

VeinScope



Tear Vein Pattern Analysis for Early
Disease Mapping

Presented by: **Group-12**

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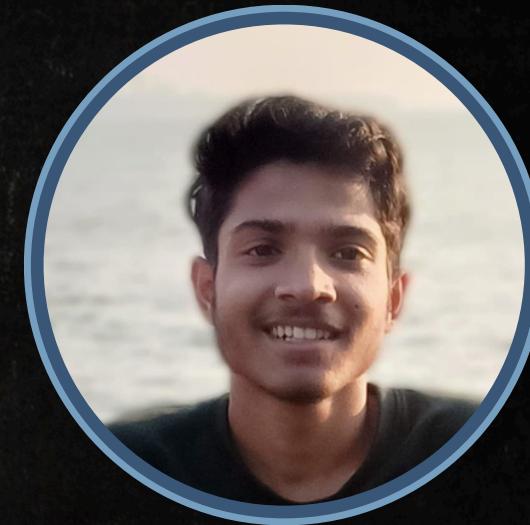
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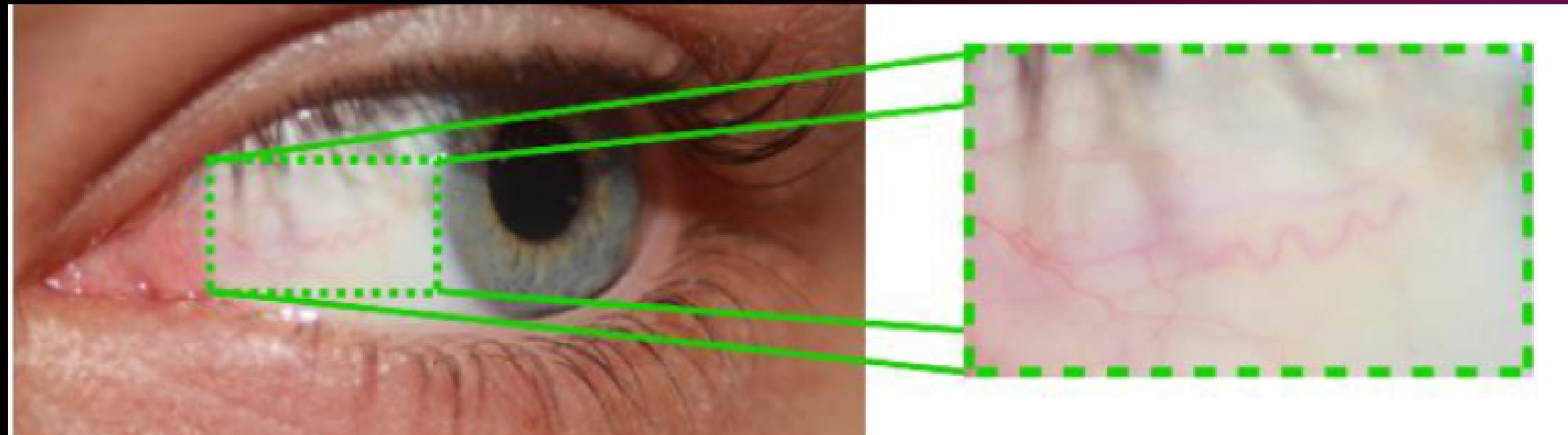


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Problem Overview



The goal is to analyze close-up eye images, especially the tear region, to:

