

A visual Brain-Computer Interface for gaze-free communication

Arne Van Den Kerchove

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The Locked-in Syndrome: a functioning mind trapped in a paralyzed body

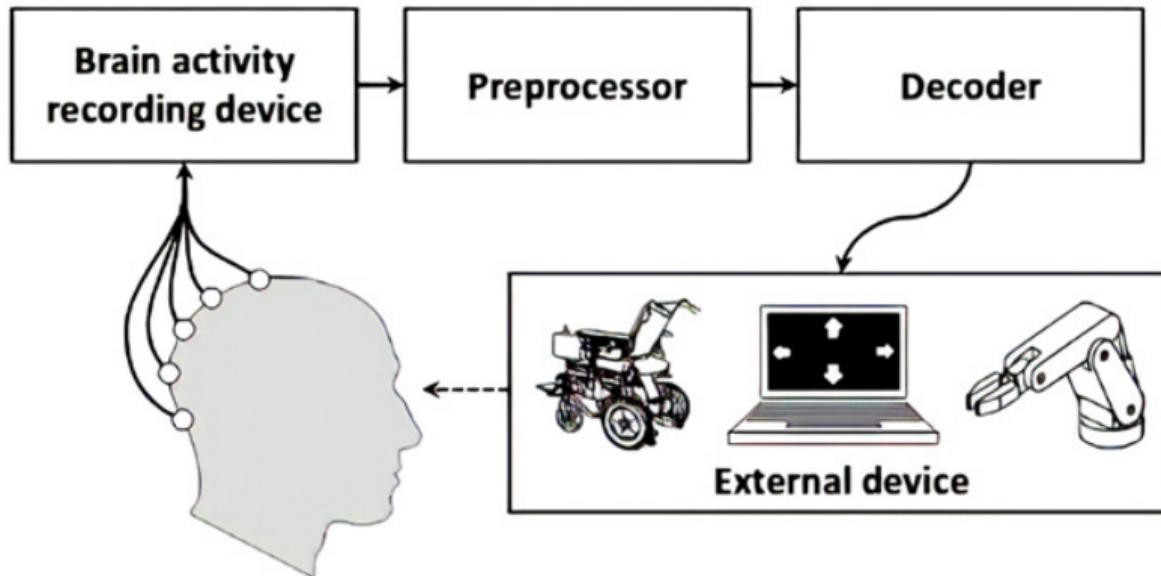


Severe paralysis leads to
Locked-in Syndrome, due to

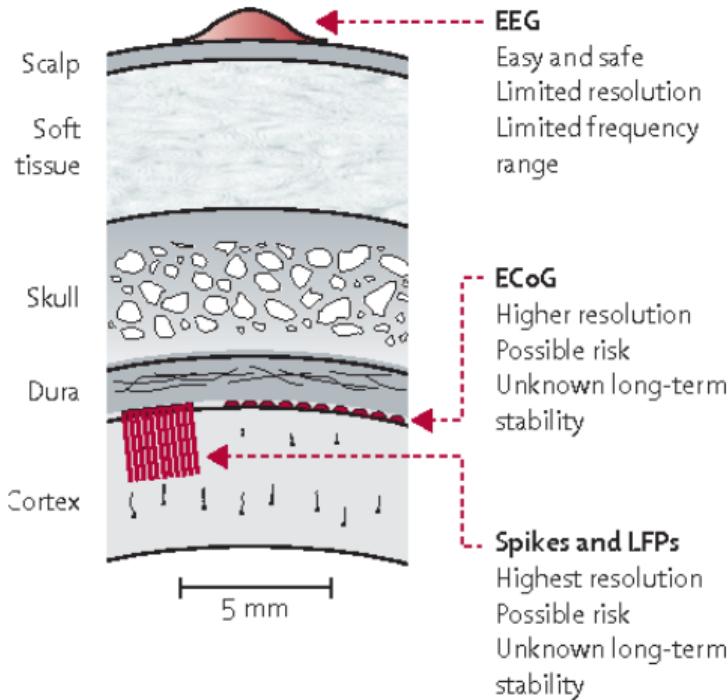
- ▶ Stroke
- ▶ Traumatic brain injury
- ▶ Neurodegenerative diseases
- ▶ ...

