Classifying Colon Cancer Colonoscopy Images Using Edge Histograms

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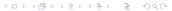


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- 3 Implementation
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- 5 Problems



Task - Overview



Task - Overview

What?

- Colon cancer colonoscopy images
- Edge histograms
- KNN: K Nearest Neighbors Classification
- Leave one patient out

Task - Overview

How?

- Preprocess images
- Perform edge detection
- 3 Extract features (e.g. edge lengths)
- 4 Compute Edge Histogram
- 5 Classify with KNN
- 6 Analyze the results



Edge Detection



Overview

- What are Edges
- Edge Detection
- First Derivative
- Second Derivative
- Canny Edge Detection



Edges:

Edges are pixels, in which the image intensity function changes its magnitude



(a) Original Image

(b) Image after Edge Detection

Figure: Edge Detection using Canny

Edge Detection:

Almost every Edge Detector uses either the first derivative or the second derivative of the intensity function.

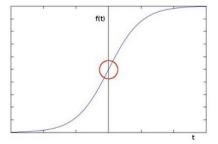


Figure: Intensity function



First Derivative:

Sobel-, Roberts-, Robinson-, Kirsch-Operator

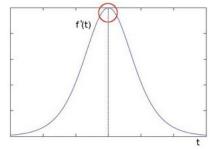


Figure: Intensity function - First derivative



Second Derivative:

Laplace-, Mexican-Hat-Operator

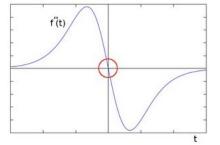


Figure: Intensity function - Second derivative



Canny Edge Detection:

- Low error rate
- Good localization
- Minimal response



Steps:

- Filter out noise using Gaussian filter
- Find the intensity gradient using Sobel-Operator $G = \sqrt{G_x^2 + G_y^2}$ or $G = |G_x| + |G_y|$
- 3 Non-maximum suppression
- 4 Hysteresis

Implementation



Overview

- Development and Frameworks
- Image Enhancement
- Edge Detection
- Edge Histograms
- Edge Lengths
- Edge Orientation
- Image Classification



Development and Frameworks

Developed using:

- Java 8
- OpenCV for Java
- Eclipse Neon



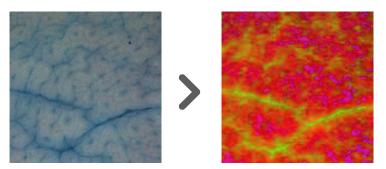
Image Enhancement and Conversion

- Color Space Conversion
- Normalization
- CLAHE
- All part of OpenCV



Image Enhancement and Conversion

RGB to HSV



Imgproc.cvtColor(srcMatrix, dstMatrix, colorSpace)
colorSpace: e.g. Imgproc.COLOR RGB2HSV



Image Enhancement and Conversion

Normalization:

```
Core.normalize(srcMatrix, dstMatrix, 255, 0,
Core.NORM_MINMAX);
```

CLAHE:

```
Mat channel = new Mat();
Core.extractChannel(matrix, channel, i);
CLAHE clahe = Imgproc.createCLAHE();
clahe.apply(channel, channel);
Core.insertChannel(channel, matrix, i);
```



Image Enhancement and Conversion





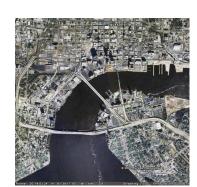




Image Enhancement and Conversion



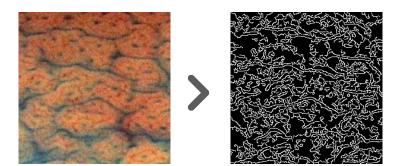


Edge Detection

- Grayscale Conversion
- Canny Edge Detector



Edge Detection



Imgproc.Canny(srcMatrix, dstMatrix,
lowThresh, highThresh);



Edge Histograms

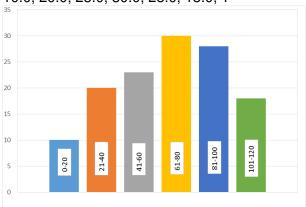
Definition

- Characteristics of an image e.g. edge lengths
- Partition characteristic attributes into bins
- In our case: edge lengths & orientations
- Length: image has5 edges of length 0 20 pixels20 edges of length 100 120 pixels