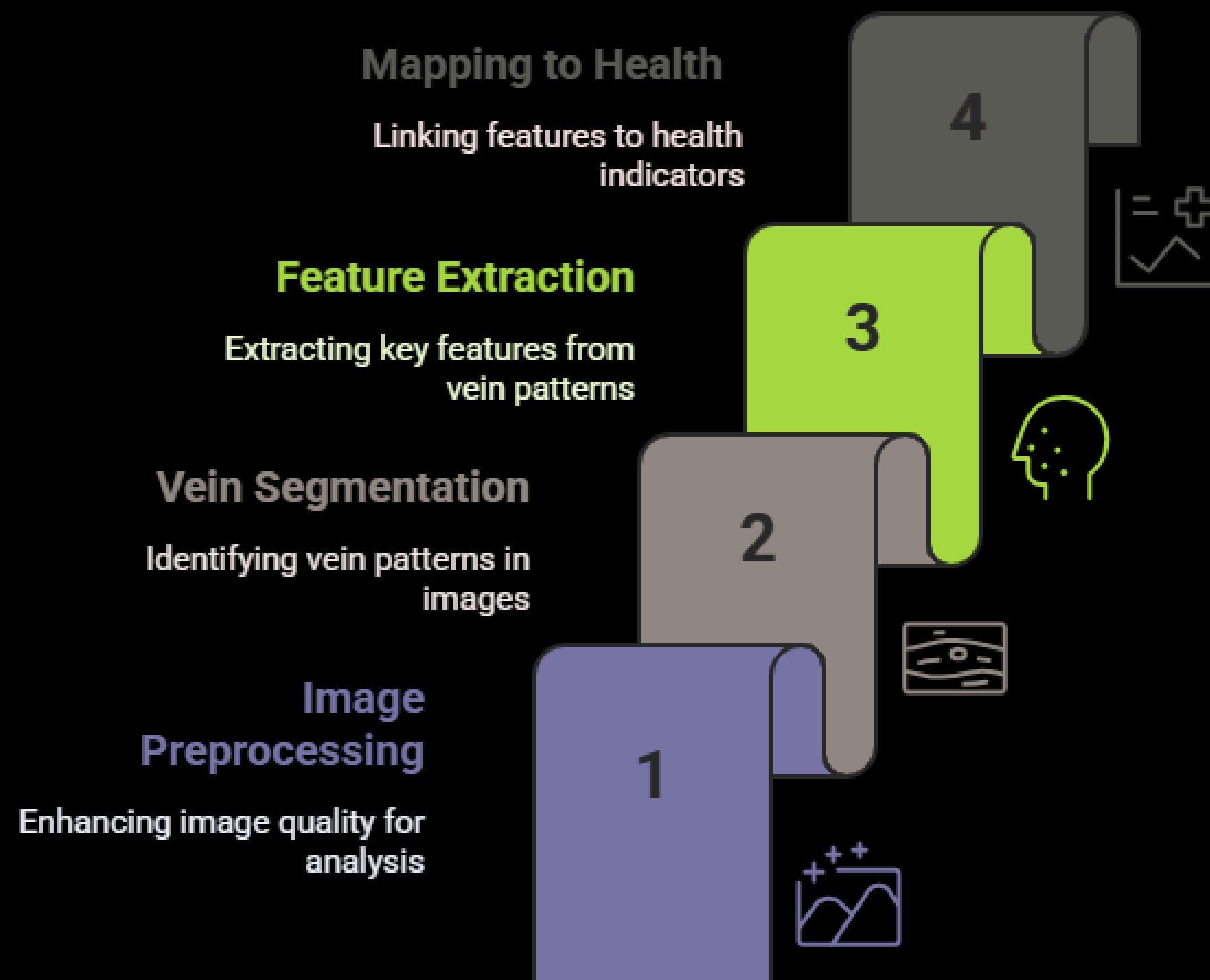
- Segment tear vein structures
- Extract and quantify vein patterns
- Optionally map features to health indicators

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Pipeline Components and Methods

Vein pattern analysis pipeline



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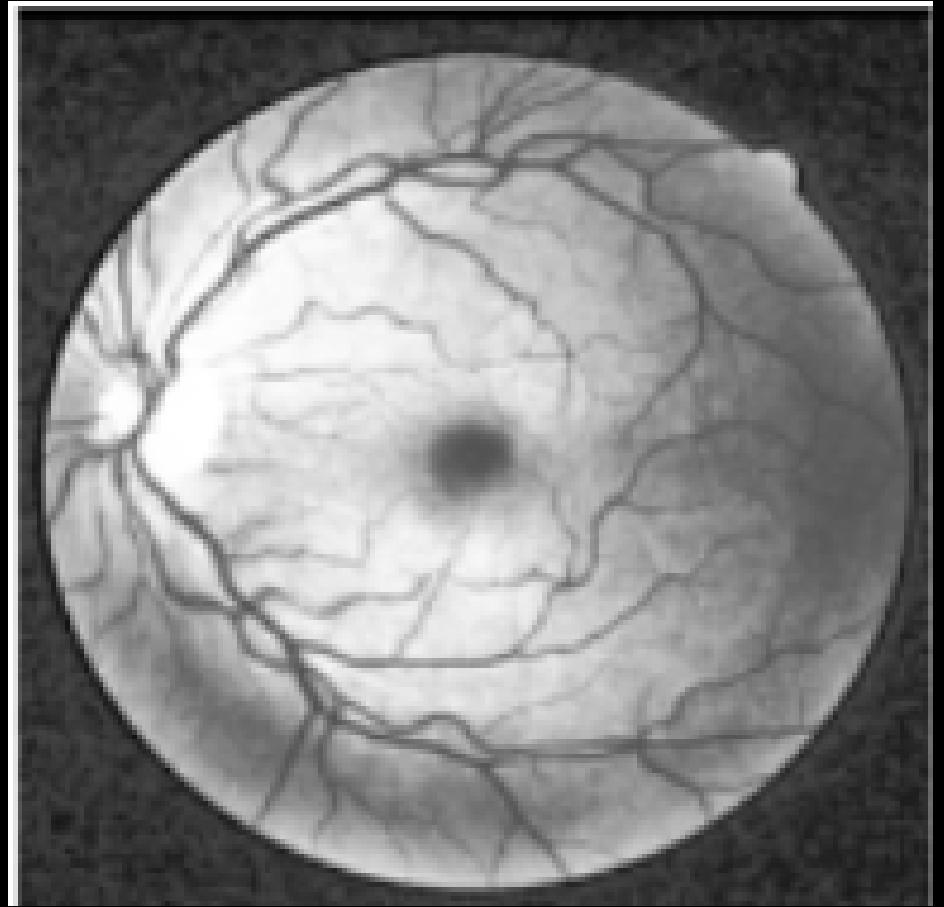
Image Preprocessing