Communication requires
assistive technology

A solution: the Brain-Computer Interface (BCI)



Recording the brain activity

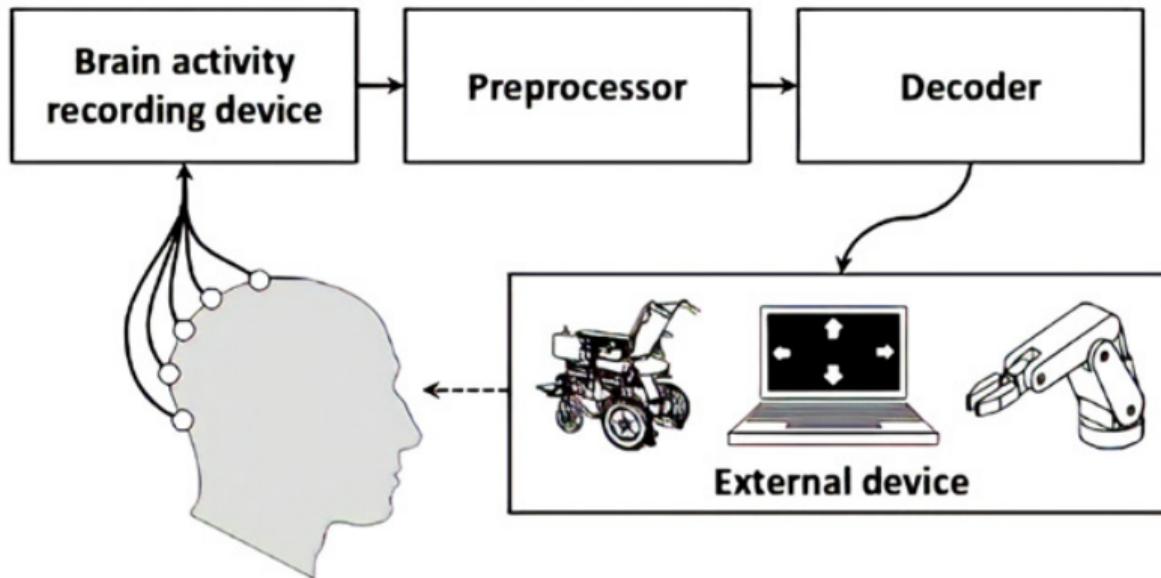


EEG measures the electrical field on the scalp: noise and some brain activity

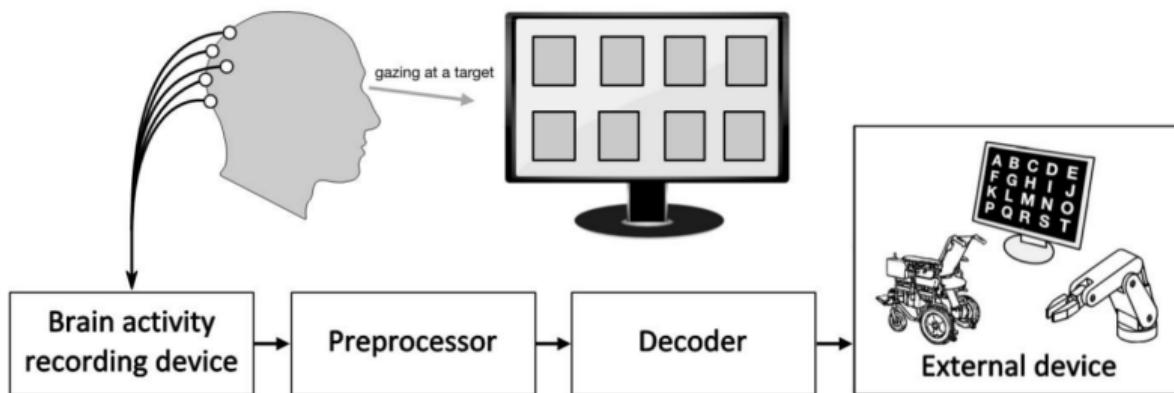
- + Non-invasive
- + Cheap
- + No durability issues
- Limited resolution
- Low signal-to-noise ratio



