

## Peer Review for Evan Petrimoulx's Paper

### Summary of Presented Topic:

This paper investigates the use of deep learning models, particularly Convolutional Neural Networks (CNNs), for enhancing MRI image quality from undersampled k-space data. The paper explains the theory behind neural networks, the impact of reducing k-space sampling on MRI image quality, and the role deep learning can play in reconstructing high-quality images from limited frequency domain data.

### General Opinion on the Overall Quality of the Paper:

- The technical level is generally appropriate, clearly aimed at classmates with knowledge of MRI and basic machine learning concepts.
- The content is mostly well-organized, with a good progression from theory to practical methods and discussions.
- The language used is scientific, clear, appropriate, and accessible.

### Detailed Comments:

#### Positive Components:

- The abstract concisely summarizes the paper's purpose, the main approaches taken, and the key outcomes clearly.
- The introduction effectively establishes the problem of extensive scan times in MRI and the potential benefit of deep learning solutions.
- The explanation of basic machine learning concepts and deep learning theory is accessible and clearly presented, providing a good background for readers new to the topic.
- The use of specific examples and figures enhances the reader's understanding of how CNNs function and their potential in MRI applications.
- The methods section provides clear details on the practical aspects of dataset handling, preprocessing, and CNN training procedures, which support reproducibility.

#### Negative Components and Suggested Changes:

- ~~• Equation (1) on page 3 is fundamental but could benefit from additional explanation about the significance of the inverse Fourier transform, specifically in MRI image reconstruction.~~
- The report lacks a thorough discussion on the potential clinical limitations and practical constraints of applying deep learning models in real-world medical settings, such as varying patient anatomies.
- The dataset used is relatively small (581 images), which the author acknowledges; however, the manuscript would benefit from further elaboration on how dataset size might specifically impact the generalization capability of the CNN.

- The paper does not explicitly address strategies for dealing with common issues in deep learning for MRI, such as artifact detection and handling, which is an important aspect of real-world implementation.
- Additional figures or analyses demonstrating the improvements made by CNN over traditional reconstruction methods would strengthen the claims made.

Overall, this manuscript presents a strong foundation but would greatly benefit from expanded discussions on practical limitations, more rigorous techniques, and clearer descriptions of future research directions.