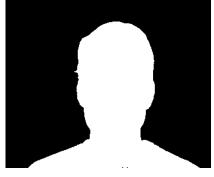
Image Segmentation

- Gray-level thresholding
- Supervised vs. unsupervised thresholding
- Binarization using Otsu's method
- Locally adaptive thresholding
- Maximally stable extremal regions
- Color-based segmentation
- Region labeling and counting
- Region moments

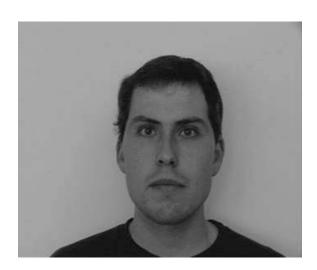




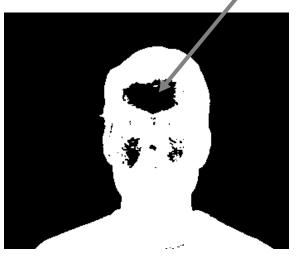




Gray-level thresholding



Original image Peter f[x,y]



Thresholded Peter m[x,y]

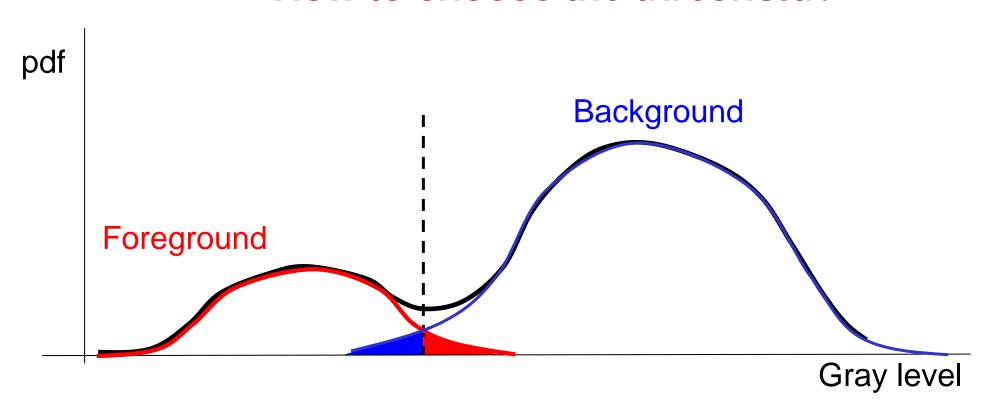


How can holes be filled?

 $f[x,y] \cdot m[x,y]$



How to choose the threshold?



Unsupervised thresholding

 Idea: find threshold T that minimizes within-class variance of both foreground and background (same as k-means)

$$\sigma_{within}^{2}\left(T\right) = \frac{N_{Fgrnd}\left(T\right)}{N}\sigma_{Fgrnd}^{2}\left(T\right) + \frac{N_{Bgrnd}\left(T\right)}{N}\sigma_{Bgrnd}^{2}\left(T\right)$$

Equivalently, maximize between-class variance

$$\begin{split} &\sigma_{between}^{2}\left(T\right) = \sigma^{2} - \sigma_{within}^{2}\left(T\right) \\ &= \left(\frac{1}{N}\sum_{x,y}f^{2}\left[x,y\right] - \mu^{2}\right) - \frac{N_{Fgrd}}{N}\left(\frac{1}{N_{Fgrd}}\sum_{x,y\in Fgrnd}f^{2}\left[x,y\right] - \mu_{Fgrnd}^{2}\right) - \frac{N_{Bgrnd}}{N}\left(\frac{1}{N_{Bgrnd}}\sum_{x,y\in Bgrnd}f^{2}\left[x,y\right] - \mu_{Bgrnd}^{2}\right) \\ &= -\mu^{2} + \frac{N_{Fgrnd}}{N}\mu_{Fgrnd}^{2} + \frac{N_{Bgrnd}}{N}\mu_{Bgrnd}^{2} = \frac{N_{Fgrnd}}{N}\left(\mu_{Fgrnd} - \mu\right)^{2} + \frac{N_{Bgrnd}}{N}\left(\mu_{Bgrnd} - \mu\right)^{2} \\ &= \frac{N_{Fgrnd}\left(T\right) \cdot N_{Bgrnd}\left(T\right)}{N^{2}}\left(\mu_{Fgrnd}\left(T\right) - \mu_{Bgrnd}\left(T\right)\right)^{2} \end{split}$$

