



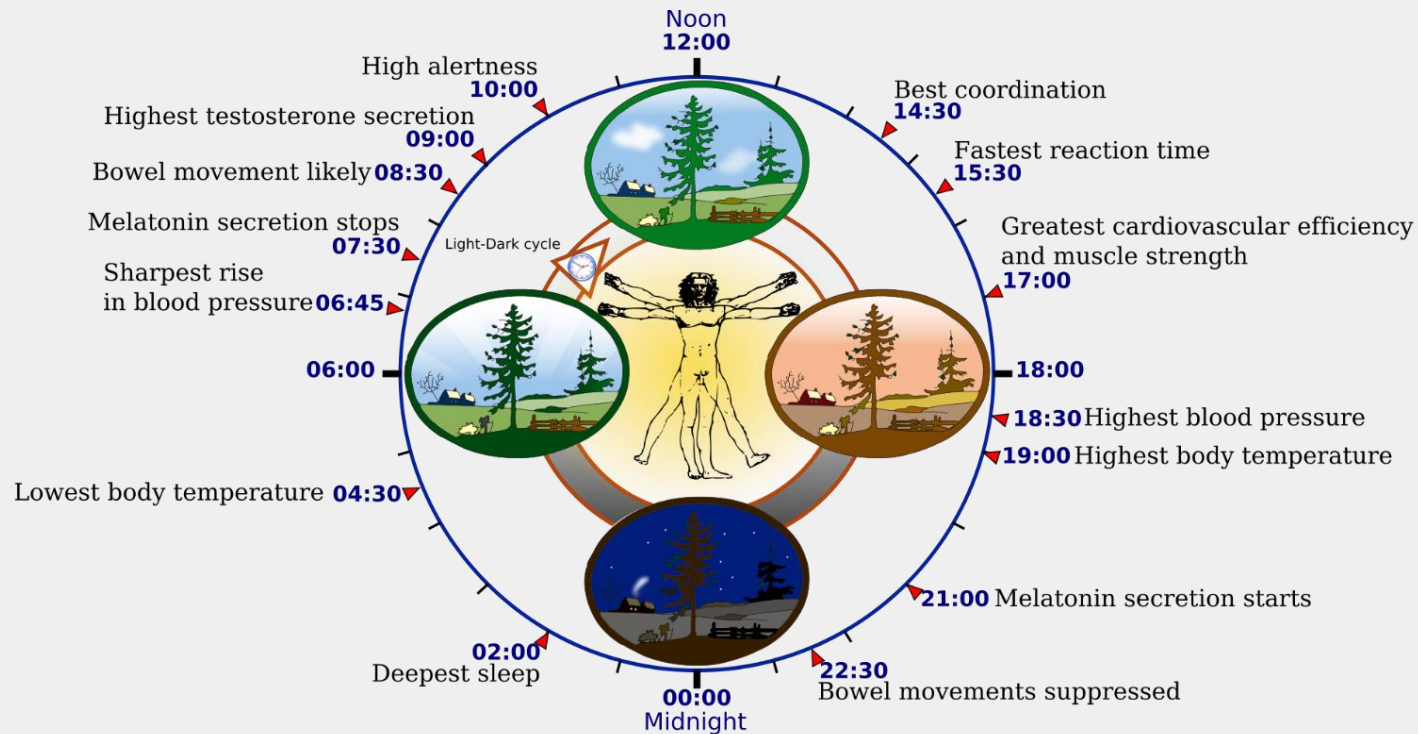
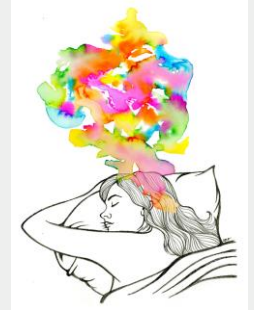
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LightSleepNet: A Lightweight Deep Model for Rapid Sleep Stage Classification with Spectrograms

