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# Predicting Alzheimer's from PET/MR Images with Deep Learning

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# Introduction

Alzheimer's disease is a devastating neurodegenerative condition that affects millions of people worldwide. Early detection and accurate diagnosis are critical for effective treatment and intervention. In recent years, the integration of medical imaging techniques, such as Positron Emission Tomography (PET) and Magnetic Resonance Imaging (MRI), with deep learning algorithms has shown promise in predicting Alzheimer's disease at its early stages. This article explores the potential of using deep learning to predict Alzheimer's disease from PET/MR images, highlighting its significance in early diagnosis and potential future applications. Alzheimer's disease is characterized by progressive cognitive decline, memory loss and behavioral changes. It not only affects the quality of life of patients but also places a significant burden on their families and healthcare systems. Early detection is crucial because treatments and interventions are more effective in the early stages of the disease. PET and MRI imaging provide valuable insights into the brain's structural and functional changes associated with Alzheimer's. PET and MRI are two distinct imaging modalities that offer complementary information about the brain. PET scans involve the injection of a radioactive tracer that highlights specific biochemical processes in the brain, such as the accumulation of beta-amyloid plaques and tau tangles, which are hallmark features of Alzheimer's disease. Meanwhile, MRI provides high-resolution images of brain structure, including atrophy patterns and white matter changes [1].

# **Description**

Combining PET and MRI in a single examination, known as PET/ MR imaging, offers a holistic view of both functional and structural brain alterations associated with Alzheimer's disease. PET/MR imaging has the potential to enhance the accuracy of early diagnosis and better understand the disease's progression. Deep learning, a subset of artificial intelligence, has shown remarkable capabilities in processing and interpreting complex medical images. Convolutional Neural Networks (CNNs) are a type of deep learning architecture particularly well-suited for image analysis tasks. These networks can automatically learn and extract relevant features from medical images, enabling them to identify subtle patterns and abnormalities that might be challenging for human experts to discern. PET and MR images need to be preprocessed to enhance their quality and normalize their intensity. This step ensures that the deep learning model receives consistent and informative input. Deep learning models, particularly CNNs, excel at automatically extracting relevant features from images. In the context of Alzheimer's prediction, CNNs can capture patterns related to beta-amyloid deposition, brain atrophy and other biomarkers [2].

The deep learning model is trained using a labeled dataset consisting of PET/MR images from both Alzheimer's patients and healthy individuals. During training, the model learns to differentiate between these two groups

\*Address for Correspondence: Qianrui Zhang, Department of Neurology, Dalian Medical University, Dalian, China; E-mail: qianruizhang@gmail.com Copyright: © 2023 Zhang Q. This is an open-access article distributed under the terms of the Creative Commons Attribution License, which permits unrestricted use, distribution, and reproduction in any medium, provided the original author and source are credited.

Received: 31 June, 2023, Manuscript No. jbr-23-116555; Editor Assigned: 03 August, 2023, PreQC No. P-116555; Reviewed: 17 August, 2023, QC No. Q-116555; Revised: 22 August, 2023, Manuscript No. R-116555; Published: 28 August, 2023, DOI: 10.37421/2684-4583.2023.6.206 by optimizing its internal parameters. After training, the model's performance is evaluated on separate datasets to assess its ability to predict Alzheimer's disease accurately. Cross-validation techniques are often used to ensure robustness. The ultimate goal is to develop a deep learning model that can be used in clinical practice to assist radiologists and clinicians in making more accurate Alzheimer's diagnoses [3]. This may involve the integration of the model into existing healthcare systems. Deep learning models can detect subtle changes in brain images, enabling early diagnosis when interventions are most effective.

Unlike subjective assessments by human experts, deep learning models provide objective and consistent results. Early diagnosis and intervention can lead to cost savings by delaying disease progression and reducing the need for extensive long-term care. High-quality labeled datasets are essential for training deep learning models. Access to large and diverse datasets can be a bottleneck in developing accurate models. Deep learning models are often considered "black boxes," making it challenging to understand the features or patterns they use for predictions. Privacy concerns and ethical considerations must be addressed when dealing with sensitive medical data. Combining information from different modalities, such as PET, MRI and clinical data, may improve prediction accuracy [4].

Monitoring changes in patients over time can provide insights into disease progression and response to treatment. Leveraging pre-trained deep learning models on large datasets can improve the performance of models with limited labeled data. Integrating deep learning models into clinical workflows to aid radiologists and clinicians in making more informed decisions. Predicting Alzheimer's disease from PET/MR images with deep learning holds great promise in early diagnosis and intervention. By leveraging the complementary strengths of PET and MRI and harnessing the power of deep learning, we can detect subtle brain changes associated with Alzheimer's disease. While challenges remain, ongoing research and advances in technology are paving the way for more accurate and accessible tools for Alzheimer's prediction, ultimately improving the lives of those affected by this devastating condition [5].

### Conclusion

Medical imaging has revolutionized our ability to visualize the human brain and detect abnormalities associated with neurodegenerative diseases, including Alzheimer's. PET and MRI scans have become indispensable tools in Alzheimer's research and diagnosis, offering unique insights into the structural and functional changes in the brain. PET scans are used to visualize metabolic processes in the brain. They involve injecting a radioactive tracer into the patient's bloodstream, which is then absorbed by brain cells. The level of tracer absorption indicates the brain's metabolic activity. In Alzheimer's disease, the PET scan can reveal reduced glucose metabolism, a hallmark of the disease. MRI scans provide detailed structural information about the brain. They can detect atrophy in specific brain regions, which is a common sign of Alzheimer's. Combining structural MRI with other imaging techniques, like PET, can enhance the diagnostic accuracy. Analyzing PET/MR images is a highly complex and time-consuming task, often relying on expert radiologists' visual assessments. The variability in human interpretation can lead to inconsistencies and potential errors in diagnosis.

Alzheimer's disease begins with subtle changes in the brain that can be challenging to detect in the early stages. Current diagnostic methods often rely on identifying advanced disease markers, missing the opportunity for early intervention. The wealth of imaging data generated by PET/MR scans can overwhelm healthcare providers. Analyzing these data manually is resourceintensive and prone to human error. Deep learning, a subset of machine learning, has shown remarkable promise in the field of medical imaging, particularly in the early detection and diagnosis of Alzheimer's disease. Deep learning algorithms can be trained to recognize subtle patterns and abnormalities in PET/MR images, making them valuable tools for automating and enhancing the diagnostic process.

# **Acknowledgement**

None.

## **Conflict of Interest**

None.

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How to cite this article: Zhang, Qianrui. "Predicting Alzheimer's from PET/MR Images with Deep Learning." *J Brain Res* 6 (2023): 206.