Purpose:

- Improve visibility and isolate the region of interest (ROI).

Methods:

- **CLAHE (Contrast Limited Adaptive Histogram Equalization):** Enhances contrast in small regions to make veins more visible.
- **Green Channel Extraction:** Veins show high contrast in the green channel of RGB images.
- **Grayscale Conversion + Histogram Stretching:** Improves contrast and suppresses noise.
- **ROI Cropping:** Using SAM model Use eye landmark detection or manual ROI to focus on the tear region.

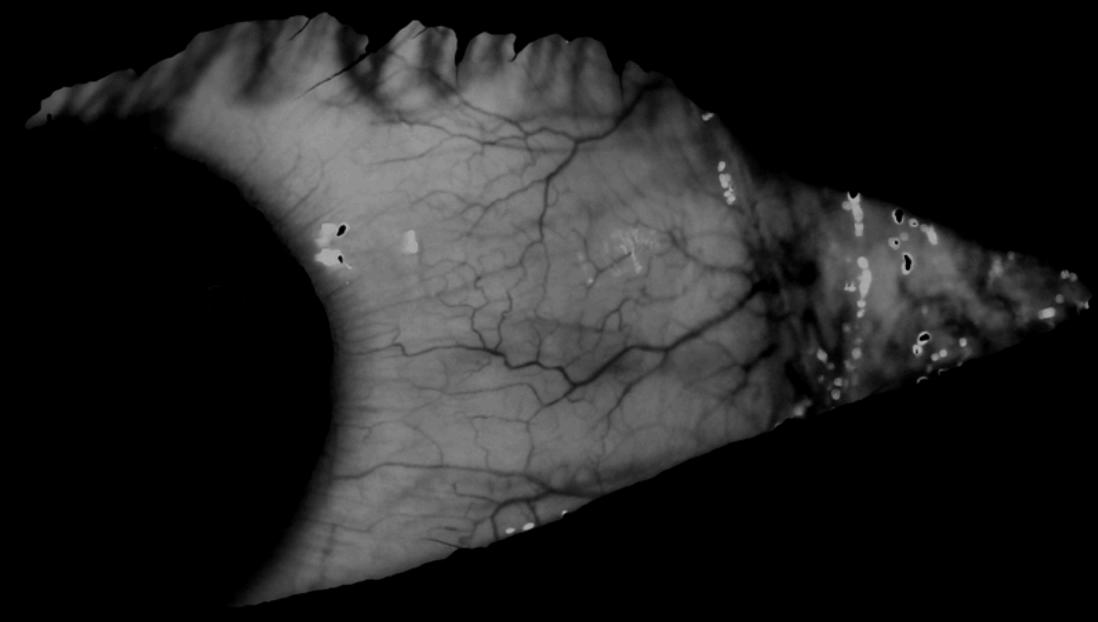


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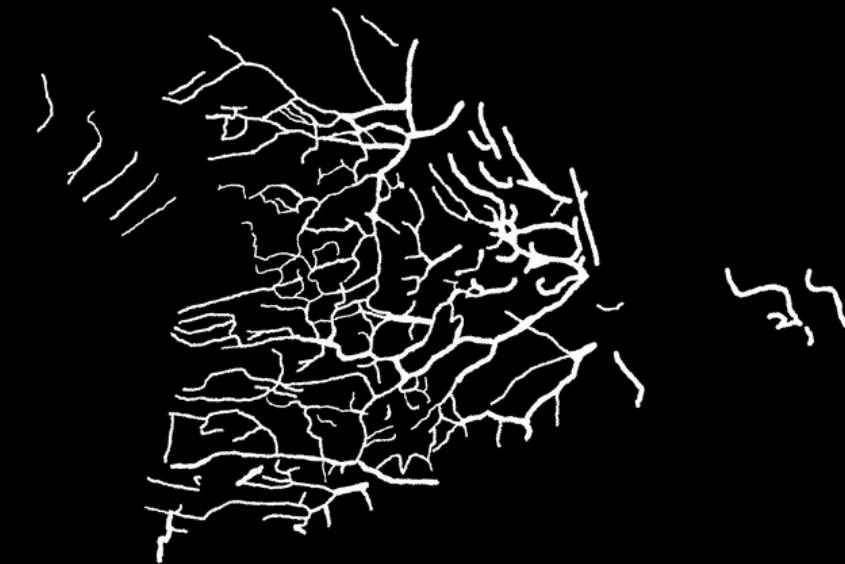