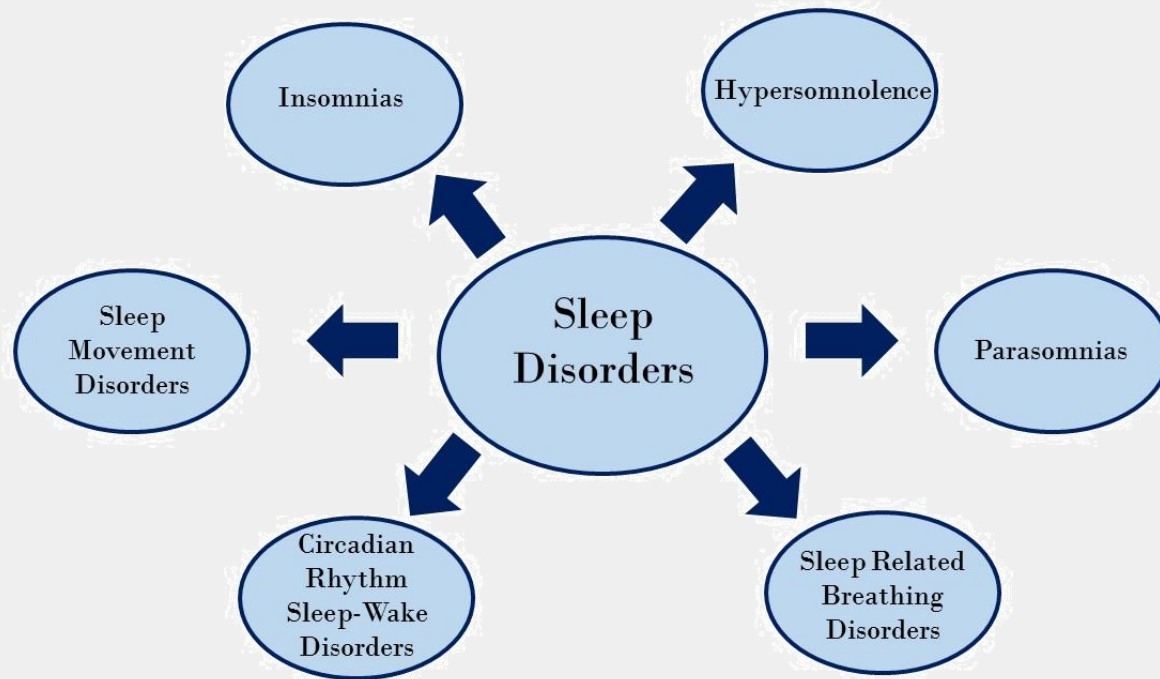
Authors:

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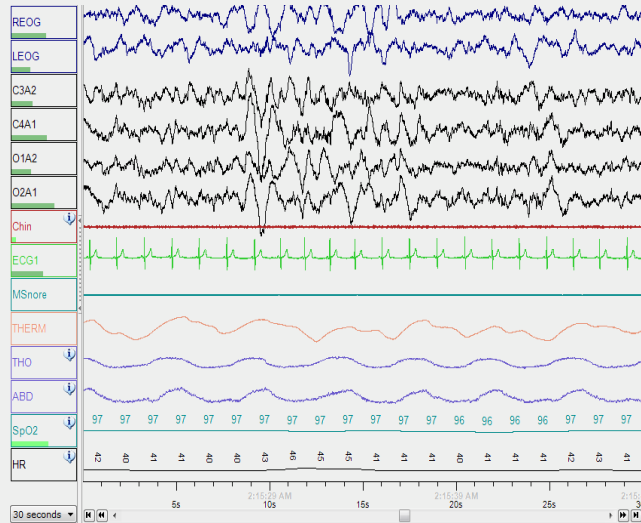
The human biological clock



Sleep disorders



Hypnogram labelled by clinician



Polysomnography (PSG) recordings

Visual inspection

Time-consuming,
Subjective error



W: wake; REM: Rapid Eye Movement; N1-N3 (Non-REM)