[Otsu, 1979]

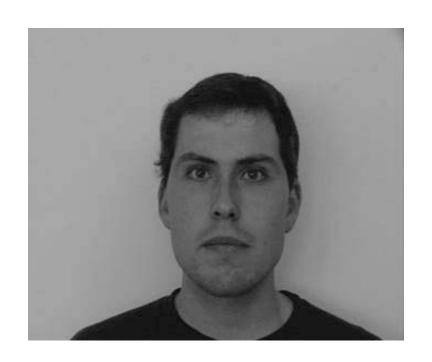
Algorithm: Search for threshold T to maximize

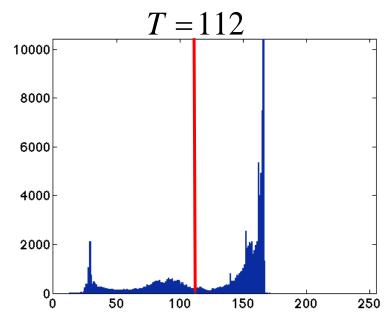
$$\sigma_{between}^{2}\left(T\right) = \frac{N_{Fgrnd}\left(T\right) \cdot N_{Bgrnd}\left(T\right)}{N^{2}} \left(\mu_{Fgrnd}\left(T\right) - \mu_{Bgrnd}\left(T\right)\right)^{2}$$

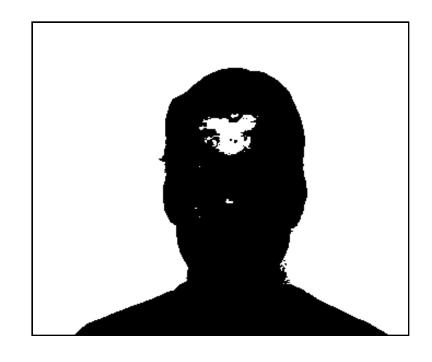
Useful recursion for sweeping T across histogram:

$$\begin{split} N_{Fgrnd}\left(T+1\right) &= N_{Fgrnd}\left(T\right) + n_{T} \\ N_{Bgrnd}\left(T+1\right) &= N_{Bgrnd}\left(T\right) - n_{T} \\ \mu_{Fgrnd}\left(T+1\right) &= \frac{\mu_{Fgrnd}\left(T\right)N_{Fgrnd}\left(T\right) + n_{T}T}{N_{Fgrnd}\left(T+1\right)} \\ \mu_{Bgrnd}\left(T+1\right) &= \frac{\mu_{Bgrnd}\left(T\right)N_{Bgrnd}\left(T\right) - n_{T}T}{N_{Fgrnd}\left(T+1\right)} \end{split}$$

[Otsu, 1979]

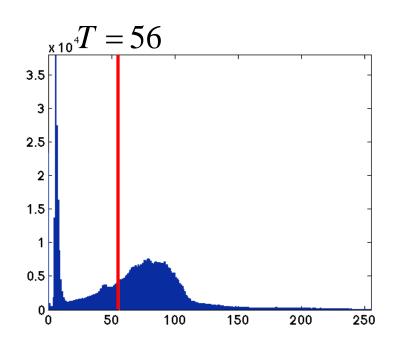
















The Stanford Daily

Tuesday, September 18, 2012 ◆ 13

Stanford defensive lineman Josh Mauro put the pressure on USC's

Handing out the USC game balls

Stepfan: Taylor: It all starts and ends with Stantford's workhorse. Taylor and excepting you could ask for and more against USC. He garne-tying touchdown on the ground, another score through the air and the consistent ground-and. The middle from the defensive line that kept Matt Barkley rattled. It may be a first mark the constraint of the constraint o pound to wear down the Trojans at in support, but Mauro went above the end. His 213 total yards of and beyond the call of duty to help

