulated data (Hartmeyer et al., 2023; Spitschan et al., 2022). This issue  
 34 increases the time and expertise necessary to handle such data and also compromises the  
 35 FAIRness (findability, accessibility, interoperability, reusability) (Wilkinson et al., 2016) of the  
 36 results, especially for meta-analyses (Vries et al., 2024).

37 LightLogR (Figure 1) was designed to be used by researchers who deal with personal light  
 38 exposure data collected from wearable devices. These data are of interest for various disciplines,  
 39 including chronobiology, sleep research, vision science and epidemiology, as well as for post-  
 40 occupancy evaluations in architecture and lighting design. The package is intended to streamline  
 41 the process of importing, processing, and analysing these data in a reproducible and transparent  
 42 manner. The package is available on GitHub (Zauner, Hartmeyer, et al. (2023b)) and CRAN  
 43 (Zauner et al. (2024a)), has a dedicated website for documentation and tutorials (Zauner,  
 44 Hartmeyer, et al. (2023a)), and releases are archived on Zenodo (Zauner et al. (2024b)).

45 LightLogR's key features include:

- 46 ■ a growing list of supported devices with pre-defined import functions tailored to their  
 47 data structure (17 at the time of writing, see Table 2),
- 48 ■ preprocessing functions to combine different time series, aggregate and filter data, and  
 49 find and deal with implicitly missing data,
- 50 ■ visualization functions to quickly explore the data. These function are based on the  
 51 popular ggplot2 (Wickham, 2016) plotting package and are designed to be easily  
 52 customizable to construct publication-ready figures (see, Figure 2),
- 53 ■ a large and growing set of metrics that cover most if not all major approaches found in  
 54 the literature (at the time of writing 61 metrics across 17 metric families, see Table ),  
 55 accessible via a consistent function interface.



**Figure 2:** Light logger data can powerfully convey insights into personal light exposure and health-related outcomes. LightLogR facilitates the import and combination of different data sources into a coherent data structure, as seen here by combining environmental daylight availability and personal light exposure with data from a sleep diary. The visualization functions in the package further allow customization to produce publication-ready figures. This figure was created with the 'gg\_day()' function. The creation process is part of a tutorial (Zauner, Hartmeyer, et al., 2023c) on several key functions in the package.

Device Name	Manufacturer
Actiwatch Spectrum	Philips Respironics

Device Name	Manufacturer
ActLumus	Condor Instruments
ActTrust	Condor Instruments
DeLux	Intelligent Automation Inc.
GENEActiv <sup>1</sup>	Activeinsights
Kronowise	Kronohealth
Lido	Lucerne University of Applied Sciences and Arts
LightWatcher	Object-Tracker
LIMO	École nationale des travaux publics de l'État (ENTPE)
LYS Button	LYS Technologies
Motion Watch 8	CamNtech
melanopiQ Circadian Eye	Max Planck Institute for Biological Cybernetics
XL-500 BLE	NanoLambda
OcuWEAR	Ocutune
Speccy	Monash University Malaysia
SpectraWear	University of Manchester
VEET	Meta Reality Labs

Table 2: Devices supported for import in version 0.4.1

Metric Family	Submetrics	Note	Documentation
Barroso	7		barroso_lighting_metrics()
Bright-dark period	4x2	bright / dark	bright_dark_period()
Centroid of light exposure	1		centroidLE()
Disparity index	1		disparity_index()
Duration above threshold	3	above, below, within	duration_above_threshold()
Exponential moving average (EMA)	1		exponential_moving_average()
Frequency crossing threshold	1		frequency_crossing_threshold()
Intradaily Variance (IV)	1		intradaily_variability()
Interdaily Stability (IS)	1		interdaily_stability()
Midpoint CE (Cumulative Exposure)	1		midpointCE()
nvRC (non-visual circadian response)	4		nvRC(), nvRC_circadianDisturbance(), nvRC_circadianBias(), nvRC_relativeAmplitudeError()
nvRD (non-visual direct response)	2		nvRD(), nvRD_cumulative_response()
Period above threshold	3	above, below, within	period_above_threshold()

<sup>1</sup>Available after processing of the data using GGIR (Migueles et al., 2019).

Metric Family	Submetrics	Note	Documentation
Pulses above threshold	7x3	above, below, within	pulses_above_threshold()
Threshold for duration	2	above, below	threshold_for_duration()
Timing above threshold (TAT)	3	above, below, within	timing_above_threshold()
<b>Total:</b>			
<b>17 families</b>	<b>61 metrics</b>		

56 : metrics available in version 0.4.1

57 LightLogR is already being used in several research projects across scientific domains, including:

- 58     ▪ an ongoing cohort study to collect light exposure data across different geolocations (Guidolin et al., 2024),
- 59     ▪ an ongoing cohort study to collect year-long datasets of various types of environmental and behavioral data (Biller et al., 2024),
- 60     ▪ a novel power analysis method for personal light exposure data (Zauner, Udovicic, et al., 2023),
- 61     ▪ an intervention study on the effects of light on bipolar disorder (Roguski et al., 2024),
- 62     ▪ an intervention study on exposure to bright light during afternoon to early evening on later evening melatonin release in adolescents (Lazar et al., 2024),
- 63     ▪ an observational study on the wearing compliance of personal light exposure (Stefani et al., 2024),
- 64     ▪ an observational study on the differences in light exposure and light exposure related behavior between Malaysia and Switzerland (preregistration in progress),
- 65     ▪ an intervention study on sex and seasonal changes in human melatonin suppression and alerting response to moderate light (publication in progress),
- 66     ▪ an observational study on light exposure, sleep, and circadian rhythms in hospital shift workers (publication in progress).

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