Shortcomings

- Time-consuming
- Subjective error
- Well trained (demanding)

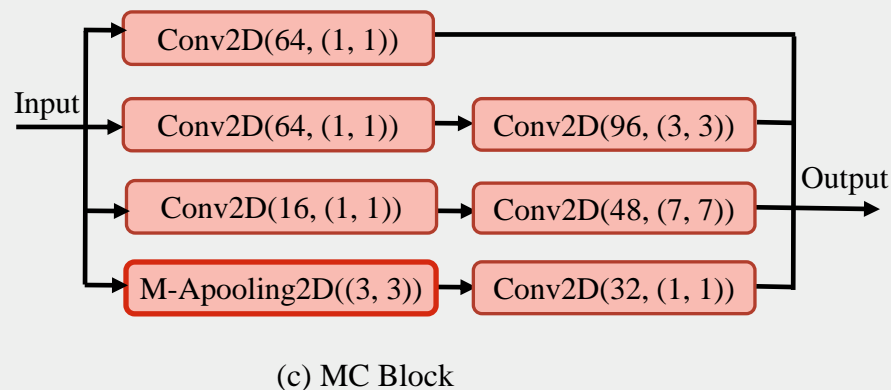
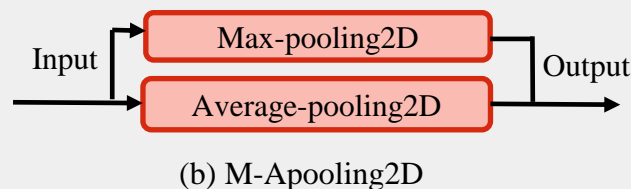
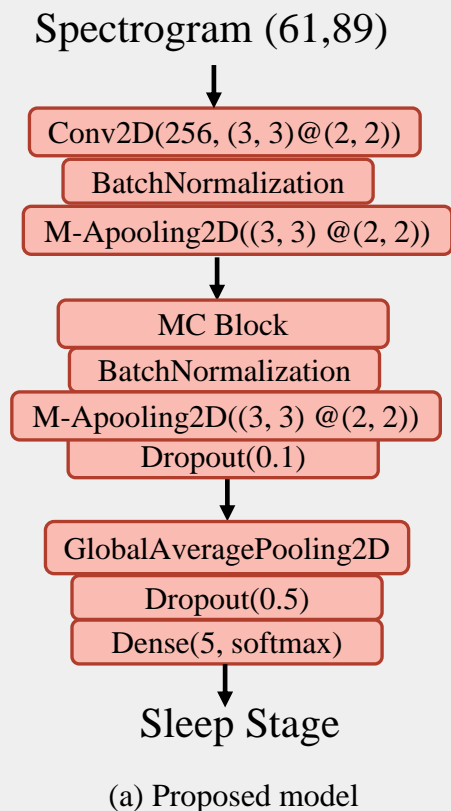
Experimental databases



- Sleep Heart Health Study (SHHS): 100 subjects, EEG C4
- Sleep-EDF Database Expanded (Sleep-EDF, 2018): 78 subjects, EEG Fpz-Cz
- Sleep-EDF Database (Sleep-EDF, 2013): 20 subjects, EEG Fpz-Cz

Dataset	W	N1	N2	N3	REM	Total
SHHS-100	23708	3010	41207	14306	14989	97220
Sleep-EDF	69518	21522	69132	13039	25835	199046
Sleep-EDF-V1	10917	2804	17799	5703	7717	44220

