

Distinct circadian temperature rhythm patterns in critical illness myopathy: secondary analysis of two prospective trials

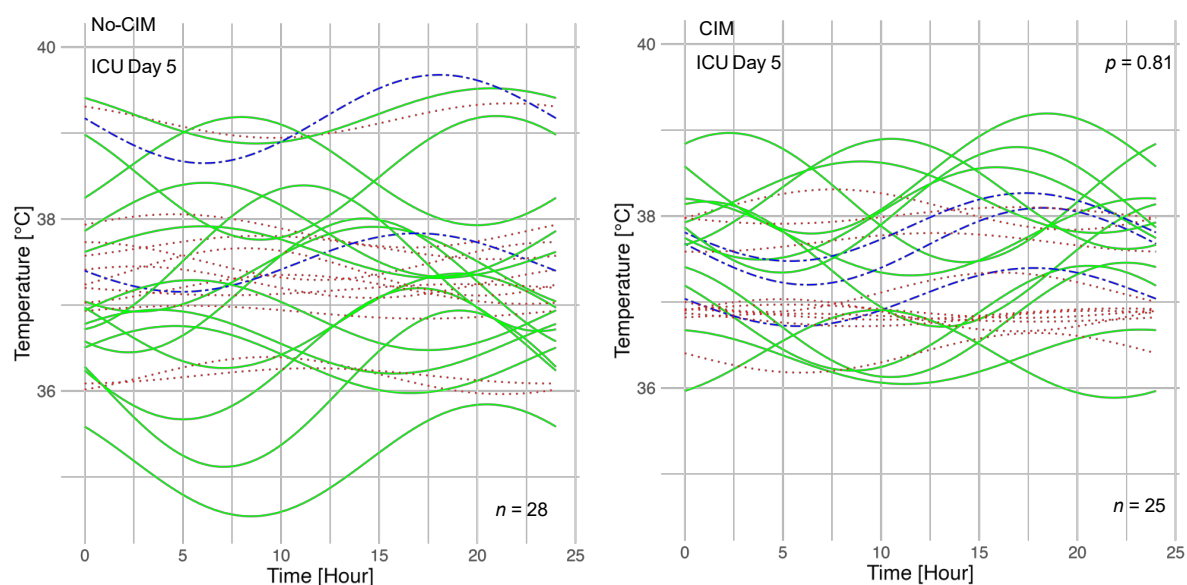
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Additional File 1

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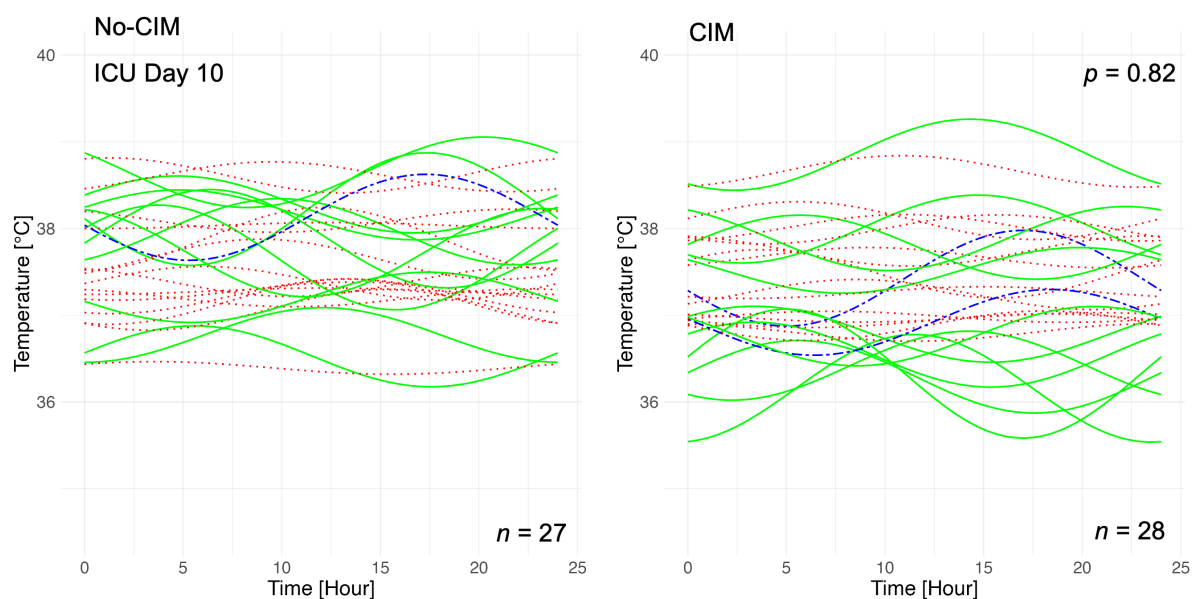


