Artificial Intelligence based detection of Parkinson's disease in Magnetic Resonance Imaging brain scans

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Background

Candidate neuroprotective treatments for PD are highlighting the need for early diagnostic tests. A number of exploratory imaging techniques have suggested that early pathological brain changes may be detectable using dedicated experimental MRI sequences.

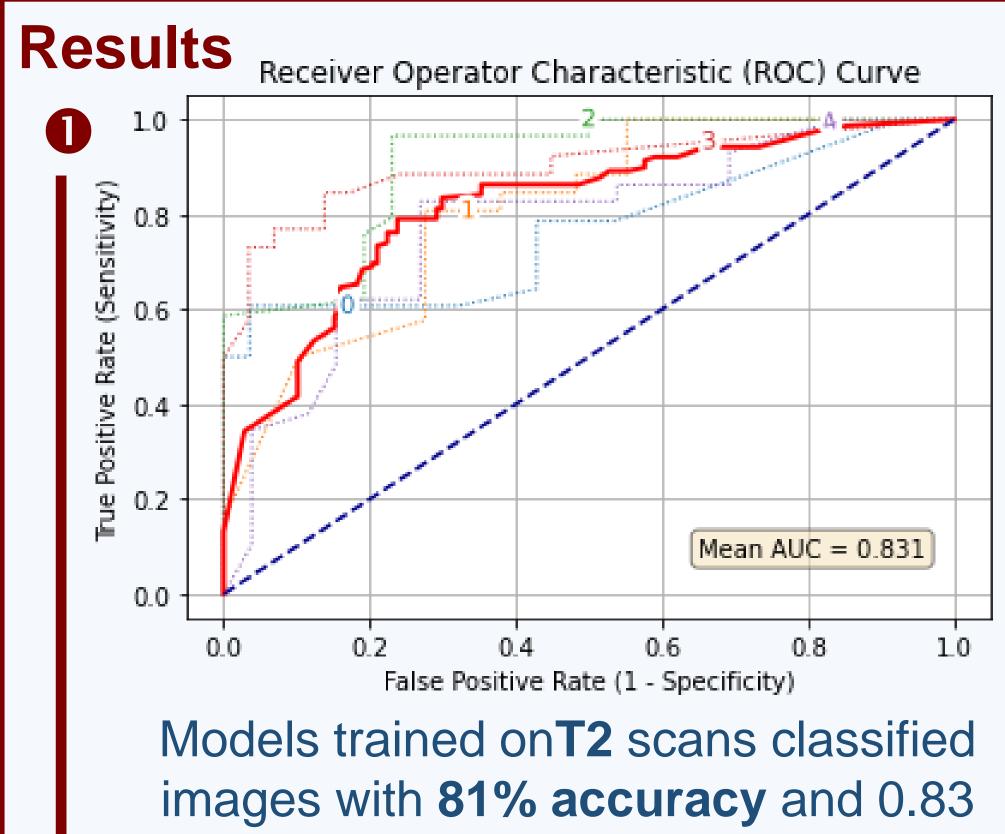
We explored whether machine learning (ML) might be employed to detect such brain changes on routine MRI scans. A subset of ML known as deep learning (DL) has shown great promise in diagnostic medical imaging, sometimes matching or even exceeding the diagnostic performance of radiologists.

DL offers the potential of automated diagnosis by detecting patterns that might be invisible to the human eye. DL methods have sometimes been criticised for being "black boxes", but newly emerging explainability methods are allowing the decisions made by DL models to be better interpreted.

Methods

We trained a convolutional neural network to classify 138 PD and 60 control brain MRI images acquired from the Parkinson's Progression Marker Initiative (PPMI) database. Models were assessed using 5-fold cross-validation.

We used Deep SHapley Additive exPlanations (DeepSHAP) to calculate and visualise the contribution of individual pixels to the model's prediction.



Mean AUC = 0.894 0.0 False Positive Rate (1 - Specificity)

Receiver Operator Characteristic (ROC) Curve

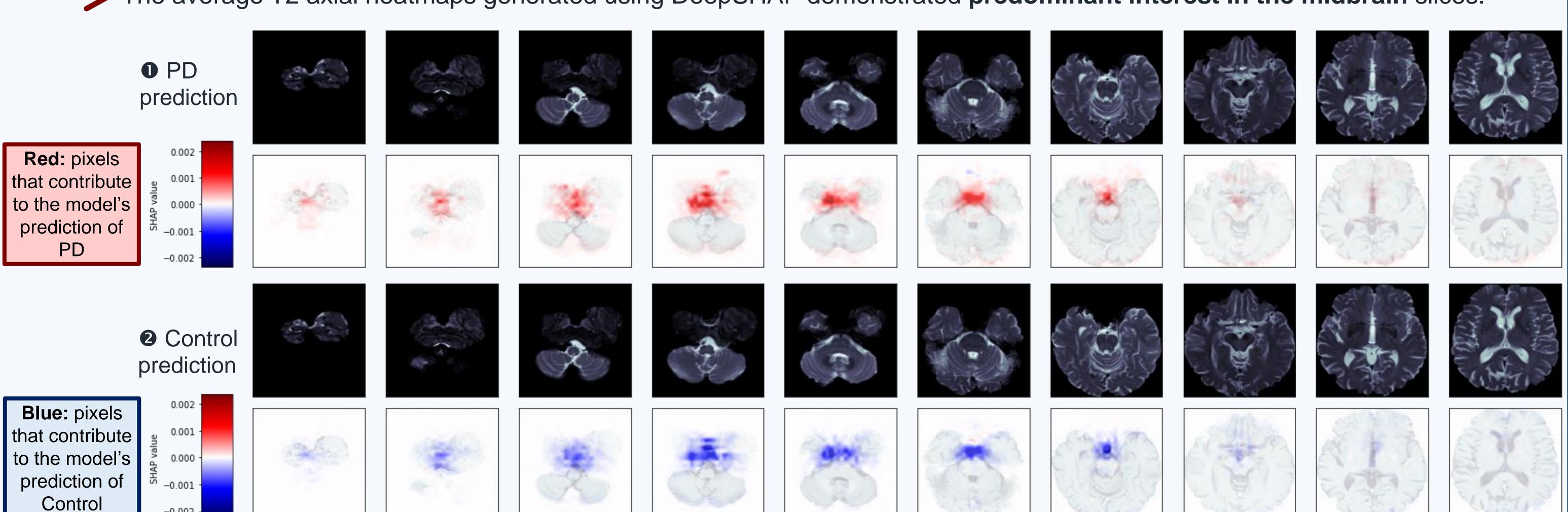
Receiver Operator Characteristic (ROC) Curve 3 (Sensitivity) Mean AUC = 0.855 False Positive Rate (1 - Specificity)

ROC area-under-the-curve (AUC).

Models trained on **proton density** scans classified images with 84% accuracy and 0.89 ROC AUC.

Models trained on a combined dataset of axial T2 and proton density MRI scans classified images with 79% accuracy and 0.86 ROC AUC.

The average T2 axial heatmaps generated using DeepSHAP demonstrated predominant interest in the midbrain slices:



Conclusion

Our models exhibited good diagnostic performance. The use of explainable AI highlighted regions of interest consistent with the known neuropathology of PD, providing a focus for future work. We will validate these models in a large dataset of routinely collected NHS MRI scans, many of which precede onset of motor symptoms.



-0.002





