Introduction:

Alzheimer's disease (AD) is a progressive neurological disorder that affects memory, thinking, and behavior. Early detection is crucial for improving patient care and management, but it remains a significant challenge due to the complexity of contributing factors and the long progression period of the disease. This project aims to develop a predictive model to assess the likelihood of an individual having Alzheimer's disease based on a range of features, including demographic details, medical history, lifestyle factors, clinical measurements, and cognitive assessments.

Below is a summary of the modeling approach for this project.

Goals:

- To obtain a good baseline model
- Select models that produce high accuracy rates using all the features
- Adapt an approach which improves the accuracy for early detection
 - By excluding the highly correlated features in modeling
 - Restricting to the dataset with no memory complaint

Models Explored:

We will be exploring the following models in this project.

- 1. Logistic Regression
- 2. Random forest classifier
- XGBoost classifier
- 4. Gradient boosting classifier
- 5. Adaboost classifier
- 6. KNN classifier
- 7. Support Vector machine
- 8. Decision Tree
- 9. LDA
- 10. QDA
- 11. Naĭve Bayes

Metrics Checked:

To evaluate the performance of the models, we will be considering the following metrics.

- Accuracy Score
- F1 Score
- Precision Score
- Recall Score
- Confusion Matrices