

# 1 Review of paper on statistical denoising and 2 segmentation

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4 October 2019

## 5 1 Introduction

6 The paper proposes a six step algorithm to denoise OCT images and identify  
7 the six retinal layers. Subsequently the thickness of the layers are found out.  
8 This is an important feature for disease detection. The algorithm identifies  
9 the following six layers in the image: nerve fiber layer (NFL) inner plexiform  
10 layer (IPL), inner nuclear layer (INL), outer plexiform layer (OPL), outer nu-  
11 clear layer (ONL), photoreceptor outer segments (POS).

## 12 2 Algorithm

### 13 2.1 Step 1: Alignment of the A-scans

14 An OCT B scan consists of several A scans. The A scans constitute the columns  
15 of a B scan. Often the misalignment of the A scan columns can occur. They can  
16 be realigned with the help of a reference layer which is assumed to be straight.  
17 retinal pigment epithelium (RPE) inner boundary, bottommost contour marked  
in figure.

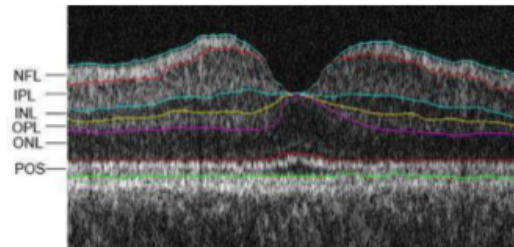


Figure 1: Layers of the eye

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19 In order to detect the edges corresponding to this layer, a two step approach  
 20 is used. A Gaussian filter with  $\sigma = 10$  is initially used and few prominent edges  
 21 are chosen. Next  $\sigma = 3$  is used and here more edges are identifiable. Only  
 22 the ones coinciding with the first pass are taken. These correspond to the RPE,  
 23 using these edges the A scans are aligned. The Gaussian filtering done here  
 24 does not affect the subsequent steps of the algorithm. It is only used for the  
 25 purpose of detecting the RPE and aligning the A scans.

## 26 2.2 Step 2: Pixel mapping

27 Certain edges between layers are more prominent than the others, for eg. NFL  
 28 to ganglion cell layer represents a change in approximate gray level value from  
 29 1800 to 1100 (12-bit data), whereas a transition from OPL to ONL translates to  
 30 a change in approximate gray level value from 1200 to 900. In order to make all  
 31 edges equally prominent, the pixel intensities are mapped using the following  
 32 functions.  $G_i(x, y) = 0.5(1 + \frac{\text{erf}(f(x, y) - t_i)}{\sqrt{2}\sigma_i^2})$  To determine  $t_i$  and  $\sigma_i$  we feed the  
 33 pixels which are between the RPE and ILM to the expectation maximisation  
 34 algorithm. It is assumed that the pixels in this region are a combination of three  
 35 Gaussian distributions. The EM algorithm provides a maximum likelihood  
 36 estimate of the means and the variances using which the values for  $\sigma_i$  and  $t_i$  are  
 37 fixed. Here we have  $i=1,2$  i.e two mapping functions  $G_1$  and  $G_2$ . the boundary  
 38 between NFL and IPL+GCL and the boundary between IS/OS and RPE was  
 39 determined with  $G_1$ , while the remaining boundaries were determined with  
 $G_2$ . The plots are as shown.

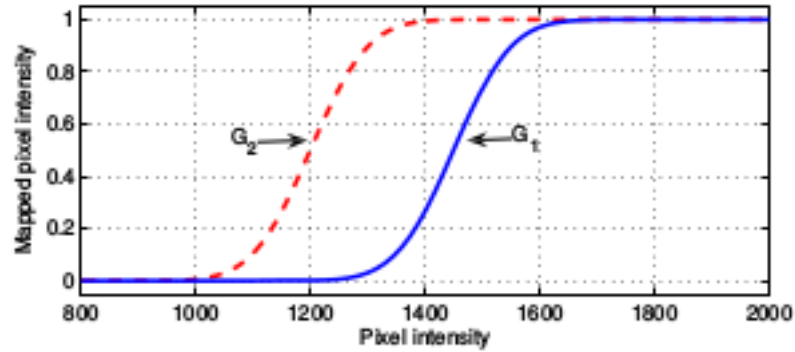


Figure 2: Mapping functions