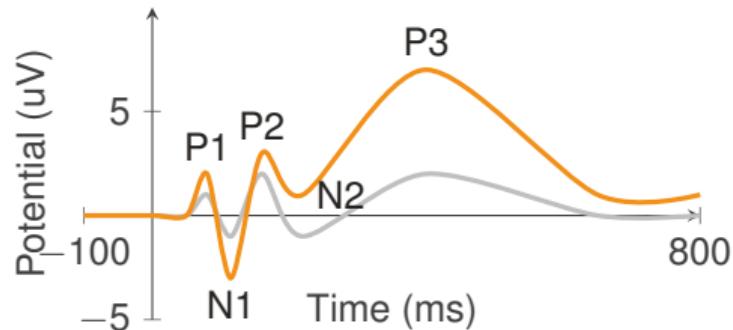
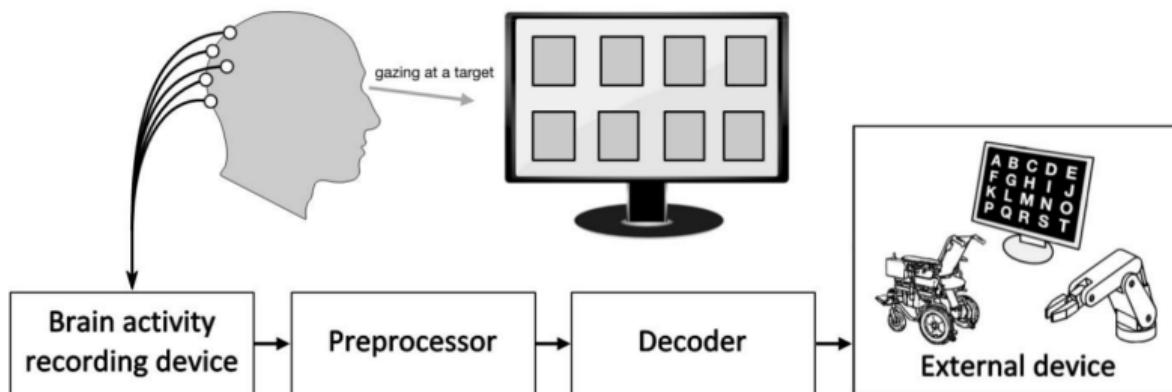
The visual event-related potential paradigm



The visual event-related potential paradigm



The visual event-related potential paradigm



Problem: Eye motor impairment prevents gazing at targets

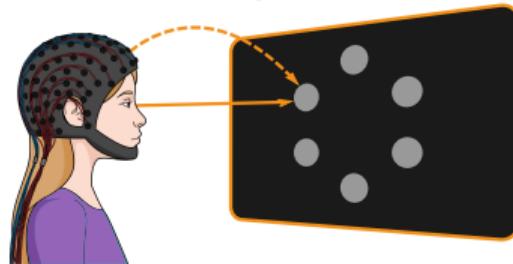
Visual skills related to disease affect BCI operation

Fried-Oken et al., 2020

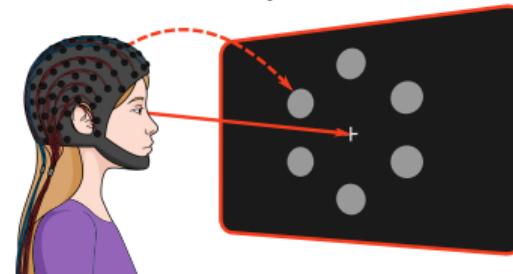
- ▶ Visual fixation
- ▶ Eyelid function
- ▶ Ocular motility
- ▶ Binocular vision
- ▶ Involuntary movement
- ▶ Field of vision