Image Preprocessing

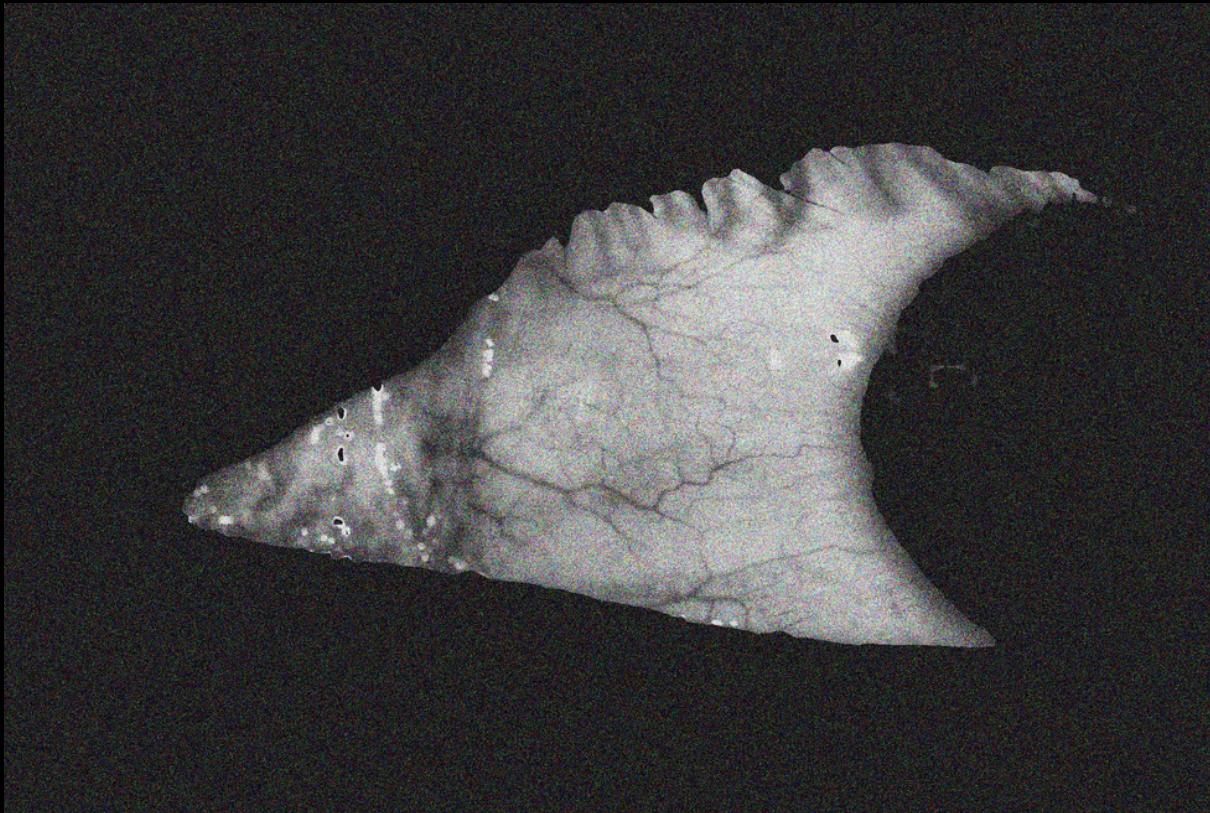
Image Augmentation



Segmented



Augmented Images



Segmented



de



Image Preprocessing

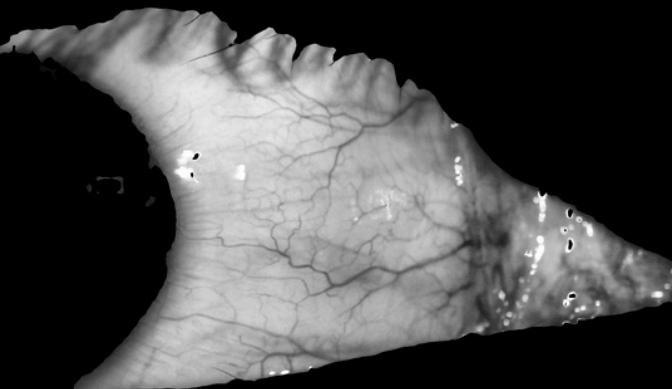


Original

Without SAM, Green
Channel Extraction
(CLAHE)



With SAM



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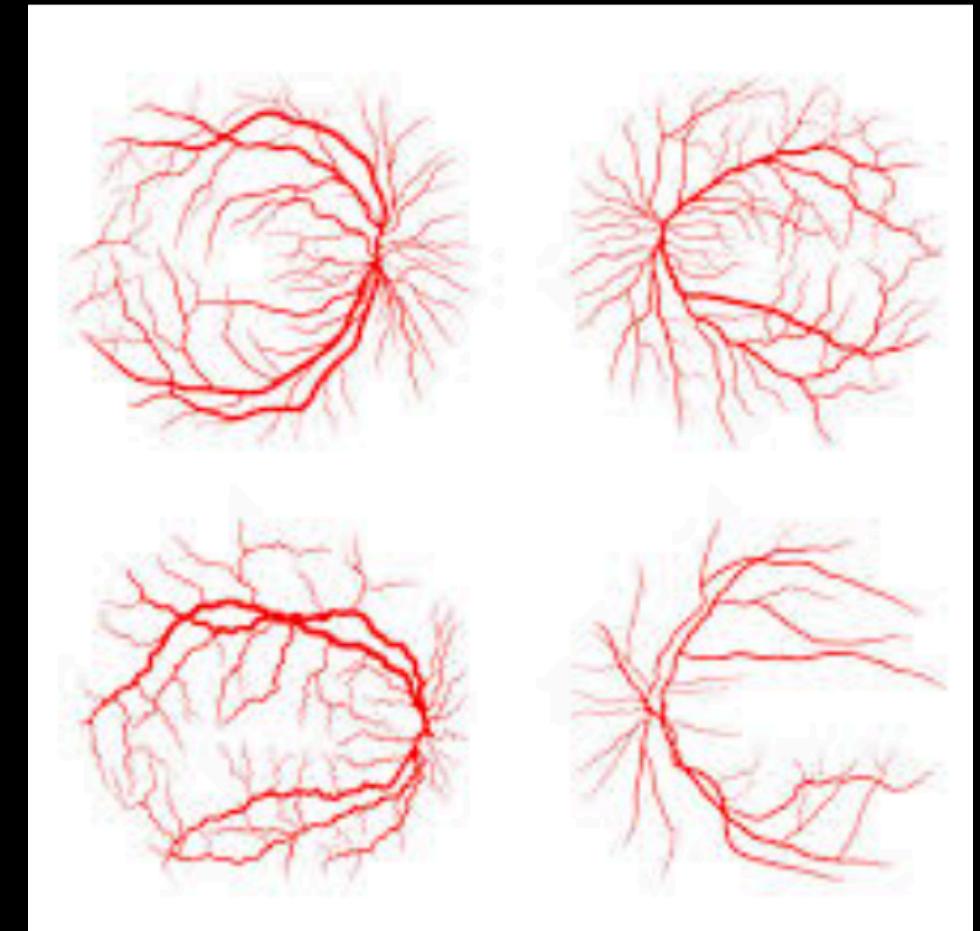
Vein Pattern Segmentation

Purpose:

- Detect thin vascular structures from images.

Methods:

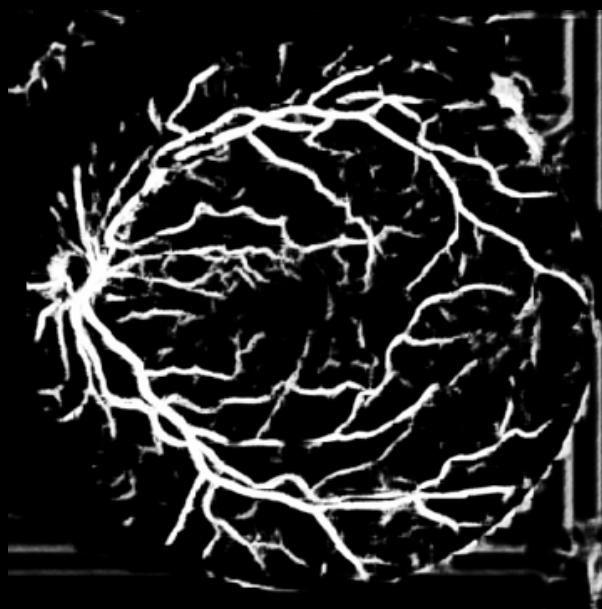
- **U-Net:** A convolutional encoder-decoder network with skip connections—standard for biomedical image segmentation.
- **ResNet:** Pretrained on ImageNet (RGB images)
- **nnUNet:** Dynamically adjusts preprocessing, architecture, and training based on data



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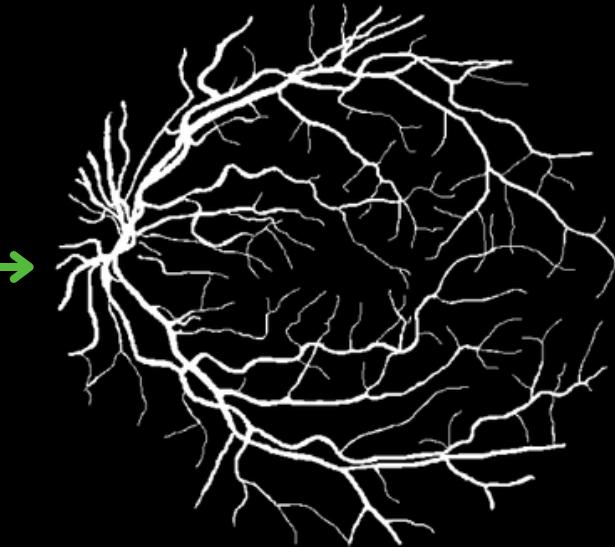
Vein Pattern Segmentation