Fig. S1 Body temperature rhythms (a) ICU day 5 and (b) ICU day 10. Cosinor analysis in no-CIM and CIM patients, classified relative to the healthy bed rest group by *CompareRhythms*: orange *loss* (amplitude loss), green *change* (phase shift), blue *same* amplitude and phase. In the initial phase of critical illness, both groups showed temperature rhythms with phase-shifts and amplitude loss. Rhythm class distribution in no-CIM vs. CIM groups on ICU day 5 and ICU day 10, Chi-square test. Data were presented at *ESICM Lives 2024*, the Abstract has been published (1).

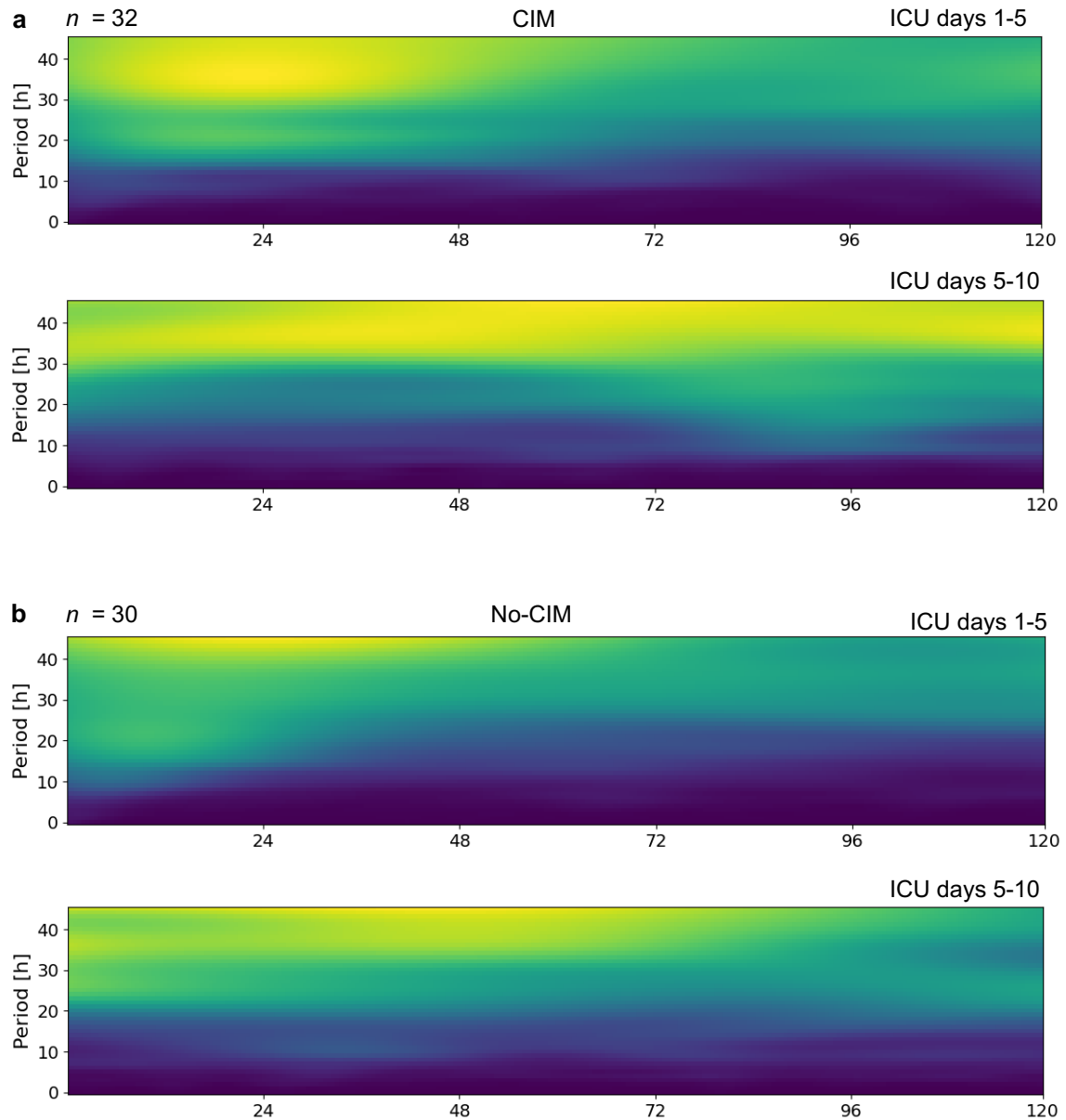


Fig. S2 Wavelet spectrogram ICU days 1-5 and ICU days 5-10 Bright yellow regions indicate average temperature rhythm periods at the given time point during ICU stay with a high relative power. Temperature rhythm period varies as a function of time in the ICU in both, the CIM and no-CIM groups. Early during ICU stay, the periodic signal appears more diffuse, while over time, a strengthening of the 24-hour period is observed in the wavelet power spectrum in both groups (Fig.7). Visual group-level comparison does not suggest a relevant difference between groups.