Overt visuospatial attention

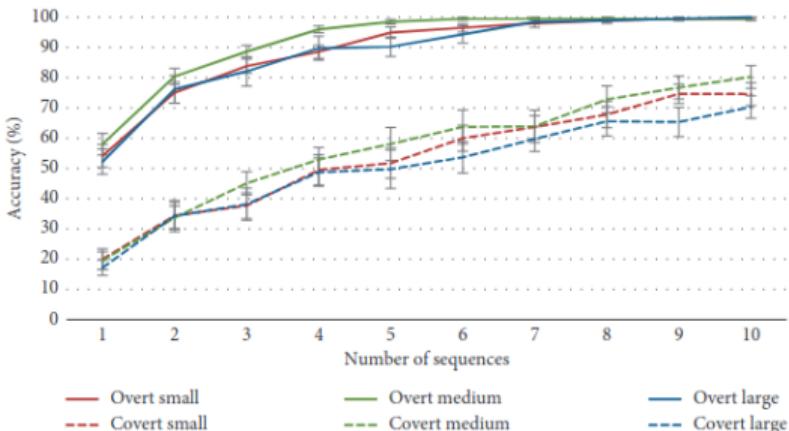


Covert visuospatial attention

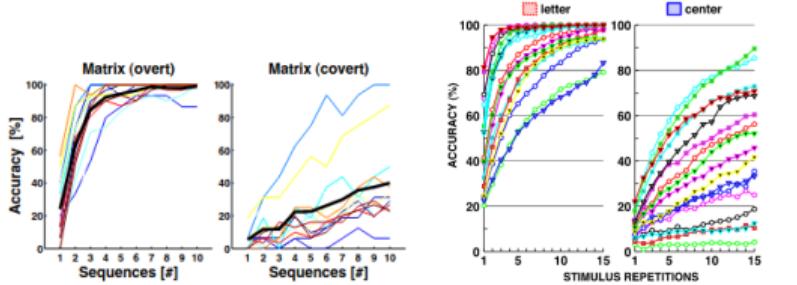


Covert attention decoding performs poorly

- ▶ Multi-target, spatial interfaces
- ▶ Long established, not solved
- ▶ Accuracy at or below usability threshold of 80%



Brunner et al., 2010

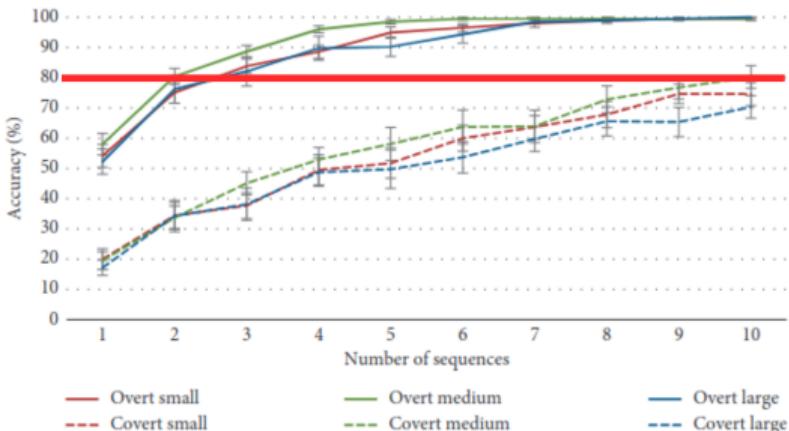


Treder and Blankertz, 2010

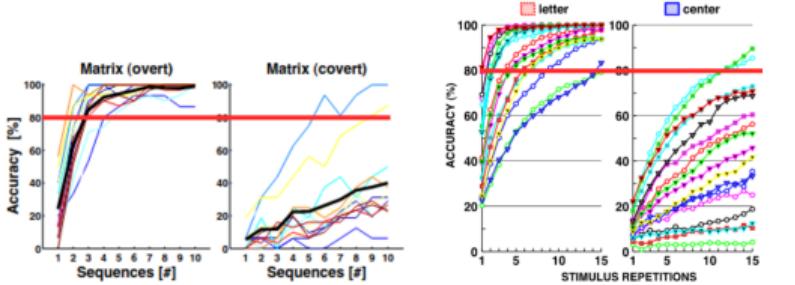
Ron2019

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Treder and Blankertz, 2010

Ron2019

Balance the bandwidth of a **visual ERP BCI** with the needs of individuals with **eye motor impairment** through **improving decoding** of covert attention.

A visual Brain-Computer Interface for gaze-free communication

Introduction

C1: General ERP decoding algorithms

C2: Gaze-independent ERP decoding

C3: End-user case studies

Conclusion

A visual Brain-Computer Interface for gaze-free communication

Introduction

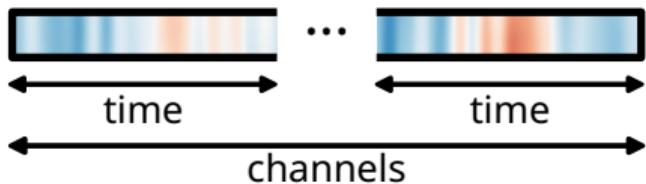
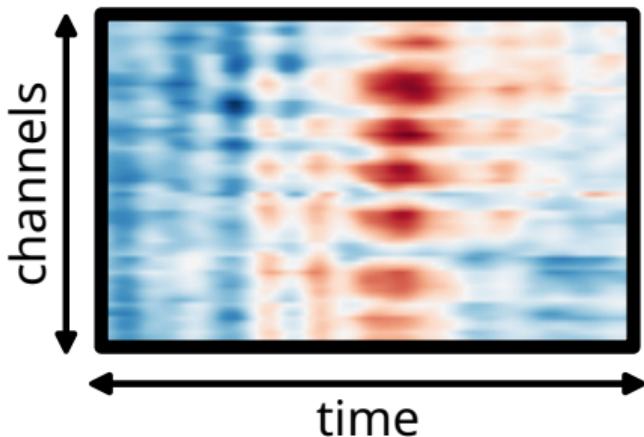
C1: General ERP decoding algorithms

