

Highlights

Bayesian classification of Parkinson's disease patients using fMRI data

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Bayesian classification of Parkinson's disease patients using fMRI data

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ABSTRACT

The goal of resting-state functional magnetic resonance imaging (fMRI) is to investigate the brain's functional connections by using the temporal similarity between blood oxygenation level dependent (BOLD) signals in different regions of the brain "at rest" as an indicator of synchronous neural activity. However having different images of patients' brains despite the sequential time based information, may contain enough data to investigate their brains and enables us to recognize some diseases. In this paper, we propose a novel method to automatically extract these non time sequence based information using a Bayesian deep learning algorithm based on convolutional neural network (CNN). Instead of using some predefined points of interest (POIs) we use the whole data in training phase so those points of the brain which do not contain related information about the disease will be ignored automatically by the trained model. Moreover this method does not take any assumptions about disease, patients, etc., makes it a possible universal disease diagnosis approach to differentiate diseases having an impact on brain functionality. This method is a supervised algorithm with small amount of calculations using three dimensional CNN. Each fMRI scan (which contains t time slices of brain) of patients will be divided into t different 3D images enabling us to make our dataset much bigger in number and calculations way simpler. Subsequently all of these images are fed to a Bayesian network similar to LeNet-5 (but in three dimensions) to train our model. Then to determine if a person is suffering from Parkinson's or not, we test his/her t fMRI images and get t different results which leads to a fraction (probability) of how unhealthy his/her brain is and if that fraction is above 0.5 we can classify that sample as a Parkinson's patient.

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A. My Appendix

Appendix sections are coded under \appendix.

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CRedit authorship contribution statement

Amin Amini: Conceptualization of this study, Methodology, Software.

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