Segmented



Original



Ground Truth



Segmented



Original



Ground Truth



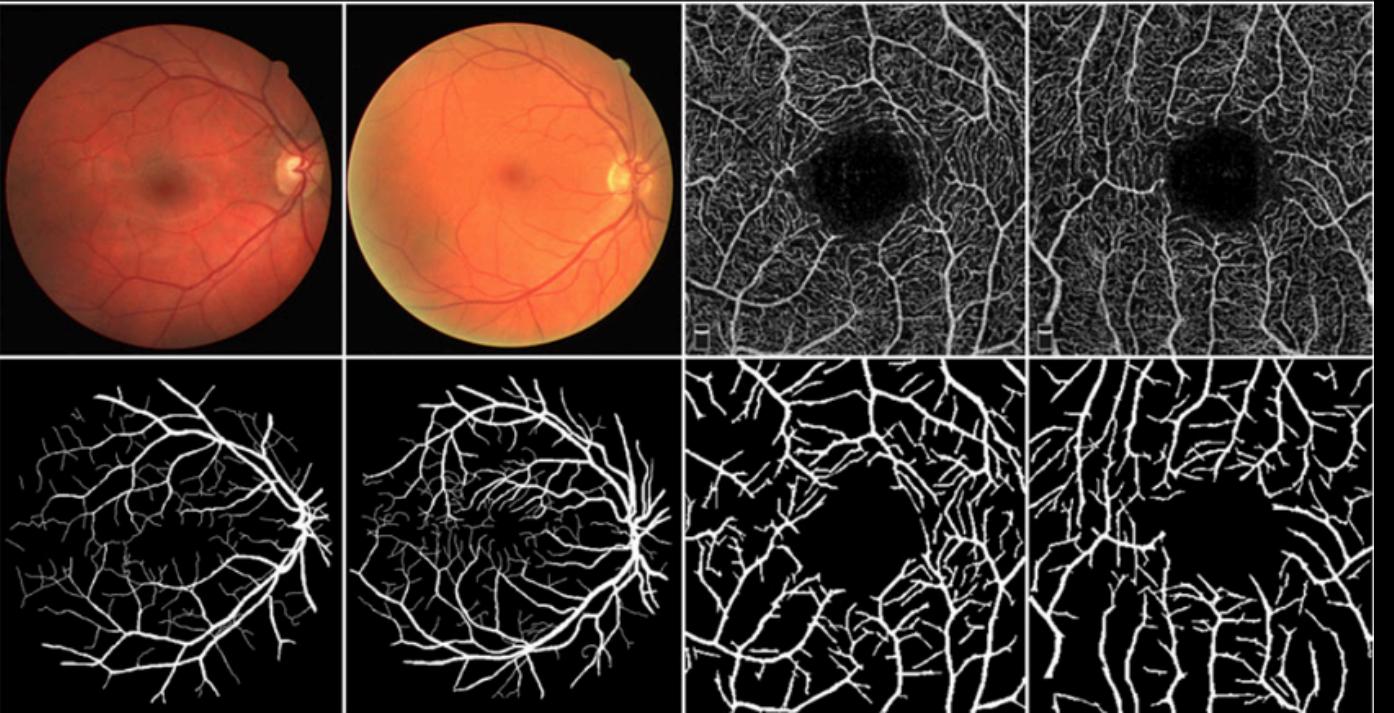
Feature Extraction

Purpose:

- Quantify vein patterns for downstream analysis.

Techniques:

- **Fractal Dimension:** Measures complexity of vein branching.
- **Vein Density:** Ratio of vessel pixels to total ROI.
- **Tortuosity:** Curviness of vein paths.
- **Orientation Histograms:** Angles of vessel branches to detect symmetry/asymmetry.



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Mapping to Health Indicators

Goal:

- Associate vein patterns with risk profiles.

Approaches:

- VeinScope extracts key morphological features from the binary vein masks using scikit-image and OpenCV libraries.

3.2 Health Indicator Mapping