C2: Gaze-independent ERP decoding

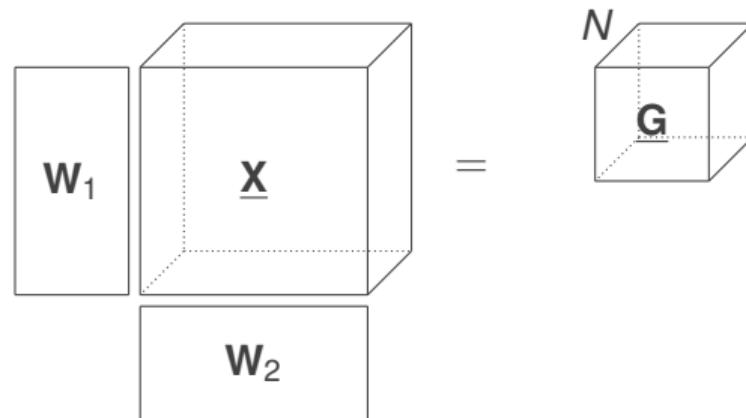
C3: End-user case studies

Conclusion

Exploit channel-time structure of ERP data
for regularization



Higher-order discriminant analysis

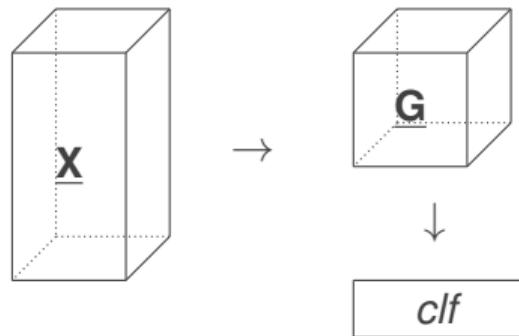


$$\underline{\mathbf{G}} = \underline{\mathbf{X}} \times_1 \mathbf{W}_1 \cdots \times_{K-1} \mathbf{W}_{K-1}$$

s.t. $\underline{\mathbf{G}}$ is maximally **discriminant** between classes.

