

DeepSleep

A Ballistocardiographic-based Deep Learning Approach for Classifying Sleep Stages

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How can a sleep classification system be modelled using Ballistocardiographic (BCG) sensor data, and achieve a performance comparable with Polysomnography (PSG)?

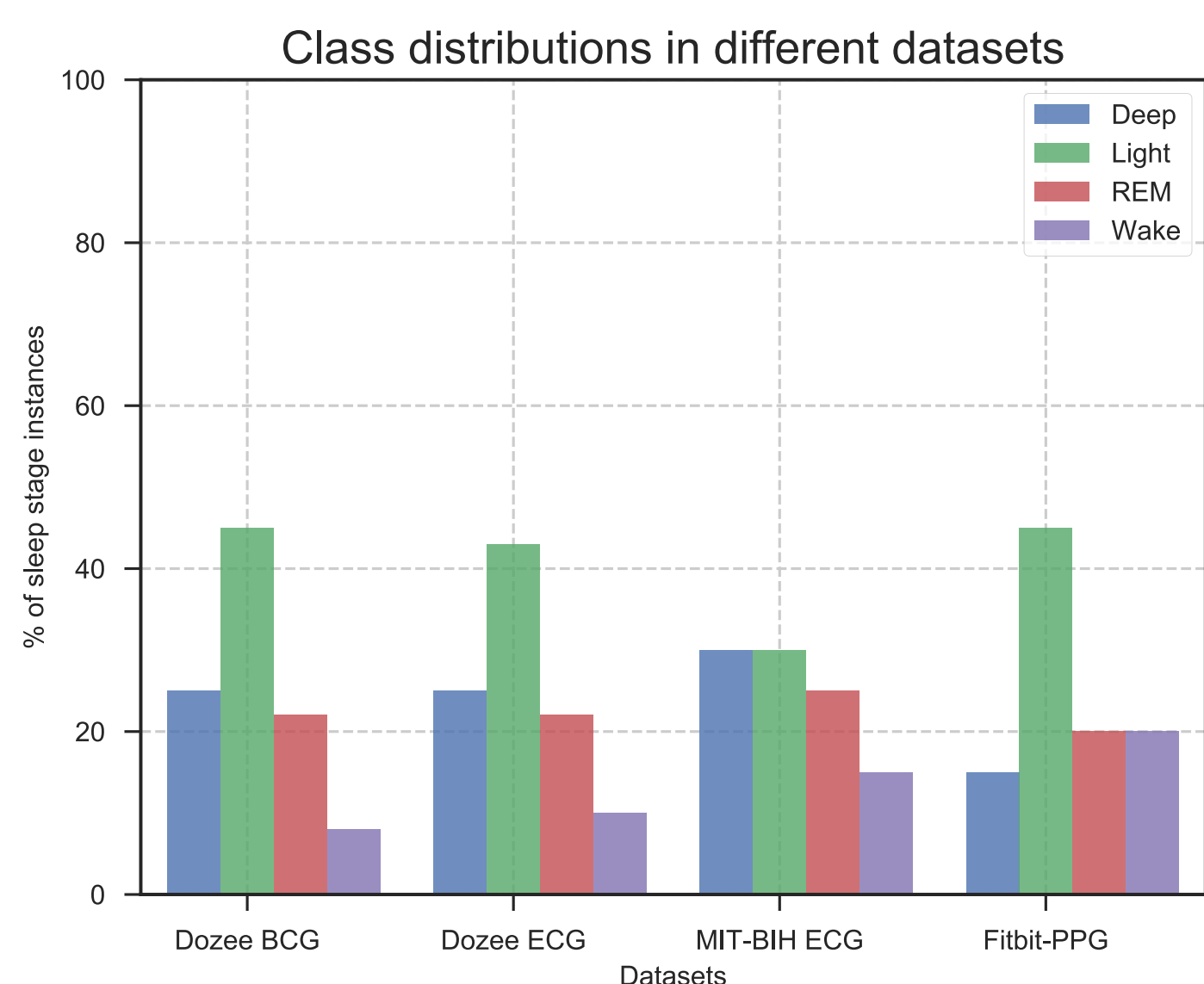
- PSG considered most accurate method for diagnosing sleep-related problems
- However, it is expensive, complex, time-consuming, and uncomfortable for users
- We propose a transfer learning approach using BCG data