Edge Histograms

Simple Example: 10.0, 20.0, 23.0, 30.0, 28.0, 18.0, 1



Edge Histograms

- Histogram data for each example image collected in a hist-file.
- Specify Category. Here: Cancer stage
- Example:

```
210.0,3.0,170.0,142.0,126.0,93.0,32.0,16.0,1
192.0,2.0,181.0,139.0,119.0,87.0,32.0,17.0,2
143.0,1.0,172.0,147.0,128.0,91.0,30.0,16.0,3
```



Edge Lengths

Prequisites

- Image with detected edges
- Edges white
- Rest black
- Not entirely given → threshold set at grayscale 200

Edge Lengths

Algorithm

Iterate over all pixels

- Check if pixel is white → new edge found
- **2** Check immediate neighbours: if white \rightarrow add pixel to edge
- Follow white path until no more connected white pixels
- 4 Add all passed pixels to collection of used pixels
- 5 Add 1 to bin with detected length
- 6 Continue iterating and start at 1



Edge Lengths

Algorithm

```
function measureEdge(pixel) {
    length = 1
    for each surrounding pixel p:
        if p == white
            length += measureEdge(p)
    return length
}
```

Edge Orientation

- Requires edge detected image
- Use sobel operator to detect edges
- Extract edge orientations from image (OpenCV)
- Partition edges into bins
- Bin content: pixels part of edge with certain orientation
- Bin: range of angles



Image Classification

- Classify all example images → create file with histograms
- 2 Create list of feature vectors
- 3 Enhance image to be classified
- 4 Create feature vector
- 5 KNN: Compare new vector all vectors in list
 - \rightarrow euclidean distance
- Select K vectors with smallest distance
- Classification: category found most often



Paramter Optimization

Difference between a good and barely functional program

- Selection of input images
- Thresholds for edge detection
- Edge Lengths: number of bins, range of lengths in bins
- Weights of features: all equally significant Lengths: per edge
 Orientation: per pixel



Results



Evaluation

Normalizing

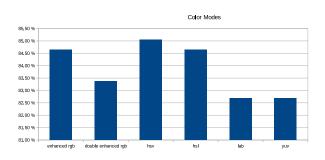
- Two features: Length, Orientation
- Histograms results are normalized
- from 0.0 to 1.0
- 0.0 is the lowest occurring value (usually 0.0 anyway)
- 1.0 is the highest occuring value

Results

- Total % of correct classified images
- Class 2 weights stronger than class 3 (fewer images)



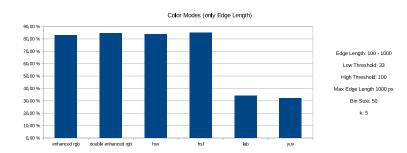
Colormodes



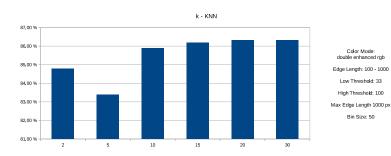
Edge Length: 100 - 1000 Low Threshold: 33 High Threshold: 100 Max Edge Length 1000 px Bin Size (Length): 50 Bin Size (Orientation): 20



Colormodes (without Edge Orientation)

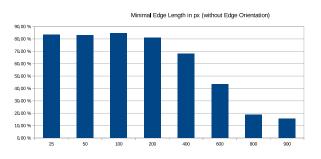


k - KNN



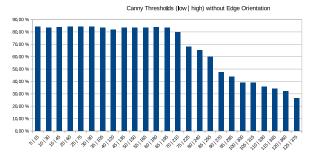


Edge Length (without Edge Orientation)



Color Mode: double enhanced rgb Low Threshold: 33 High Threshold: 100 Max Edge Length 1000 px Bin Size: 50 k (KNN): 5

Canny Thresholds (without Edge Orientation)



Color Mode: double enhanced rgb Edge Length: 100-1000 px Bin Size: 50 k (KNN): 5



Problems



Problems

Encountered Problems

- Understanding OpenCV (bad java documentation)
- Fine tuning of parameters (unlimited combinations)
- Different parameter sets for different images
- Normalizing features to weight equally
- Can get over 80% correctness without recognizing class 3

Thank You!

