Retinal OCT Classification Challenge

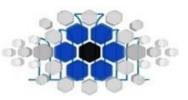
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Outline



- Denoising
- 2 Segmentation
- Feature Extraction
- Supervised Learning
 - Support Vector Machine (SVM)

Denoising



We have used various low pass filters for speckle noise reduction which are introduced in Medical Imaging.

Speckle noise reduction methods

- Median Filter
- Adaptive Weighted Median Filter
- Fourier Ideal Filter
- Butterworth Filter
- Wavelet Filter
- Homomorphic Ideal Filter
- Homomorphic AWMF Filter
- Homomorphic Butterworth Filter
- Homomorphic Wavelet Filter



Denoising



Original Image

Homomorphic Wavelet Filter (Double Level Decomposition) Band Eliminated = HH

Median filter windows size [5 5]

Median filter windows size [5 5]

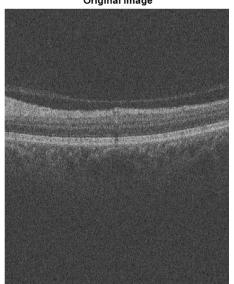
Median filter windows size [5 5]

Image after denoisieng

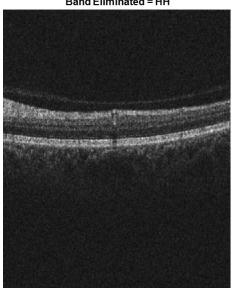
Denoising



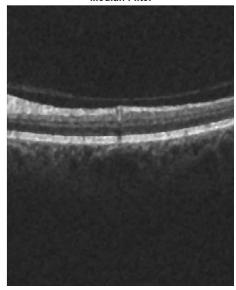
Original Image



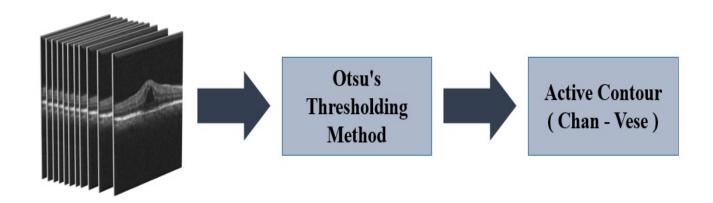
Homomorphic Wavelet Filter Analysis Band Eliminated = HH



Median Filter



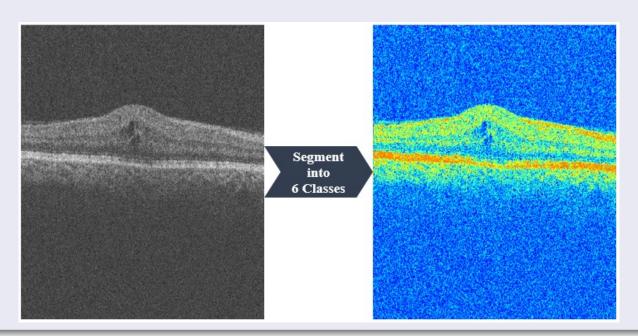






Otsu's Thresholding Method

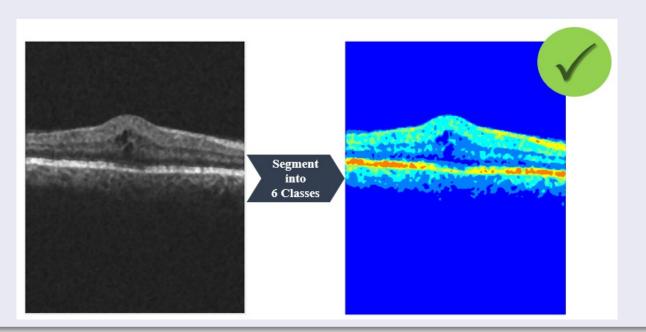
For original OCT, Extract 5 thresholds and also we have 6 classes:





Otsu's Thresholding Method

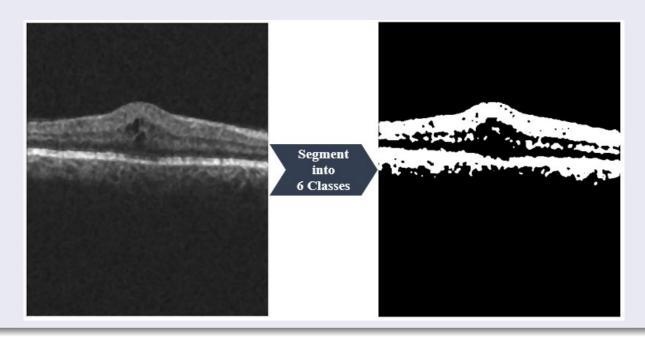
For denoised OCT, Extract 5 thresholds and also we have 6 classes:





Otsu's Thresholding Method

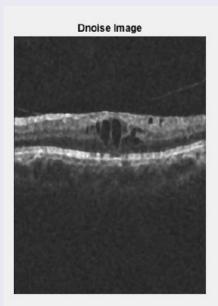
Extract all pixel in denoised OCT with intensity \geqslant threshold(2) to have binary mask with max point in Retina





Active Contour

Apply active contour with initial mask on denoised OCT:



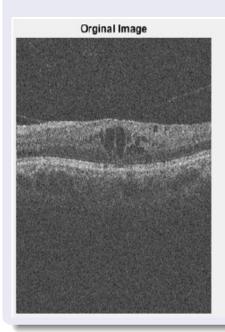






Active Contour

Apply active contour with initial mask on original OCT:



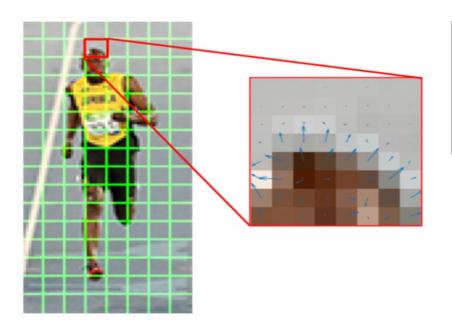




Feature Extraction



Histogram of Oriented Gradients (HOG)



2 3 4 4 9 3 4 2 2 5 11 17 13 7 9 3 4 11 21 23 27 22 17 4 6 23 99 165 135 85 32 26 2 91 155 133 136 144 152 57 28 98 196 76 38 26 60 170 51 165 60 60 27 77 85 43 136 71 13 34 23 108 27 48 110

Gradient Magnitude

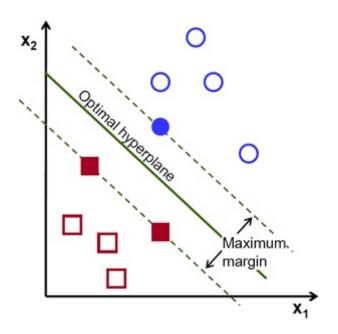
80 36 5 10 0 64 90 73 37 9 9 179 78 27 169 166 87 136 173 39 102 163 152 176 76 13 1 168 159 22 125 143 120 70 14 150 145 144 145 143 58 86 119 98 100 101 133 113 30 65 157 75 78 165 145 124 11 170 91 4 110 17 133 110

Gradient Direction

Support Vector Machine (SVM)



The operation of the SVM algorithm is based on finding the hyperplane that gives the largest minimum distance to the training examples. Twice, this distance receives the important name of margin within SVM's theory. Therefore, the optimal separating hyperplane maximizes the margin of the training data.



Nonlinear SVM



Kernel Trick:

- Polynomial Kernel
- Radial Basis
 Function Kernel