Collecting BCG-based heart sensor signals



- 51 recordings, 25 subjects
- Training = 41; Validation = 7; Test = 3
- Ground truth annotated by 2 doctors from NIMHANS [1]
- Cohen's kappa, $k = 0.80$

Dataset distribution



Measuring sleep quality

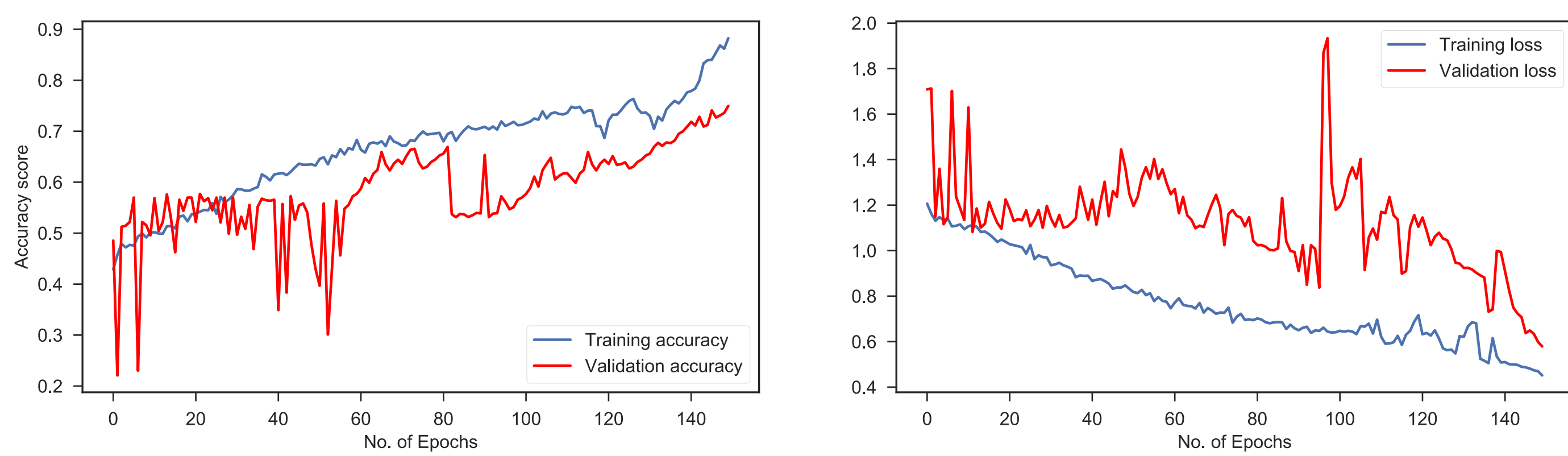
SATED questionnaire for perceived sleep quality

- SATED: Satisfaction, Alertness, Timing, Efficiency, Duration
- 16 subjects
- Scores recorded 1hr and 24 hr after PSG recording

Objective sleep quality

$$SQ = \frac{REM \text{ (min)} + NREM \text{ (min)} - \text{Awakening (min)}}{\text{Total Sleep (min)}}$$

