

Final Report: Brain Dataset

Roman Podolski, Philipp Bergmann, Dominik Irimi, Manuel Nickel,
Christoph Dehner

Technische Universität München

*roman.podolski@tum.de, philipp.bergmann@tum.de, dominik.irimi@tum.de,
manuel.nickel@tum.de, dehner@in.tum.de*

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Overview

1 Dataset

2 Methodology

3 Results

4 Pitfalls

Dataset

Experiment recording human grasp and lift tasks¹

- 12 participants, 5 different types of series recorded each
- EEG: 32 electrodes recorded at 5kHz
- EMG: 5 signals at 4kHz
- kinetic: 36 signals at 500Hz
- objects to grasp with different surface friction/weights (165g - 660g)
- preprocessing: trials provided in windowed format (event timing relative to window)



¹Data source: Luciw, M. D., Jarocka, E. & Edin, B. B. FigShare <http://dx.doi.org/10.6084/m9.figshare.988376> (2014). ↗ ↘ ↙

Experiment

Experimental procedure:

- event/commands signaled visually by LED to participant
- participant starts moving hand to grasp object
- grasp object
- move to target position
- hold position
- move object back to initial position
- hand release object
- move hand back to resting position

Methodology

t-SNE

- expectation
- result

Methodology

Data preparation

- Input normalization to -1/1 range (tanh activation optimization)
- imagine one lifting trial as a single learning sample
 - recordings have different length, therefore equalize it!
 - zero padding → learning in danger of being misguided
 - tail cut → targets fall off, fails to learn sometimes
- separation into sets of 300 data point records length **improves learning.... Why did Smagt tell us to do that??**)
- Subsampling (10Hz) of EMG data within one trial
- consideration of target vector definition: one dim multi class vector vs one-hot-encoding
- data set split: train/valid/test → 0.8/0.1/0.1

Methodology

Recurrent Neural Network

- assuming predictability in human planning → history matters
- network shape:
 - 100 neurons (trial and error showed fewer will fail)
 - 1 layer
 - optimizer: adadelta
 - batch size 50 (limit set by available hardware)
- Bernoulli cross entropy loss at output layer
- Data vector shape
 - input $[300 \times 2428 \times 5] \rightarrow [\text{slice}, \text{trial length}, \text{sensors}]$
 - target $[300 \times 1320 \times 2] \rightarrow [\text{slice}, ???, \text{targets}]$

Methodology

RNN

- Weight initialization from uniform normal dist
- spectral radius
- Important weights
 - skip 150 samples till transient oscillation

LSTM

- usage of breeze library implementation ²
- finally not used because satisfying results achievable by RNN already

²[https://github.com/breeze-no-salt/breeze v0.1 \(2016\)](https://github.com/breeze-no-salt/breeze v0.1 (2016))

Results (1): t-SNE

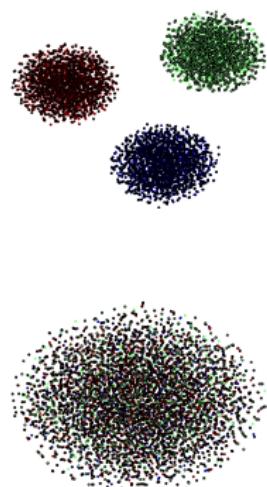
Visualizing high-dimensional EMG and EEG data using t-SNE.

Why?

- Get a *feeling* for the data
- See if notion of *grasping* could possibly be captured by a Feedforward NN

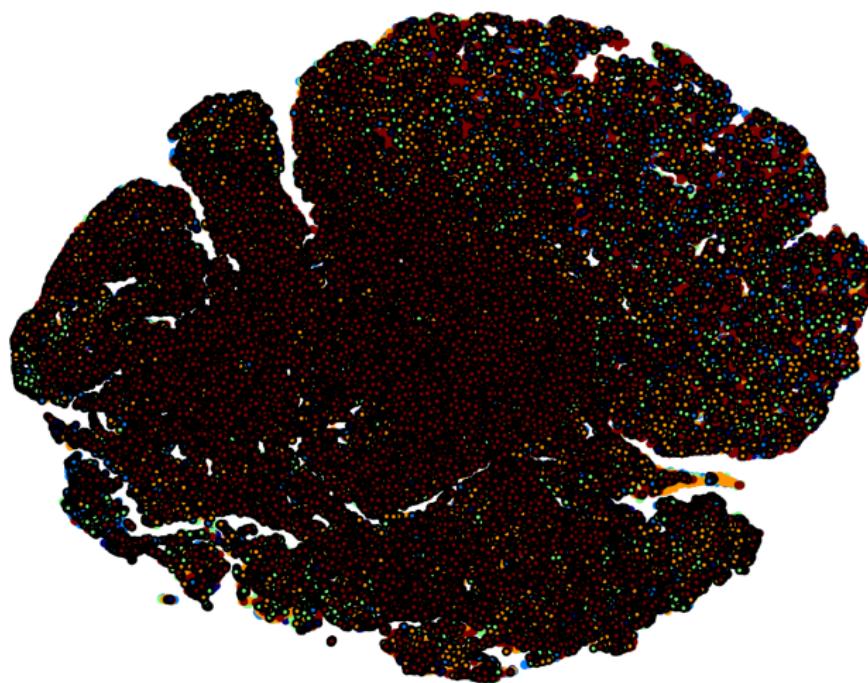
Idea

- *Grasping*-states and *Not Grasping*-states clustered respectively → Use Feedforward NN to classify *Grasping*.
- No apparent structure in the plot → Use Recurrent NN.