Based on published literature and clinical research, the extracted features were mapped to potential health indicators:

Table 3.2: Mapping of Feature Patterns to Health Indicators

Feature Patterns	Potential Health Indicators
High tortuosity & branch density	Diabetic Retinopathy
Reduced vessel width and area	Aging, Dehydration
Dense branching & thick veins	Hypertension
Irregular width & branching	Proliferative Vascular Diseases
Sparse veins, many endpoints	Circulatory blockages, Dry Eye Syndrome

These mappings are intended for screening purposes only

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Evaluation Metrics

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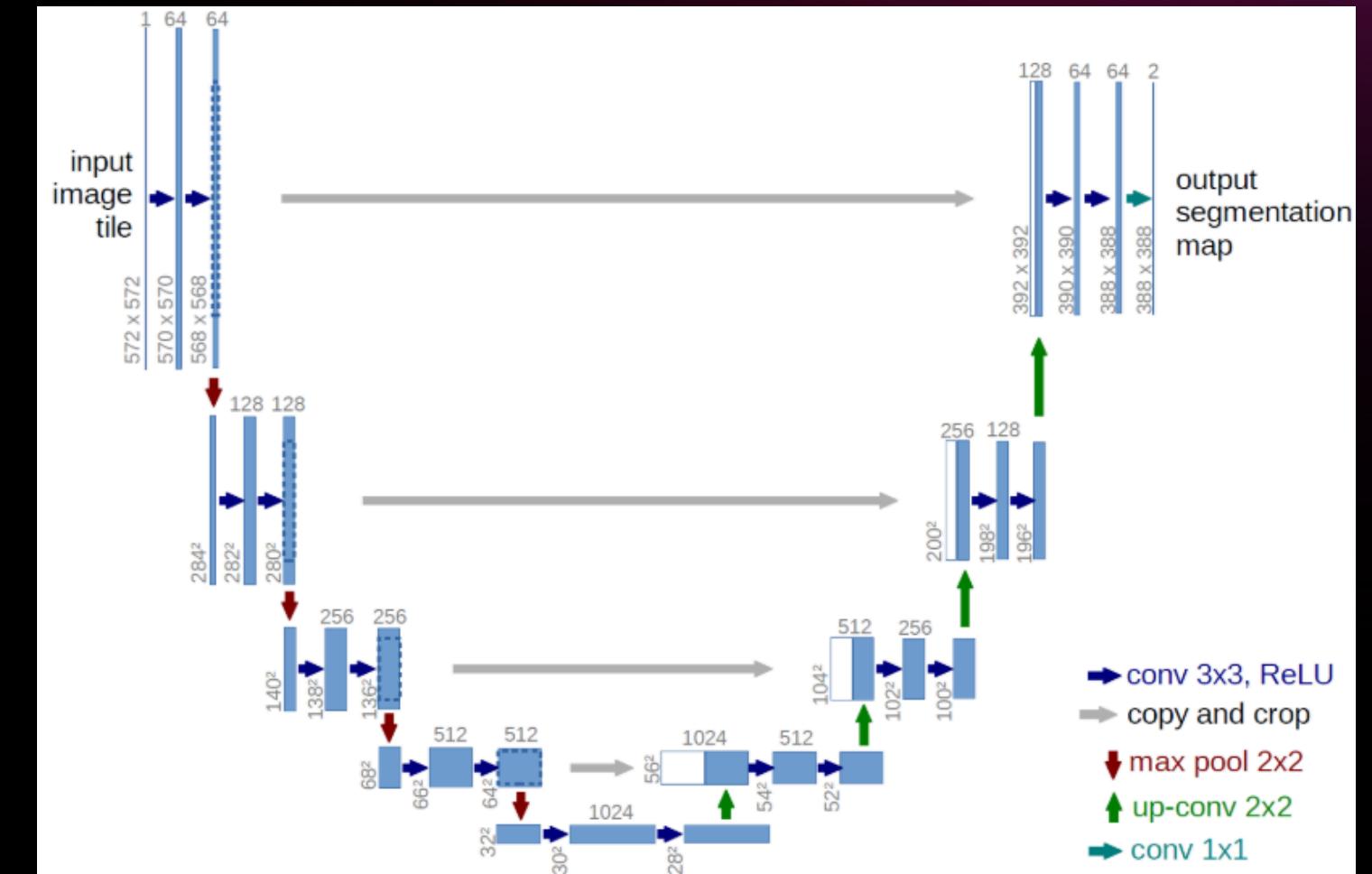
Task	Metric
Segmentation	Dice Score
Qualitative	Sharpness, continuity
Health Mapping	Cluster compactness