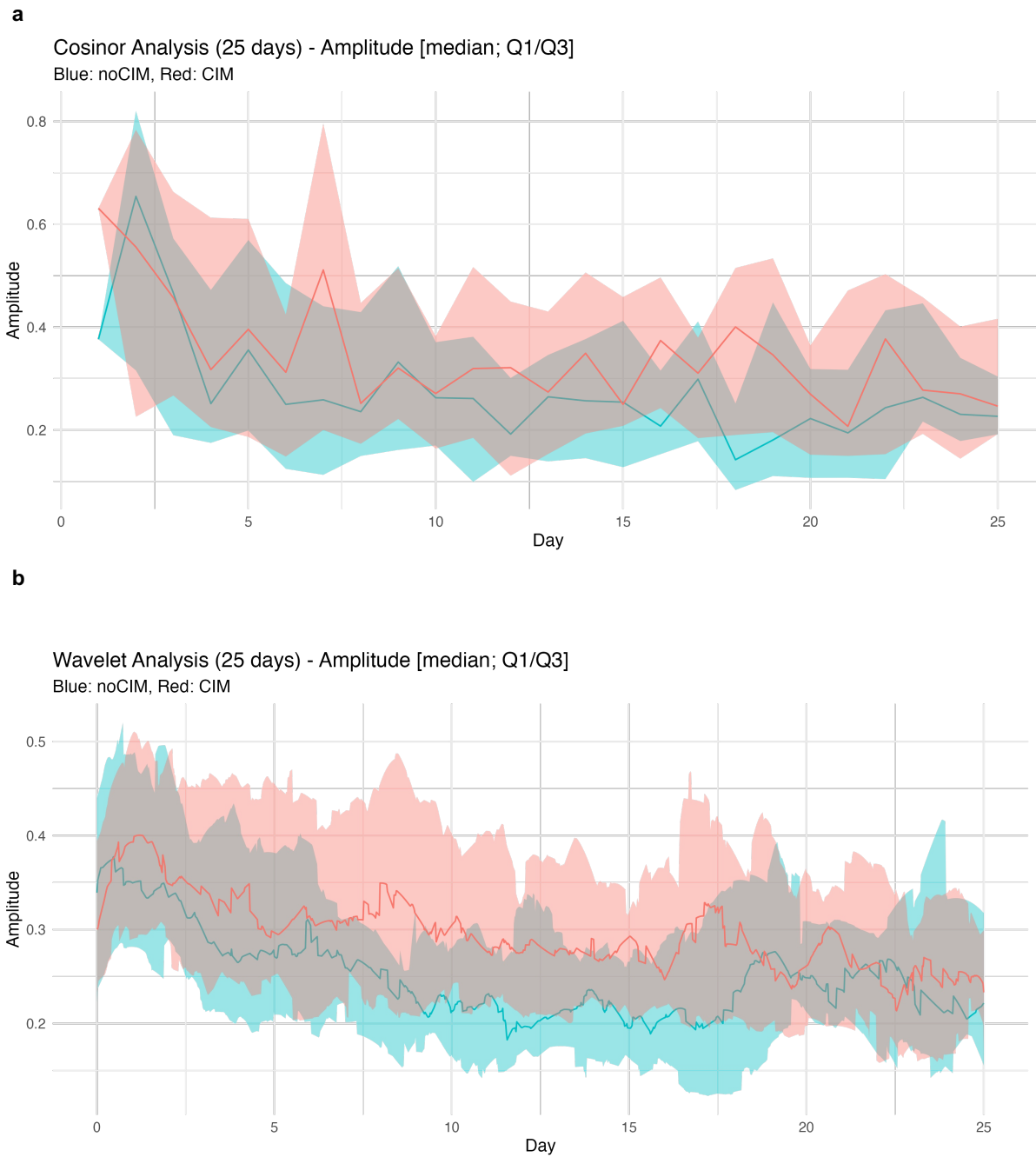


Fig. S3 Comparison of temperature rhythm amplitude in CIM and no-CIM groups using different analysis methods. (a) Cosinor analysis: Temperature rhythm amplitude is consistently higher in CIM compared to no-CIM patients on most ICU days between day 1 and 25. **(b) Wavelet analysis:** Temperature rhythm amplitude is likewise consistently higher in CIM compared to no-CIM patients across the majority of ICU days between day 1 and 25.

Table S1 Multivariable linear regression model for temperature rhythm mesor*n*=62 patients (no-CIM *n*=30, CIM *n*=32)

Response Variable	Variable	β	95% CI of β	β standardized	<i>p</i>	R^2	R^2 adjusted	<i>p</i> model
Mesor	Intercept	36.95	[36.86;37.03]		<0.01	0.48	0.47	<0.01
	CIM	-0.10	[-0.15; -0.05]	-0.08	<0.01			
	Age	0.00	[0.00/0.00]	0.01	0.71			
	Sex	0.05	[0.00/0.10]	0.04	0.06			
	Fever	0.92	[0.87/0.96]	0.68	<0.01			

B regression coefficient, β standardized regression coefficient, *CI* Confidence interval, CIM Critical illness myopathy

Table S2. Multivariable linear regression model for temperature rhythm amplitude*n*=62 patients (no-CIM *n*=30, CIM *n*=32)