HODA and the Tucker tensor core model

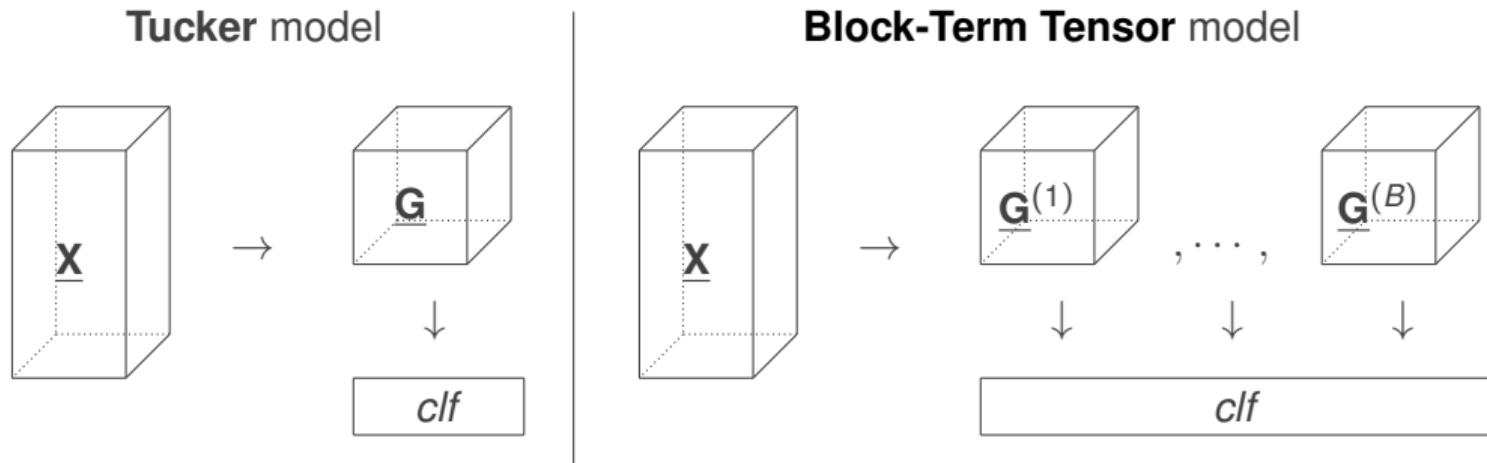
Tucker model



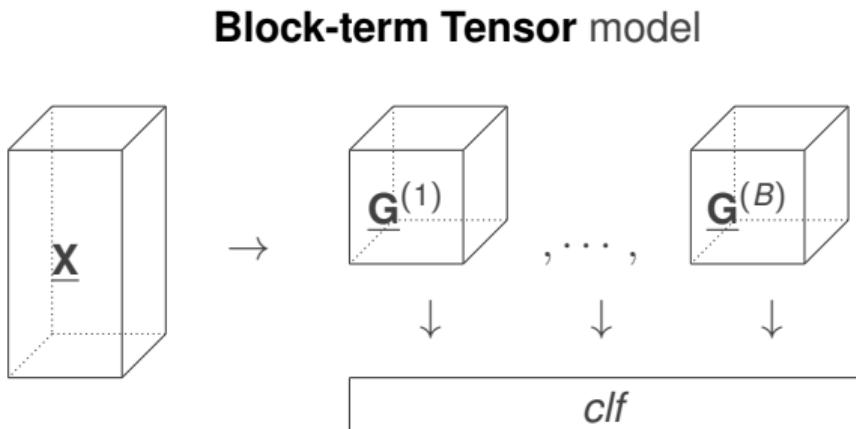
Strong assumptions
on data structure

- + Efficient
- + Regularizing constraints
- + Increased flexibility
- Redundant features
- More parameters to tune

Towards a Block-Term Tensor model



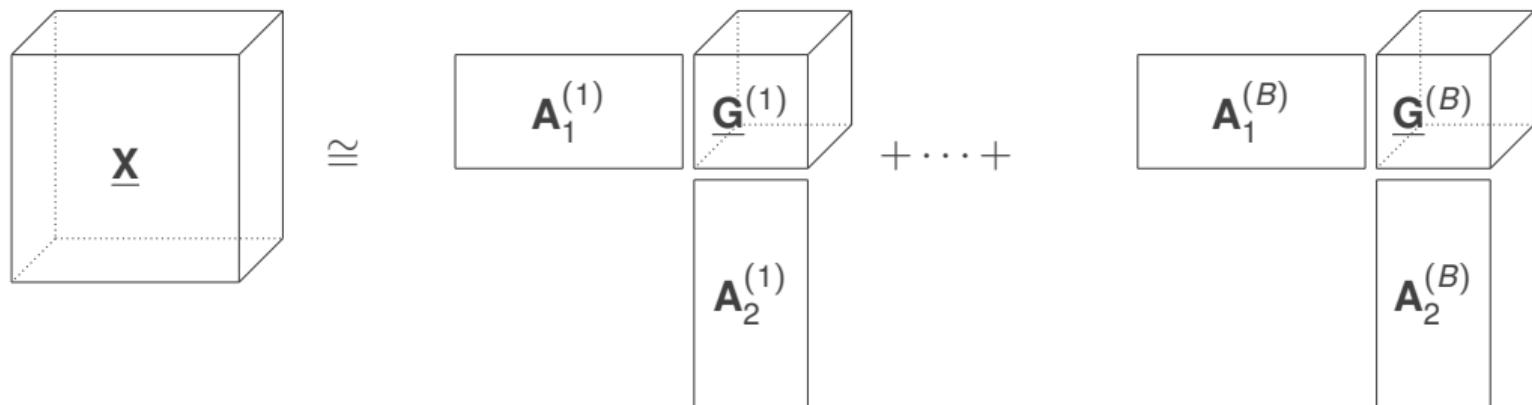
Block-term Tensor Discriminant Analysis



Weaker assumptions
on data structure

- + Efficient
- + Regularizing constraints
- Manual rank selection
- Redundant features
- Cannot model full covariance structure