Results (1): t-SNE

Visualizing EMG data using t-SNE



Attributes

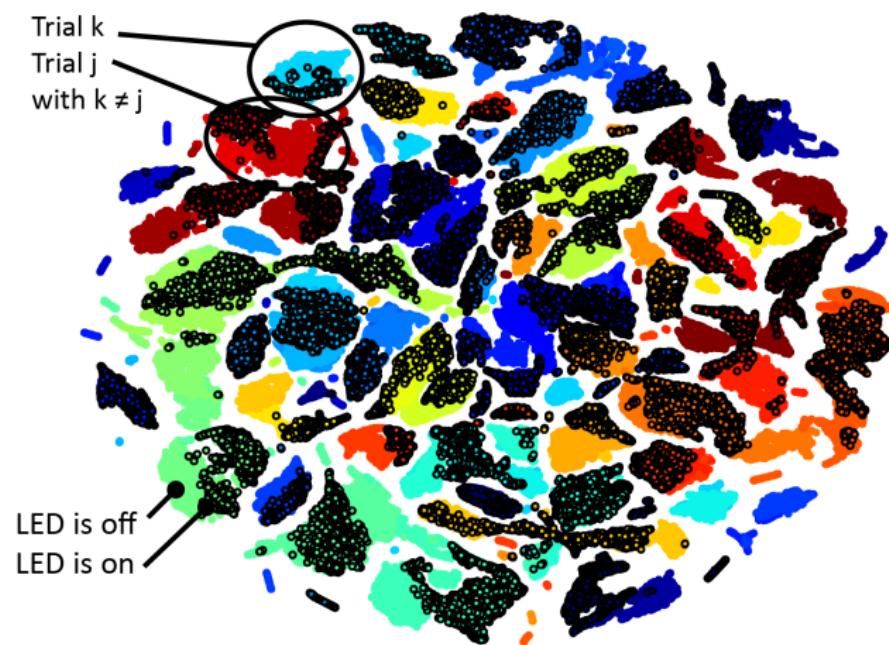
- Data of 34 trials
- Each color stands for one trial
- Black spots mark *Grasping-states*

Conclusion

No apparent structure
→ Feedforward NN
not sufficient, use
Recurrent NN instead.

Results (1): t-SNE

Visualizing EEG data using t-SNE



Attributes

- Data of 34 trials
- Each color stands for one trial
- Black spots mark *Grasping-states*

Conclusion

Trials cluster →
Weird? Use
Feedforward NN to
classify trials?

Results (1): t-SNE

Further analysis of EEG data

Verifying correct function of implementation

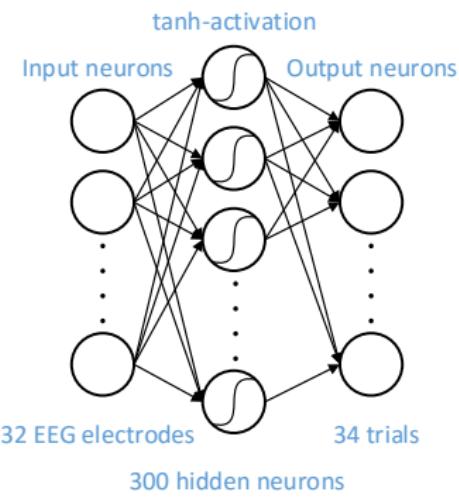
Check code over and over, use reference data sets to make sure the plot is indeed correct → Everything seems fine

If there is some truth to the t-SNE plot, classification of trials should be possible

Classify trials using a Feedforward NN

Feedforward NN with:

- 32 EEG electrodes as input
 - 300 tanh-activated hidden neurons
 - 34 trials as output classes



Results (1): t-SNE

Results of trial classification

Depending on participant and series nearly perfect classification with up to just about 0.01% error on test set.

Even when classifying all participant 1's 296 trials still a good test error of about 15% was achieved.

Conclusion

We conclude that there indeed has to be some kind structure encoding the time of measurement within the EEG data

Results (1): t-SNE

- Figure of t-SNE of EEG data → Separability of trials
- Possible to separate with standard NN
- Figure of t-SNE of EMG data → As expected

Results (2): RNN

- Overview of the targets
- Hand move to target works good.
- Touch phase target also quite ok.
- hand move back target also (partially) sucessful
- Comparison: Training with data of one person vs. data of more participants
- etc.

Pitfalls

- Prediction do not fit to the target borders exactly
- No working early stopping criterion (so far)
- Targets within the lift phase cannot be predicted properly
- etc.

Blocks of Highlighted Text

Block 1

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Block 2

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Block 3

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Multiple Columns

Heading

- ① Statement
- ② Explanation
- ③ Example

Lorem ipsum dolor sit amet,
consectetur adipiscing elit. Integer
lectus nisl, ultricies in feugiat rutrum,
porttitor sit amet augue. Aliquam ut
tortor mauris. Sed volutpat ante
purus, quis accumsan dolor.

Table

Treatments	Response 1	Response 2
Treatment 1	0.0003262	0.562
Treatment 2	0.0015681	0.910
Treatment 3	0.0009271	0.296

Table: Table caption

Theorem

Theorem (Mass–energy equivalence)

$$E = mc^2$$

Example (Theorem Slide Code)

```
\begin{frame}  
 \frametitle{Theorem}  
 \begin{theorem}[Mass--energy equivalence]  
 $E = mc^2$  
 \end{theorem}  
 \end{frame}
```

Figure

Uncomment the code on this slide to include your own image from the same directory as the template .TeX file.

Citation

An example of the \cite command to cite within the presentation:

This statement requires citation [Smith, 2012].

References



John Smith (2012)

Title of the publication

Journal Name 12(3), 45 – 678.

The End