

Differentiable learning of image encodings for cortical visual neuroprosthetics through bio/phenomenologically-aware phosphene modeling

Maureen van der Grinten^{1*}, Jaap de Ruyter van Steveninck^{1*}, Antonio Lozano^{2*},
Laura Pijnacker¹, Bodo Rückauer¹, Pieter Roelfsema², Marcel van Gerven¹,
Richard van Wezel¹, Umut Güçlü¹, Yağmur Güçlütürk¹

¹ Donders Institute for Brain, Cognition and Behaviour, Nijmegen, the Netherlands;

² Netherlands Institute for Neuroscience, Amsterdam, the Netherlands

Email: a.lozano@ninl.knaw.nl

*These authors have contributed equally to this work

Introduction: The development of high-channel count intracortical neuroprostheses for the blind[1], alongside recent demonstrations of these systems in human patients[2], establishes the possibility of restoring a rudimentary form of vision. Alongside these developments, the interdisciplinary field of neurotechnology is met with a wide array of challenges. Among these is the task of creating meaningful visual representations constrained by a limited implant resolution and safety and hardware constraints.

In order to gain insight on phosphene vision (i.e. creating artificial light percepts by electrical stimulation of the brain), simulated prosthetic vision grants researchers the ability to generate and test scientific hypothesis allowing for the optimization of computer image algorithms that are designed/trained to generate useful visual representations for the users [3], [4]. Here, we demonstrate how we can learn to create meaningful image representations using cortical phosphene representations in a biologically plausible way. In order to do that, we integrate decades of research evidence on modeling the effects of electrical stimulation on cortical tissue, coupled with electrophysiology and psychophysics data, into a novel, phenomenologically realistic differentiable artificial vision simulation pipeline.

Material, Methods and Results: An integrative model of phosphene perception, accounting for a wide array of psychophysical and neuroscientific evidence such as cortical magnification, current spread, phosphene thresholds, the relationship between electrical stimulation parameters and phosphene brightness, size, and temporal dynamics - including phosphene fading effects- is developed. Implemented in Pytorch, this model of phosphene perception, linked to computer vision algorithms based on Deep Neural Networks, allows for differentiable end-to-end learning of phosphene-based image representations on a broad diversity of conditions. These include realistic safety stimulation constraints, dynamic encoding of video data, and encoding of naturalistic images, in real time on a single GPU.

Discussion: While the neurophysiology and psychophysics of phosphene vision regarding cortical neural implants is still on its developmental beginning, an integrative pipeline able to create biologically and phenomenologically realistic cortical simulated prosthetic vision is a Prerequisite for the creation of when creating useful visual representations. Our simulations show highly correlated predictions with respect to the empirical psychophysics literature. In addition, the differentiable nature of our proposed modelling and optimization approach allows for deep learning-based end-to-end optimization of phosphene-based visual representations tailored to realistic physical and safety constraints.

Significance: A machine learning-compatible, realistic model of cortical phosphene perceptions enables neuroscientists, neuroengineers and clinicians to narrow the gap between prosthetic vision research and clinical applications.

Acknowledgements: This work was supported by two grants (NESTOR, INTENSE) of the Dutch Organization for Scientific Research (NWO) and the European Union's Horizon 2020 research and innovation programme (under grant agreement No 899287). We thank Xing Chen for help with the compilation and reviewing of relevant literature regarding phosphene perception.

References:

- [1] X. Chen, F. Wang, E. Fernandez, and P. R. Roelfsema, "Shape perception via a high-channel-count neuroprosthesis in monkey visual cortex," *Science*, vol. 370, no. 6521, pp. 1191–1196, Dec. 2020, doi: 10.1126/science.abd7435.
- [2] E. Fernández et al., "Visual percepts evoked with an intracortical 96-channel microelectrode array inserted in human occipital cortex," *Journal of Clinical Investigation*, vol. 131, no. 23, p. e151331, Dec. 2021, doi: 10.1172/JCI151331.
- [3] M. Beyeler, G. M. Boynton, I. Fine, and A. Rokem, "pulse2percept: A Python-based simulation framework for bionic vision," 2017.
- [4] J. de Ruyter van Steveninck, U. Güçlü, R. van Wezel, and M. van Gerven, "End-to-end optimization of prosthetic vision," *Bioengineering*, preprint, Dec. 2020. doi: 10.1101/2020.12.19.423601.