Response Variable	Variable	β	95% CI of β	β standardized	<i>p</i>	R^2	R^2 adjusted	<i>p</i> model
Amplitude	Intercept	0.29	[0.25;0.34]		<0.01	0.09	0.08	<0.01
	CIM	0.07	[0.04;0.1]	0.13	<0.01			
	Age	-0.001	[-0.002/0.00]	-0.06	0.02			
	Sex	-0.02	[-0.05; 0.01]	-0.04	0.11			
	Fever	0.14	[0.11;0.16]	0.26	<0.01			

B regression coefficient, β standardized regression coefficient, *CI* Confidence interval, CIM Critical illness myopathy

Table S3 Interindividual temperature rhythm variability

Rhythm parameter	ICU Days	CIM	No-CIM	<i>p</i> -value
Amplitude variability	1-5	0.44 [0.26]	0.44 [0.33]	0.515
Mesor variability	1-5	37.25 [0.90]	37.13 [1.06]	0.735
Time of nadir variability	1-5	11.41 [5.87]	12.92 [6.14]	0.765
Amplitude variability	6-10	0.33 [0.22]	0.29 [0.27]	0.191
Mesor variability	6-10	37.29 [0.68]	37.42 [1.13]	0.617
Time of nadir variability	6-10	10.98 [5.86]	9.80 [3.64]	0.788
Amplitude variability	Last 5	0.30 [0.28]	0.21 [0.10]	0.002
Mesor variability	Last 5	37.23 [0.80]	37.26 [0.31]	0.118
Time of nadir variability	Last 5	9.56 [7.35]	8.89 [8.19]	0.902

*Results in Median [IQR]; last ICU days refers to last days with available temperature data
Brown-Forsythe Test;*

