

# Feature-Based Classification of Mouse Eye Images

Jenny Yuen, Yi Li, and Linda Shapiro

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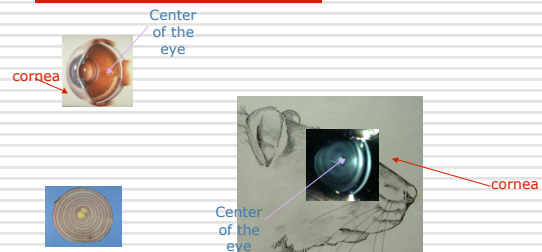
## Problem

- "...A cataract is a clouding of the eye's lens."
  - The University of Michigan Kellogg Eye Center
- There are ideas associating them to other illnesses.
- Different cataracts have different effects.
- The classification is currently manual.
- This involves a lot of time and knowledge in the area.

## Problem Statement

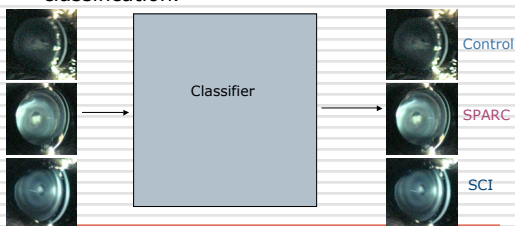
- We are given:
  - $\{C_1, C_2, \dots, C_n\}$  classes of cataracts and for each class  $C_i$ :
    - A set of mouse eye training images exhibiting the corresponding cataract.
    - A test image  $I$  that is not in the training set.
- We want:
  - A classifier that can determine the class  $C_i$  to which  $I$  belongs.

## Eye Image



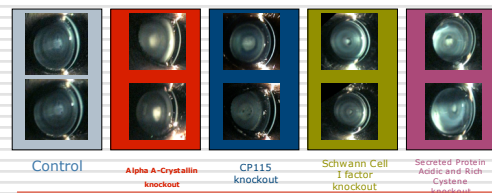
## Proposed solution

- Automate the classification.



## Training Data

- Labeled images for each category



## Approach

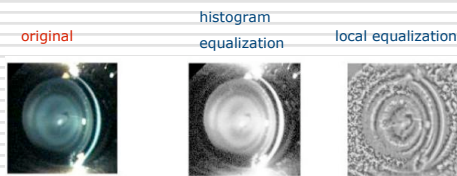
- Feature Detection
  - Ring Extraction
  - Intensity Curve
- Training

## Local/Histogram Equalization

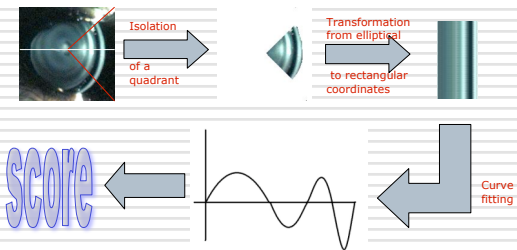
- Histogram equalization:
  - Reassigns brightness of pixels to achieve near equal numbers of distinct intensity values.
- Local equalization
  - For each pixel  $p$  in the image:
    - Take a subimage from window  $w$  around  $p$
    - $\text{LocalEq}[p] = \text{HistEq}(\text{centered at } p \text{ using } w)[p]$

## Ring Enhancement

- Use local equalization to enhance rings.



## Ring Feature Extraction



## Ring Detection (2)

- Take a horizontal axis  $x$  that passes through the middle of the image (row  $nrows/2$ ).
  - For each pixel  $p_{(x,x)}$  in row  $x$ ,
    - 1) Take the region of the image in the circular sector  $C_i$  with center  $p_{(x,x)}$  and central angle  $\theta$ .
    - 2) Map this circular sector to a rectangle  $R$  using a polar coordinate transformation.
    - 3) Represent each column of  $R$  by its average intensity.
    - 4) Compute the variance of the intensities in this curve. This is the score for  $p_{(x,x)}$ .

## Ring Detection (3)

- Let  $L_{best}$  be the curve to which the circular sector with the highest score was mapped.
- Create a feature vector with the following information from  $L_{best}$ :
  - Number, mean, and variance of maxima.
  - Number, mean, and variance of minima.
  - Mean and variance for the distance between every two consecutive maxima.
  - Mean and variance for the distance between every two consecutive minima.

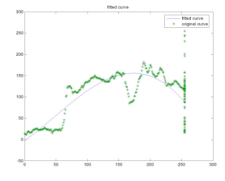
## Intensity Curve Feature

- The intensity curve of a line through the center of the eye with respect to the plane described by the image.



## Intensity Curve (2)

- We could use the coefficients of  $Fit_C$  as features to describe each class.
- Problem: Noise in the image might affect the polynomial.
- Solution: Use the Fourier Transform



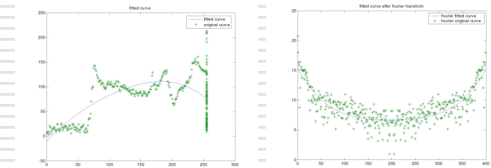
## Intensity Curve(4) – Modified Algorithm

- Let  $P = \{p | p \text{ is a pixel in the row } r/2 \text{ in the image } I\}$
- $C(x) = \{Intensity(I(r/2, x))\}$
- Apply a Fourier Transform to  $C$ :  $Fourier(C)$ .
  - Fit a polynomial on  $Fourier(C)$ . Call this polynomial fitting  $Fit_{Fourier(C)}$ .
  - Use the coefficients of  $Fit_{Fourier(C)}$  as features.

## Intensity Curve(5)

Intensity values

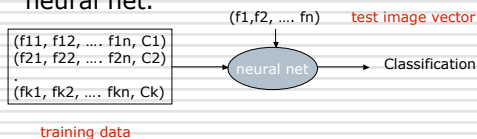
With the application of a Fast Fourier Transform



$\{coeff1_{Real}, coeff1_{Im}, coeff2_{Real}, coeff2_{Im}, coeff3_{Real}, coeff3_{Im}, coeff4_{Real}, coeff4_{Im}\}$

## Training

- Using the features from a training set, we can input this data in a neural net.

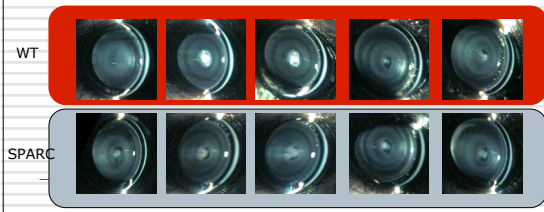


## Preliminary Results on 3 Classes

		Predicted		
		SCI	SPARC	WT
Actual	SCI	103	1	2
	SPARC	1	122	32
	WT	0	30	91
	382 images	Correctly classified	82.7 %	
	Incorrectly classified	17.2 %		
	$\chi^2$	473.44		
	p	$\leq 0.001$		

## Confusion between images

- Some of the images are so similar to each other that it is very hard, even for an expert, to determine the class.



## Conclusions and Future Work

- Performance depends on features.
- Features vary depending on the classes.
- We have:
  - Good performance identifying the SCI class.
- We want:
  - Better performance between SPARC and WT classes.
- Future work might involve human cataract classification.