



AI-Based Retinal Disease Detection Using Deep Learning

By Vidish Kumar



Introduction to Diabetic Retinopathy Detection



Global Challenge

DR causes avoidable blindness worldwide



AI Advantage

Deep learning enables faster, accurate analysis



Current Limitations

manual detection slow, needs specialists



Early Detection Importance

Crucial to prevent vision loss



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Project Objectives

Multi-class Classification

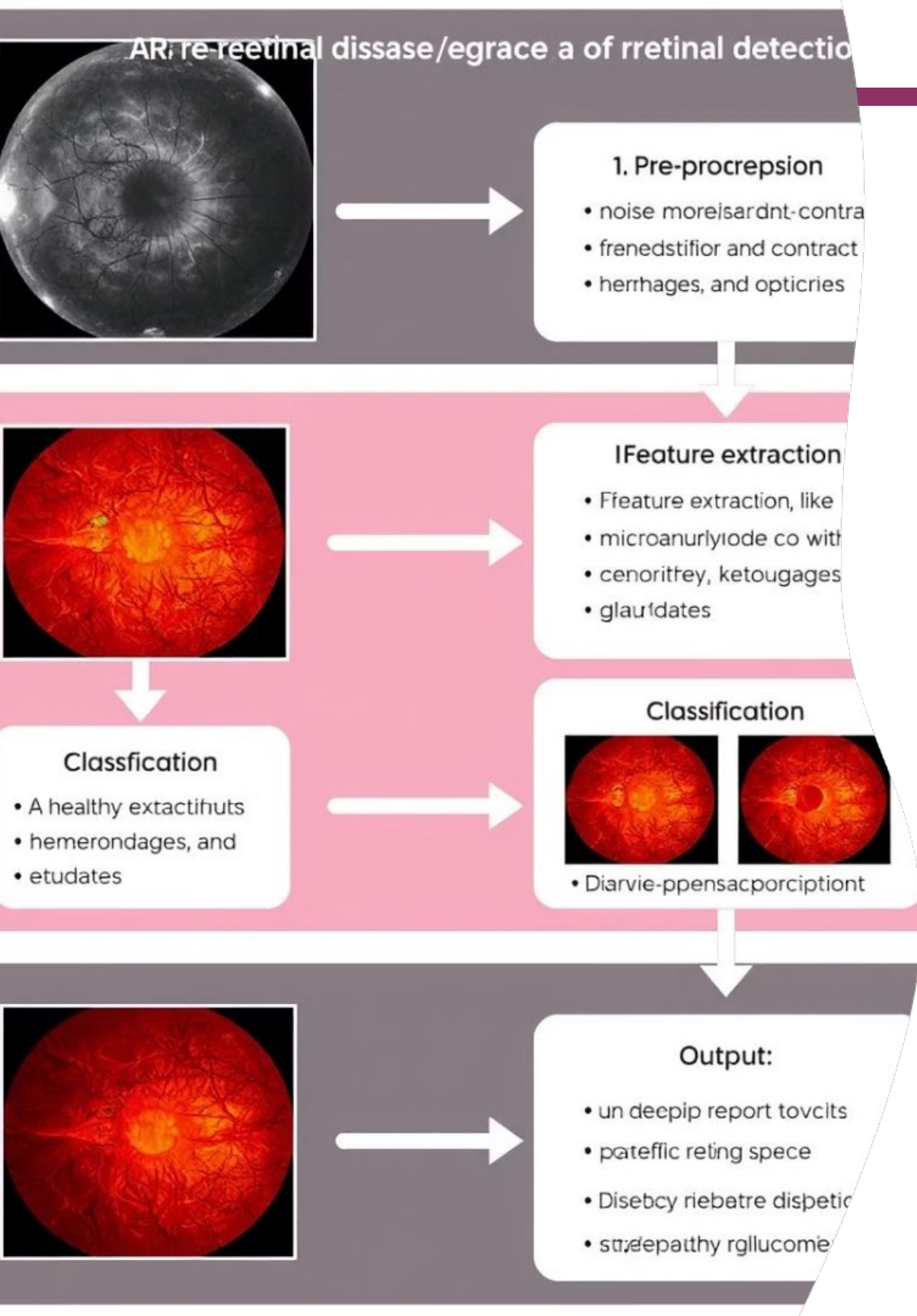
5 DR severity levels predicted

Automation

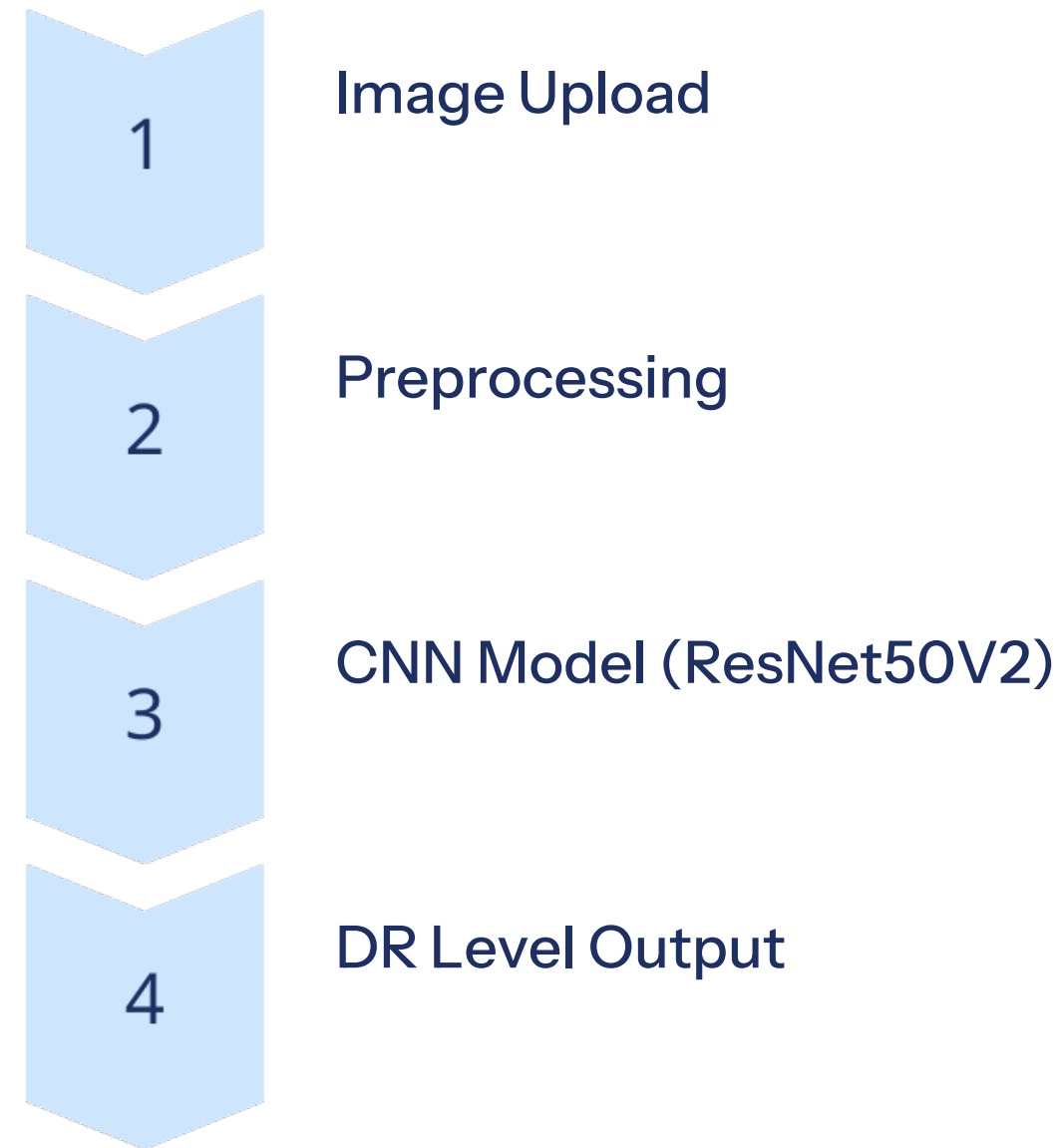
Reduce manual screening
reliance

Early Diagnosis

Prevent blindness through timely detection



System Workflow





Dataset & Preprocessing

Dataset

20,000 retinal images categorized into 5 severity classes.

Preprocessing

- Normalization for consistent brightness
- Augmentation: rotation, zoom, flip, shear, brightness