Table S4. Excluded temperature values basic statistics

	CIM	No-CIM
Patients (<i>n</i>)	32	30
Overall temperature values per group (<i>n</i>)	50553	41005
Excluded temperature values <34°C (<i>n</i> and % of overall temperature values)	515 (1.0%) mean value: 30.9 °C	270 (0.7%) mean value: 30.1 °C
Excluded temperature values >41°C (<i>n</i> and % of overall temperature values, mean value)	4 (0.01%) mean: 41.5 °C	6 (0.01%) mean: 41.4 °C
Temperature values outliers excluded (<i>n</i> and %))	50034 (99.0%)	40729 (99.3%)
Overall ICU days per group* (<i>n</i>)	1226	938
Excluded ICU days with data gaps >4h (<i>n</i> and % of overall ICU days)	325 (26.5%)	210 (22.3%)
Overall days included in Cosinor Analysis per group (<i>n</i> and % of overall ICU days)	901 (73.5%)	728 (77.7%)

* measurement outlier removal reduced both CIM and No-CIM by two days, as these consisted entirely of outlying values.

Table S5. Temperature rhythm parameters based on two-component Cosinor analysis

Rhythm parameter	CIM	No-CIM	p-value
R ²	0.74 [0.68, 0.79]	0.71 [0.65, 0.77]	0.25
Amplitude (in °C)	0.42 [0.36, 0.48]	0.33 [0.26, 0.41]	0.02
Mesor (in °C)	37.24 [37.09, 37.4]	37.41 [37.18, 37.7]	0.07
Time of temperature nadir 1 (hh:mm)	7.21 [6.28, 8.35]	7.89 [6.91, 11.18]	0.15
Time of temperature nadir 2 (hh:mm)	17.18 [15.3, 18.59]	17 [15.84, 19.14]	0.67
Missing second component (%)	0.54 [0.47, 0.63]	0.58 [0.5, 0.65]	

Brunner Munzel Test; The two-component Cosinor analysis improved model fit (R²), as expected. However, a second peak was not consistently detectable (58% of no-CIM and 54% of CIM patients showed no second peak). Importantly, the central finding of a higher amplitude in CIM remained unchanged in the two-harmonic Cosinor model.

Table S6. Body temperature data preprocessing and software packages

Filtering and pre-processing	<ul style="list-style-type: none"> Individual measurements below 34 °C were removed since the data below the physiological range were frequently single outliers (details see Table S5) Days with time gaps in body temperature ≥ 4 hours were excluded (details see Table S5) For the healthy bed rest group, 24-hour rectal temperature data after one week of bed rest were resampled to 60-minute intervals for the complete day of measurements using the <i>round_date</i> function of the <i>lubridate</i> package for R (2). The ICU patients temperature measurements were resampled to 60-minute intervals before performing rhythm classification. The <i>compareRhythms</i> package requires regular data points. We decided to perform hourly resampling (60 minutes), as a compromise between preserving clinically meaningful temperature fluctuations and allowing the use of days with fewer measurements. This was performed using the <i>round_date</i> function of the <i>lubridate</i> package for R (2), reducing the temporal resolution by taking the mean of all measurements within each hour. The cut-off value used for the distinction between <i>loss</i> and <i>change</i> categories in the <i>compareRhythms</i> classification was based on the healthy bed rest
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	group, as described in the Methods (main manuscript).
Software for data importing, pre-processing and analysis	<ul style="list-style-type: none"> ▪ Importing, cleaning, and analysis of data have been implemented using RStudio (version 2023.12.0+369) and R (version 4.3.2 (2023-10-31) (3), Python (v.3.11.3) and SciPy (v.1.11.4). ▪ The CosinorPy library (v2.1) for cosinor analysis (4), reticulate package was used to execute Python code within R markdown notebooks. ▪ Rhythm classification was determined for ICU day 5, ICU day 10, and the day before ICU discharge using the <i>compareRhythms</i> R library (5). ▪ Further R packages: broom (6), , flextable (7), ggbeeswarm(8), ggpubr(9), ggstatsplot (10), knitr (11), lubridate(12), nparcomp (13), nparLD (13), patchwork(14), reticulate (15), rmarkdown (16), table1 (17), tidyverse(18)

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