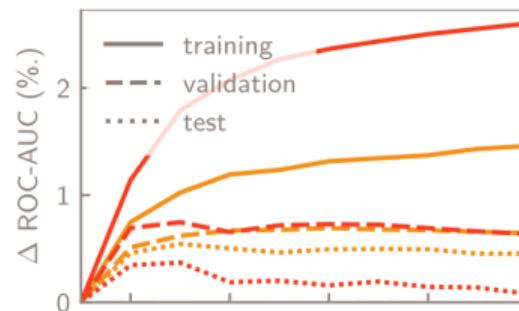
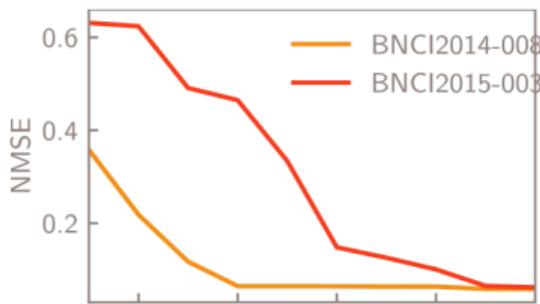
Block-term tensor model through deflation



Deflation scheme:

$$\underline{\mathbf{X}}^{(b+1)} = \underline{\mathbf{X}}^{(b)} - \underline{\mathbf{G}}^{(b)} \times_1 \mathbf{A}_1^{(b)} \cdots \times_K \mathbf{A}_K^{(b)}$$

BTTDA ouptperforms HODA on benchmark datasets



Pipelines	BNCI2014-008	BNCI2015-003
ERPCov+MDM	74.30 ± 9.77	76.79 ± 10.95
ERPCovSVD+MDM	75.42 ± 9.91	76.93 ± 11.26
XDAWN_Cov+MDM	77.62 ± 9.81	83.08 ± 7.55
XDAWN+LDA	82.24 ± 5.26	78.62 ± 7.19
XDAWN_Cov+TS+SVM	85.61 ± 4.43	82.95 ± 8.57
HODA+LDA	83.25 ± 6.25	82.57 ± 7.52
PARAFACDA+LDA	86.19 ± 4.62	84.85 ± 7.93
BTTDA+LDA	86.43 ± 4.51	85.08 ± 7.36

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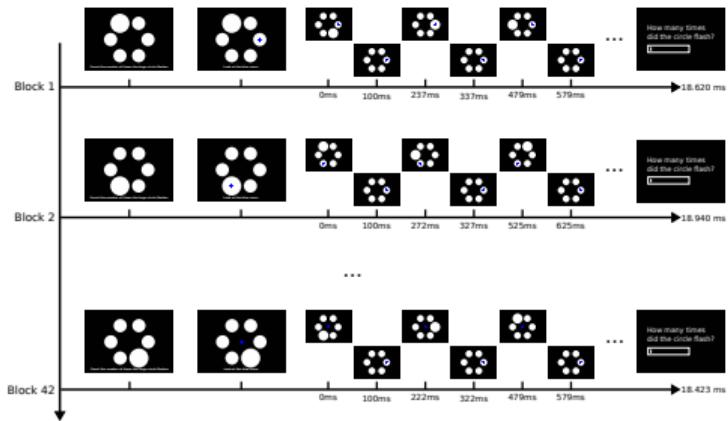
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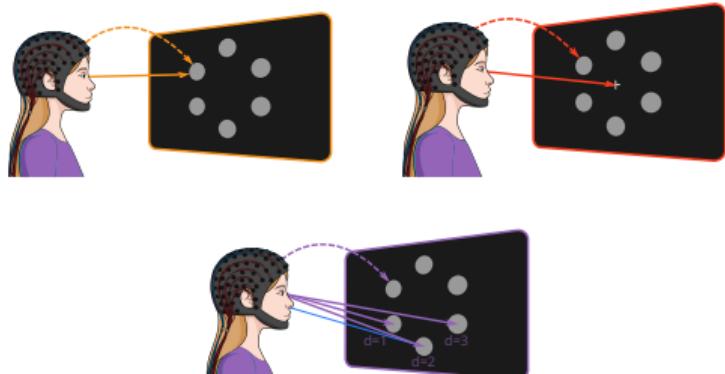
Conclusion

Covert attention ERP data collection



CVSA-ERP dataset

- $N = 15$
- ISI = 300 ± 100 ms
- ± 11.25 h recoded stimulation,
- 3 conditions based on gaze and attention cue