Fine-tuning accuracy and loss



Dataset comparisons

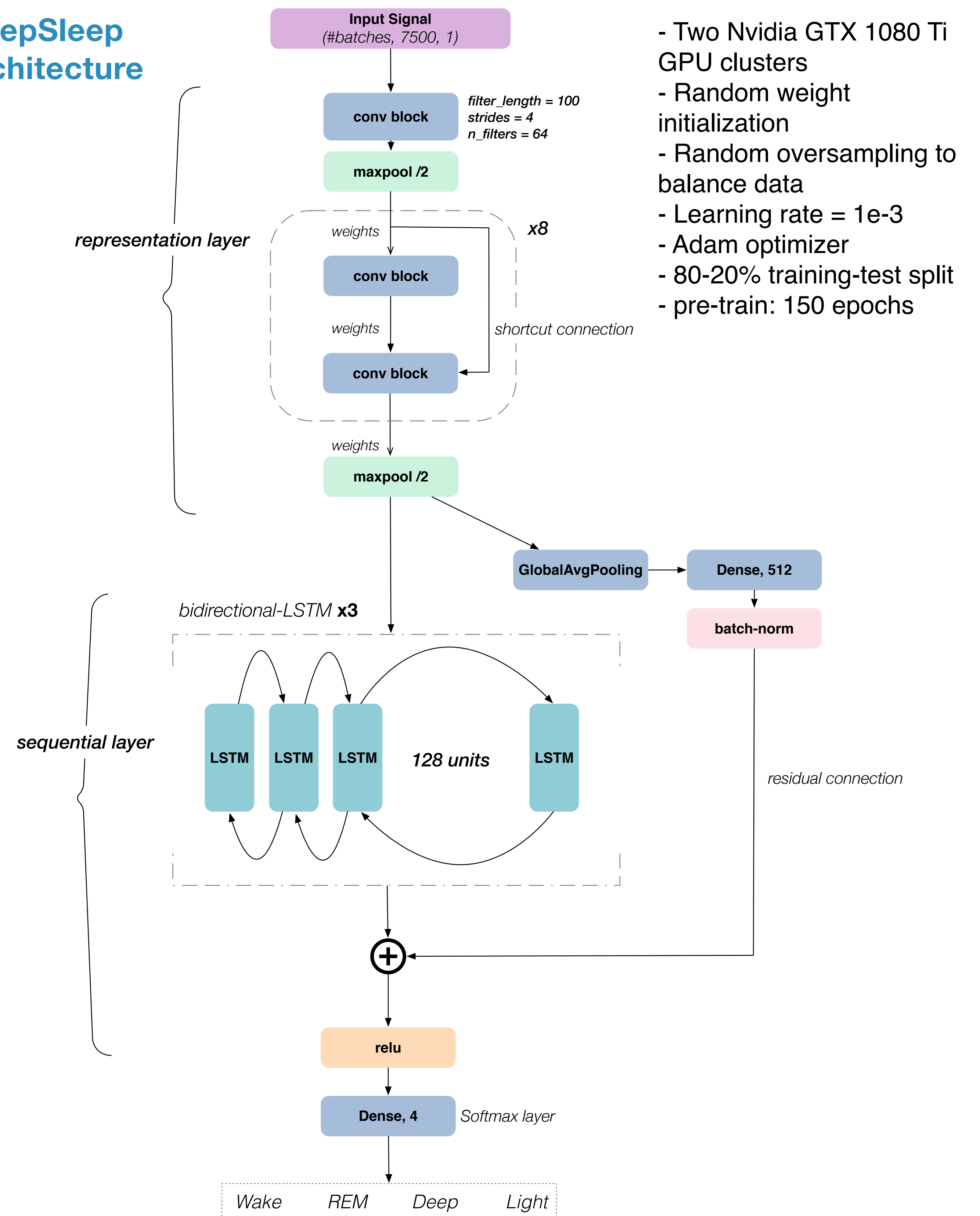
Study	Year	Sensor type	#Features	Classifier	Classes	Accuracy
Långkvist et al. [50]	2012	EEG, EOG, EMG	1	DBN, HMM	W, REM, NREM, L	72%
Samy et al. [72]	2014	BCG	6	KNN, SVM, Naive-Bayes	W, L, REM, Deep (NREM)	72%
Supratak et al. [77]	2017	EEG	1	1D-CNN + LSTM	W, REM, NREM, L	86%
Dong et al. [24]	2018	EEG, EOG	1	LSTM	W, REM, NREM, L	86%
Chambon et al. [13]	2018	EEG, EOG, EMG	1	1D-CNN	W, REM, NREM, L	87%
DeepSleep (proposed)	2018	BCG	1	1D-CNN + bi-LSTM	W, L, REM, Deep (NREM)	74%

Performance comparison between *DeepSleep* model and prior works that perform 4-class classification.

Dataset	Sensor type	#Features	#Recordings	Accuracy
Dozee BCG	BCG	1	51	74%
Dozee ECG	ECG	1	51	77%
MIT-BIH	ECG	1	80	82%
Fitbit-PPG	PPG	1	12	63%

