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Deep-learning Your Brain

Classification of movement execution and imagination using EEG signals

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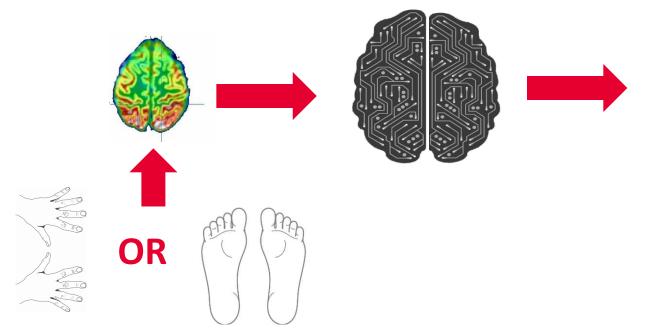
21st May 2019, University of Bern, Advanced Topics in Machine Learning, Prof. Dr. Paolo Favaro



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Classification of movement execution and imagination using EEG signals

What do we aim for?





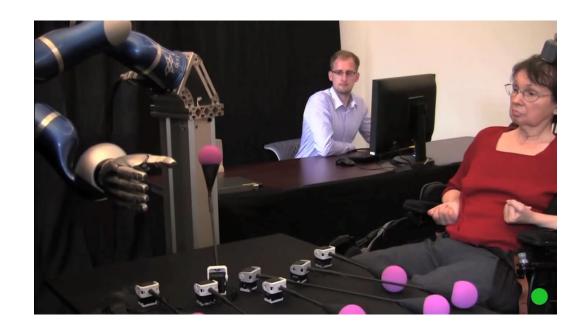






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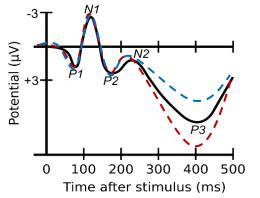
Why to read your mind? Current and future applications

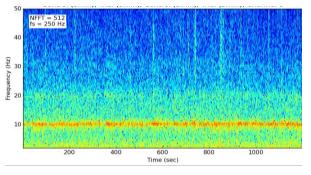


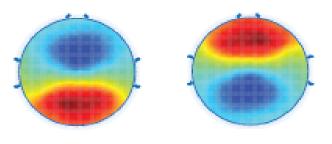




How to understand people's intention? EEG Basics







Time Domain

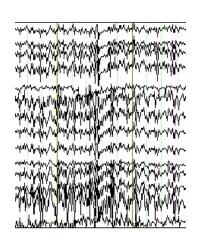
Frequency Domain

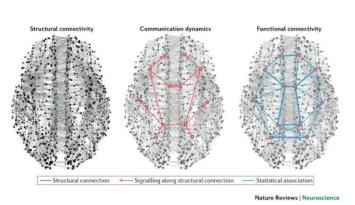
Spatial Domain



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Each person, one world Challenges in EEG analysis





Images:

Imagenet >14 mio. Images

EEG: Physionet BCI 2000109 subjects, 64 electrodes,
7 different tasks

Artefacts

Complex neural processes

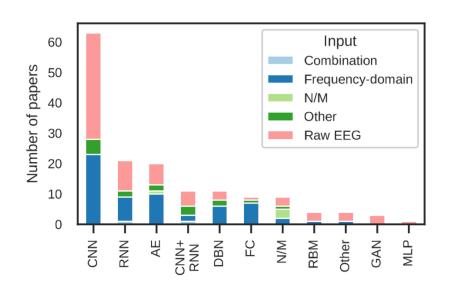
Limited Data



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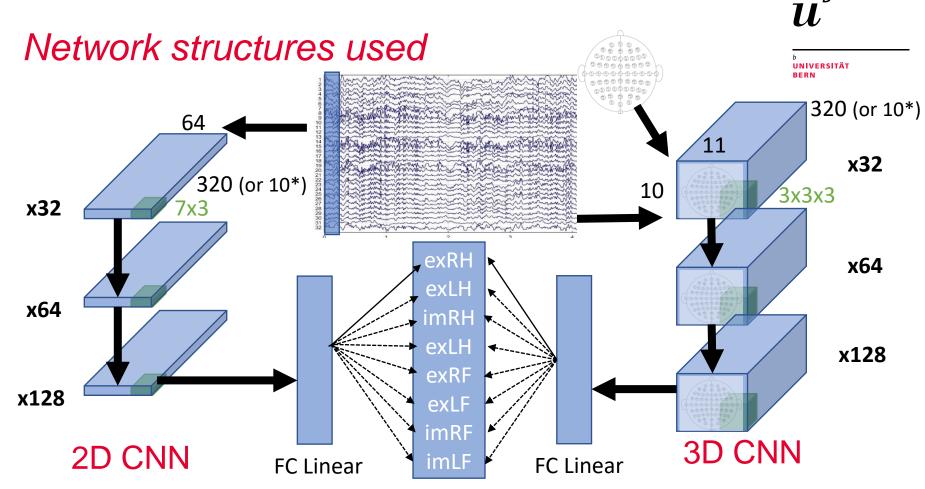
Prior attempts

Classification on EEG Data



Yannick et. al. (2019), arXiv:1901.05498

- Zhang et al. (2018), Advances in Knowledge Discovery and Data Mining
 - 20 subjects, 5 tasks
 - Sliding window (10 points, 50% overlap)
 - 3D-CNN + LSTM + RL
 - 93% accuracy
- Schirrmeister et al. (2017), Hum. Brain Mapp
 - Compared FBCSP, Deep and Shallow CNN in 5 task clasiffication
 - Best accuracy over datasets 93% in Shallow CNN



Adam optimizer, Cross Entropy Loss, *with Time Cropping, 20 Subj. (Zhang,2018)



Results

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Neural Network	Test Accuracy
3D-CNN cropped	77.1%
3D-CNN	28.9%
2D-CNN cropped	73.1%
2D-CNN	32.2%

Class	Test accuracy
Exec. Left Hand	75%
Exec. Right Hand	78%
Imag. Left Hand	75%
Imag. Right Hand	76%
Exec. Both Hand	80%
Exec. Both Feet	76%
Imag. Both Hand	76%
Imag. Both Feet	77%



Discussion and Future Works

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- Time cropping is improving performance. Better local temporal information (high frequency), in exchange of slowly changing global temporal change information.
- 3D-CNN outperforms 2D-CNN due to exploitation of spatial information.



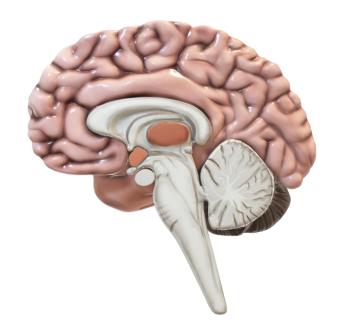
Future Steps:

- Integrate RNN to capture global temporal aspects (Zhang, 2018).
- Using transfer learning to exploit big Phisionet dataset for motor learning experiments with robots.



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Thanks for your attention! Questions...?



Take part in our motor learning experiments;)
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