

LOW LEVEL DESIGN (LLD)

Alzheimer Disease Classifier

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1. Introduction

1.1 What is a Low-Level Design Document?

The **Low-Level Design (LLD)** document provides detailed information about the internal structure of the **Alzheimer Disease Classifier** system. It describes class structures, module functionalities, and the relationships between various components. The purpose of this document is to provide a blueprint for implementing the classifier using **VGG16** for **binary classification** (Alzheimer's Positive vs. Negative).

1.2 Scope

The **LLD** includes:

- Data ingestion and preprocessing steps.
- Model architecture using VGG16 and fine-tuned layers.
- Training methodologies and evaluation metrics.
- Model inference and deployment strategy.
- Unit test cases to validate each module.



2. Architecture

Workflow Overview

1. Data Acquisition

- Medical images (MRI scans) are collected.
- o Data is stored securely for further processing.

2. Pre-processing & Augmentation

o Image resizing, normalization, and augmentation.

3. Model Training (VGG16-based)

- o Train model with labelled Alzheimer's images.
- o Use transfer learning with fine-tuned layers.

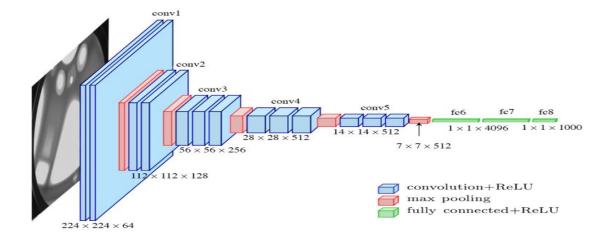
4. Model Evaluation

o Validate performance using metrics (Accuracy, AUC, Precision, Recall).

5. Model Inference & Deployment

- Deploy model via API.
- o Accept image input and return classification results.





3. Architecture Description

3.1 Data Description

- Dataset: MRI scan images labelled for Alzheimer's detection.
- Format: Images stored in JPEG/PNG format.
- Classes: Binary classification Alzheimer's Positive and Alzheimer's Negative.

3.2 Data Pre-processing

- Image Resizing: Convert images to 224x224 pixels for VGG16.
- Normalization: Scale pixel values between 0 and 1.
- Augmentation: Apply techniques like rotation, flipping, and brightness adjustment to improve model robustness.
- Dataset Splitting:

Training Set: 80%

Validation Set: 10%

Test Set: 10%

3.3 Model Training

- Base Model: VGG16 pre-trained on ImageNet.
- Fine-Tuning:



- o Freeze convolutional layers to preserve pre-trained features.
- o Add new dense layers with dropout for regularization.
- Loss Function: Binary Cross-Entropy.
- Optimizer: Adam.
- Metrics:
 - Accuracy
 - o AUC-ROC
 - o Precision/Recall
- Training Method:
 - o **Epochs:** 20-50 (based on early stopping).
 - o Batch Size: 32 or 64.

3.4 Model Evaluation

- Validation Metrics:
 - o Confusion Matrix: Evaluates TP, FP, FN, TN.
 - o AUC-ROC Curve: Measures model discrimination.
 - o **Precision-Recall Curve:** Assesses model effectiveness.
- **Cross-validation:** Used to prevent overfitting.

3.5 Model Inference and Deployment

- Inference Pipeline:
 - Accepts images via API.
 - o Applies pre-processing (resizing, normalization).
 - Runs inference using trained model.
 - Returns probability scores and classification labels.
- Deployment Strategy:
 - Model hosted using Flask / FastAPI.
 - o Containerized using **Docker**.
 - o Deployed on AWS / GCP / Azure for scalability.
- Security Considerations:



- o Data encryption in transit and at rest.
- o User authentication for API access.

4. Unit Test Cases

Test Case	Pre-Requisite	Expected Result
Verify image upload functionality	API is running	Image should be successfully uploaded
Verify image pre- processing	Image is uploaded	Image should be resized and normalized
Verify model inference	Pre-processed image is ready	Model should return classification output
Verify API response time	API is running	Response time should be < 500ms
Verify model accuracy	Trained model is available	Accuracy should be above 85%



Test Case	Pre-Requisite	Expected Result
Verify security authentication	API requires authentication	Only authenticated users can access results

Conclusion

This **Low-Level Design (LLD)** document provides a structured approach to developing the **Alzheimer Disease Classifier** using a **VGG16-based model**. It outlines data processing, model training, inference, deployment, and security considerations, ensuring a scalable and efficient classification system.