LightSleepNet



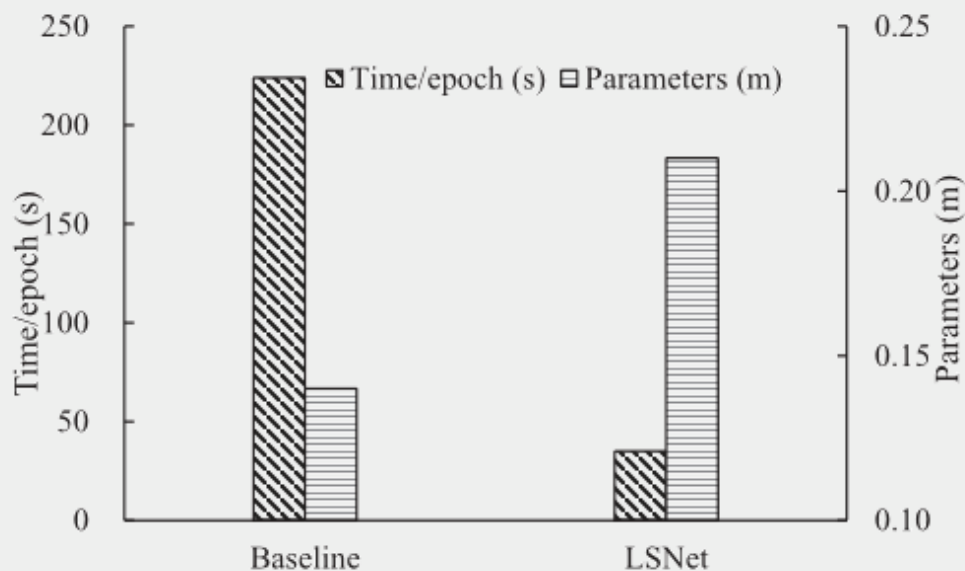
Baseline:

- Same model structure
- The filter size of $N \times N$ is replaced by the filter size of N
- Time series as the input

Experimental results



- Model parameters and computational cost



Baseline: 0.14 million parameters; 224 seconds for one epoch training

LSNet: 0.21 million parameters; 35 seconds for one epoch training

Experimental results



- Performance comparison


Accuracy (ACC), Cohen's kappa coefficient (K)

Study	Dataset	Method	Input channel	Input type	Parameters ($\times 10^6$)	Subjects	ACC(%)	K(%)
Proposed (epoch-wise)	SHHS-100	Deep CNN	C4-A1	Spectrogram	0.2	100	86.7	81.3
Proposed (subject-wise)	SHHS-100	Deep CNN	C4-A1	Spectrogram	0.2	100	85.6	79.4
Sors <i>et al.</i> [11]	SHHS	Deep CNN	C4-A1	Time series	2.2	5728	87	81
Seo <i>et al.</i> [16]	SHHS	CNN + LSTM	C4-A1	Time series	-	5791	86.7	79.8
Zhang <i>et al.</i> [12]	SHHS	CNN + LSTM	2EEG + 2EOG + EMG	Spectrogram	1.3	5793	87	82
Proposed (epoch-wise)	Sleep-EDF	Deep CNN	Fpz-Cz	Spectrogram	0.2	78	83.7	77.5
Proposed (subject-wise)	Sleep-EDF	Deep CNN	Fpz-Cz	Spectrogram	0.2	78	83.4	76.7
Supratak <i>et al.</i> [13]	Sleep-EDF	CNN + LSTM	Fpz-Cz	Time series	1.3	78	83.1	77
Mousavi <i>et al.</i> [15]	Sleep-EDF	CNN + LSTM	Fpz-Cz	Time series	21	78	80.0	73
Proposed (epoch-wise)	Sleep-EDF-v1	Deep CNN	Fpz-Cz	Spectrogram	0.2	20	88.3	84.5
Proposed (subject-wise)	Sleep-EDF-v1	Deep CNN	Fpz-Cz	Spectrogram	0.2	20	86.1	81.0
Supratak <i>et al.</i> [14]	Sleep-EDF-v1	CNN + LSTM	Fpz-Cz	Time series	21	20	82.0	76
Supratak <i>et al.</i> [13]	Sleep-EDF-v1	CNN + LSTM	Fpz-Cz	Time series	1.3	20	85.4	80
Mousavi <i>et al.</i> [15]	Sleep-EDF-v1	CNN + LSTM	Fpz-Cz	Time series	2.6	20	84.3	79



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