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Tools & Libraries

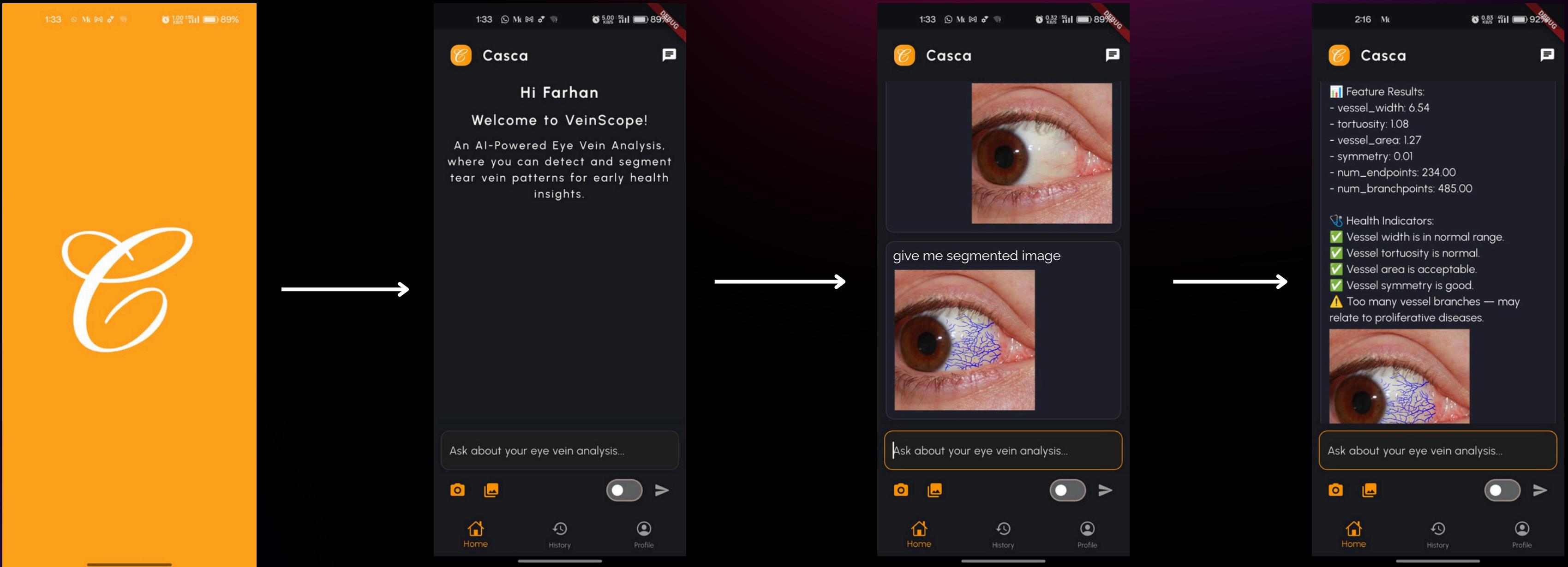
- Python, NumPy, OpenCV, scikit-image, PyTorch/Keras, Al augmentations (augmentation)
- Pretrained models (U-Net)



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Our App - Casca

An Android application named “Casca” was developed to provide a user-friendly interface for the VeinScope system.



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Thank You!