- Data split: 60% train, 25% validation, 15% test



Input Layer



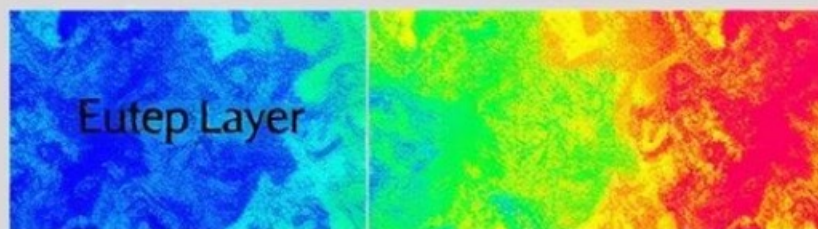
Activation
Convolutions



Pooling
Layer



Output Layer



Model Architecture: ResNet50V2 Transfer Learning

Base Model

Pretrained on ImageNet

Fine-Tuning

Last 50 layers unfrozen for training

Added Layers

Pooling, BatchNorm, Dense, Dropout, Softmax



Training & Evaluation

Training Setup

Adam optimizer with learning rate $1e-4$, 10 epochs on augmented data.

Results

Validation accuracy stable at 75.91%, test accuracy 75.40%.

Loss function: categorical crossentropy with final loss 0.7561.

Applications, Benefits & Future Work

Clinical & Remote Use

Fast, accessible screening reduces specialist workload.

Future Improvements

- Boost accuracy beyond 85%
- Detect additional eye diseases
- Optimize for mobile and edge devices
- Expand clinical trials and telemedicine support

