

Deep Convolutional and Recurrent Neural Networks for Interpretable Analysis of EEG Sleep Stage Scoring

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Sleeping Stages

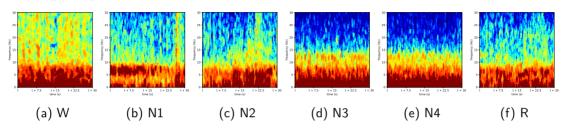


Figure: This figure illustrates a random epoch of the multi-taper spectrum for each sleeping stage. There is high similarity between sleeping stage N3 and N4.

Sleep Stage						
Dist. (in %)	12	7	46	9	6	20

Table: This table summerises the aggregates the distribution of the labels for all 20 Subjects. The distribution of the labels illustrates the sleep stages of subjects during the recordings.



Confusion Matrices

		Predicted					Normalized pred. (in %)						
		W	N1	N2	N3	N4	R	W	N1	N2	N3	N4	R
CNN	W	495	145	29	11	1	20	71	21	4	2	0	3
	N1	25	211	43	0	0	62	7	62	13	0	0	18
	N2	4	51	1313	104	17	68	0	3	84	7	1	4
	N3	0	2	11	164	64	0	0	1	5	68	27	0
	N4	0	0	0	54	91	0	0	0	0	37	63	0
	R	17	80	46	0	0	591	2	11	6	0	0	81
RNN	W	578	39	26	7	1	43	83	6	4	1	0	6
	N1	38	107	64	0	0	132	11	31	19	0	0	39
	N2	8	13	1314	102	28	92	1	1	84	7	2	6
	N3	3	0	18	125	95	0	1	0	7	52	39	0
	N4	0	0	1	60	84	0	0	0	1	41	58	0
	R	19	36	43	0	0	636	3	5	6	0	0	87

Table: Confusion matrices and normalized confusion matrices for the CNN and RNN network.



Bootstrapped Performance Metrics

Study	Precision	Sensitivity	F_1 -score	Accuracy
		71- 71 -72		
RNN	62- 65 -67	63- 66 -69	62- 64 -67	92- 92 -92

Table: **Mean** and corresponding 95% confident values computed by 100.000 bootstrap iterations with replacement.

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Sensitivity Maps

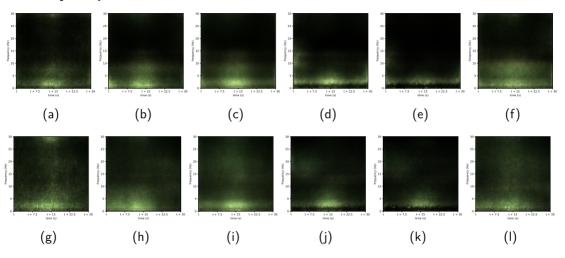


Figure: Illustration of the average sensitivity maps from the CNN (top) and the RNN (bottom) for the two validation subjects.

Conclusion

Future Study



- Successfully re-produced [1] in TF which was one of the objectives, despite using few different approaches.
- Added a LSTM cell to the of the VGGNet 16 network. The of the RNN does out-perform the baseline on average (table 3), despite the RNN does archive better classification sensitivity in the following sleeping stages W and R.
- Improvements in this projects can be archived by merging sleeping stage N3 and N4, then follow the newest regulations.
- Study the effect of stacking multiple LSTM cells and applied the LSTM cells from layers with a lower-level feature representation higher and spectral variance.

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References





A. Vilamala, K. H. Madsen, and L. K. Hansen, "Deep Convolutional Neural Networks for Interpretable Analysis of EEG Sleep Stage Scoring,"

ArXiv e-prints, Oct. 2017.

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