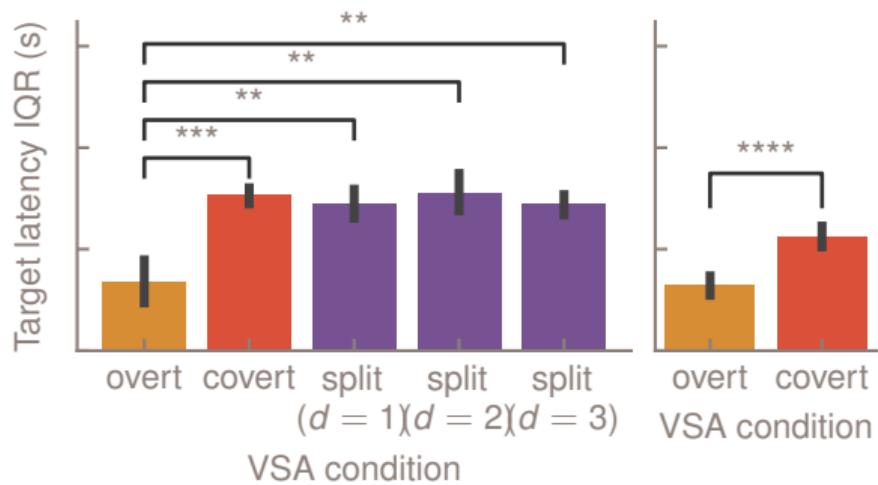
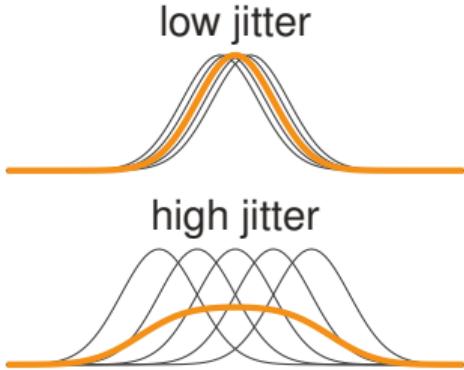


Grand average ERPs

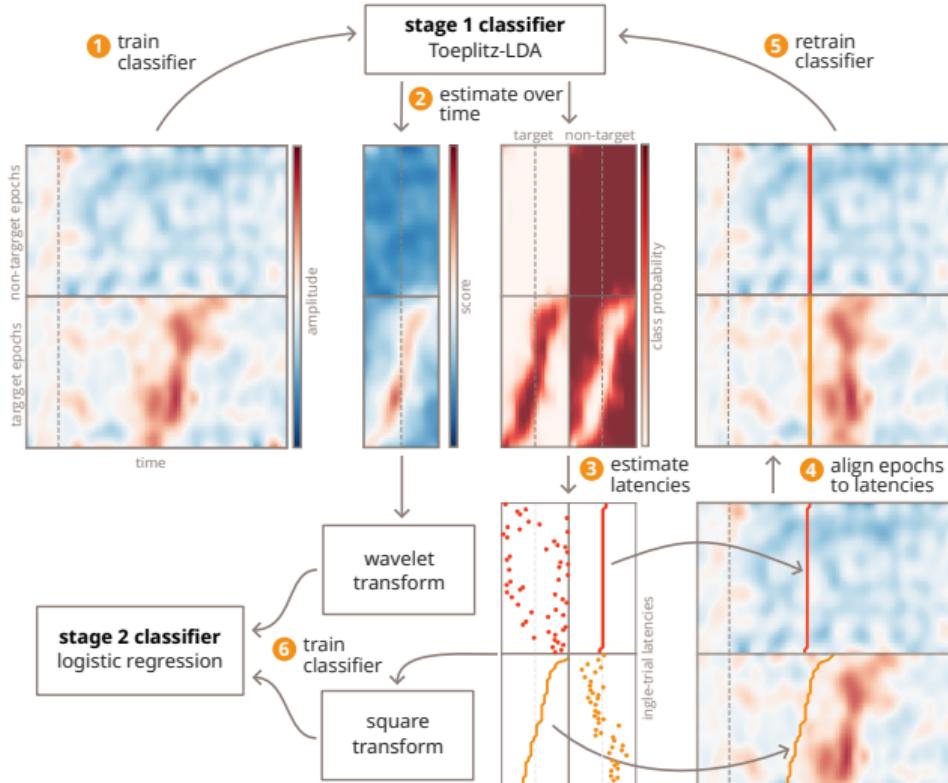
TODO: grand average ERPs and F-scores

Latency jitter decreases performance in covert and split attention

Aricò et al., 2014



Classifier-based Latency Estimation with Woody iterations



Aligned vs. non-aligned simulated data

Increase in gaze-independent decoding accuracy

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Experimental setup

- ▶ show setup
- ▶ list participants

Impaired visual skills

show table

Gaze tracking analysis

highlight specific gaze case

Usability

how many achieved higher than chance?