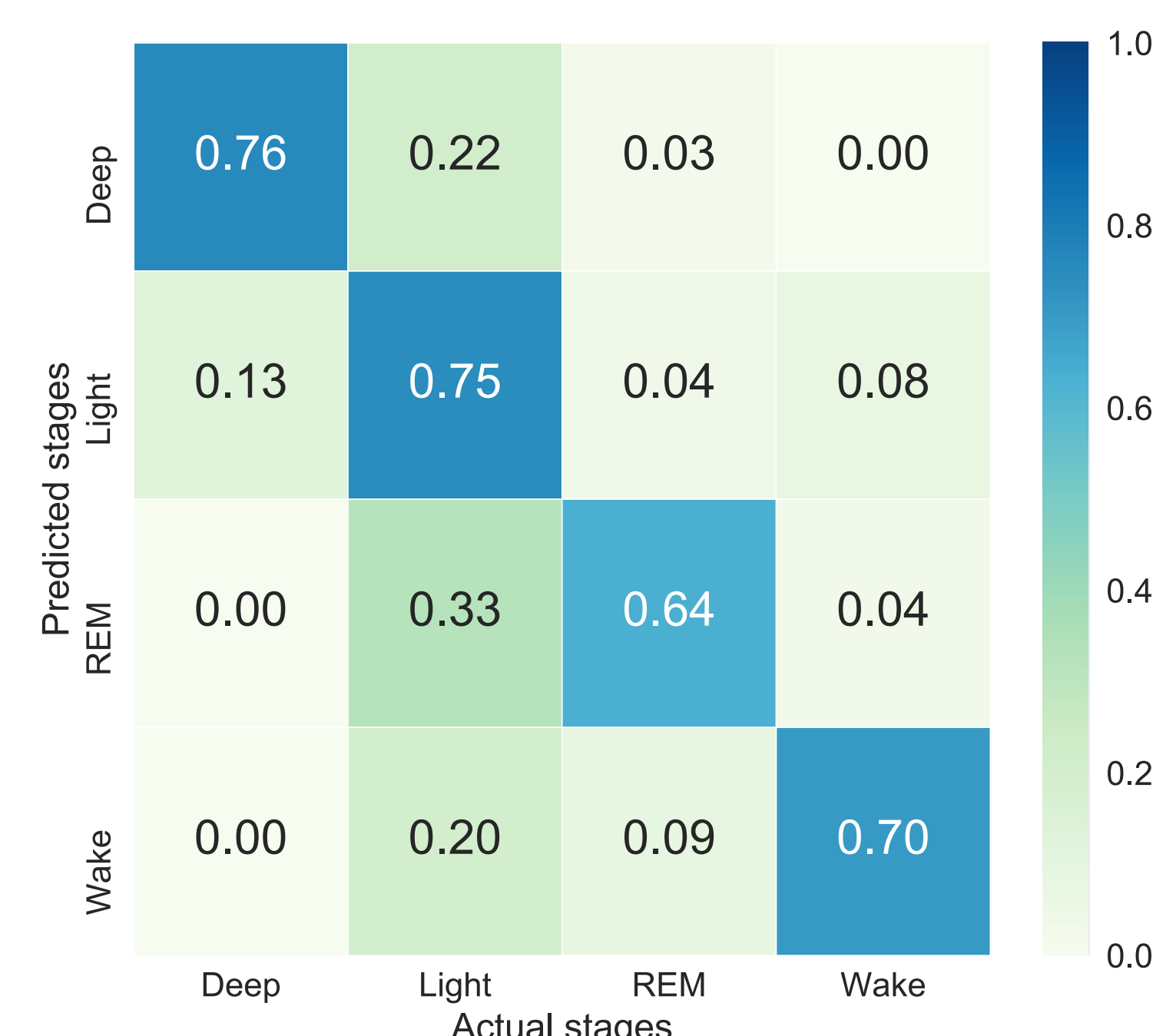
Performance of *DeepSleep* model on different datasets and sensor types

DeepSleep architecture



- Two Nvidia GTX 1080 Ti GPU clusters
- Random weight initialization
- Random oversampling to balance data
- Learning rate = $1e-3$
- Adam optimizer
- 80-20% training-test split
- pre-train: 150 epochs

Sleep classification performance



Conclusion

- Model identifies onset and period of sleep stages
- Differentiates between REM & Deep
- Avg. F1-score: 74%
- Avg. F1-score: 82% on ECG data (transfer learning)
- Positive correlation with PSG ($r = 0.48$) and SATED ($r = 0.43$), whereas $r=0.54$ between SATED and PSG

Future work

- Leave-one-out cross-validation
- Compare with non-NN approaches
- Better oversampling (e.g., Seq2seq)
- Multimodal learning

[1] NIMHANS - National Institute of Mental Health and Sciences