had fans on both sides forgetting Please see AWARDS, page 15

The winding road ahead

FOOTBALL EDITOR

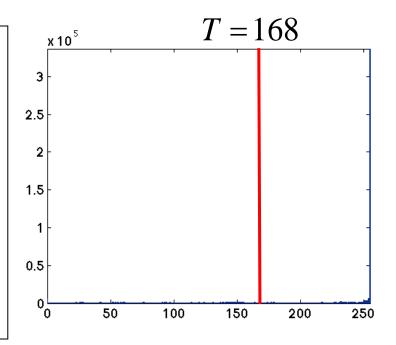
Andrew Luck may be gone, but with Saturday night's win over USC, the Stanford Cardinal put itself in position to be supported by the standard put itself in position to by number 12 You heard right though there's pleaty of work left to do, this 2012 Standon on small, in USC s delense is the swated opportunities that the standard of the standard of

schedule has the potential for a magical appear. The Rose Bowl becomes truly remarkable for Stanford to magical appear. The worst-case scenario. To get there, Stanford almost certainly late point, there are three paths, the rest of the season could take the promised Land, a berth in the BCS National Championship Game. In all Bikelhood, because Stanford is not named Alabama or LSU, the Cardinal will have to win out to carn a trip to South to Crepon and a potential remarked to win out to carn a trip to South Coregon and a potential remarked against USC for the Pac-12 title.

about returning to the site of its everyone on the schedule except

longing interest pinalty on whose the company of the champing of the company of t

schedule has the potential for a Pac-12, the Rose Bowl becomes truly remarkable for Stanford to



The Stanford Daily



Stanford defensive lineman Josh Mauro put the pressure on USC's

Handing out the USC game balls

Seaths Thefor: It all starts and ends with Stantard's workhores, and the start of t pound to wear down the Trojans at the end. His 213 total yards of and beyond the call of duty to help

nd saw most of his action at nosc

Please see AWARDS, page 15

The winding road ahead

FOURSALL ENTERM about returning to the site of its correspond to the school and t

For Stanford to even think the last two seasons, beating about returning to the site of its everyone on the schedule except

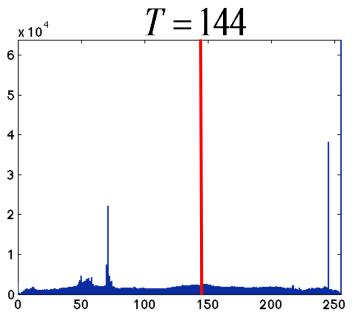
Tuesday, September 18, 2012 ◆ 13

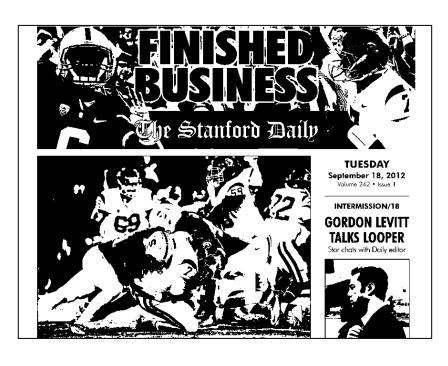
eth to do, this '2012' Standard cleam showed that it is capable of playing at a national change construction of the standard contribution of the standard contribution of the standard cleam played to the cleam pl

schedule åre games at Notre
Dame. Ortgom and UCLA, all
of whom are currently zanked in
the top 31. The next two games,
at all
of whom are currently zanked in
the top 31. The next two games,
against Arzona, are on
pathorers either. And as
standord has a showed topranked opponents in years past,
the path tanybling but straight
schedule however,
the path tanybling but straight
schedule has the potential for a
magical space.
From Standord's current vantage point, there are three paths
the rest of the season could take
Promised Land, a berth in the
BCS National Championship
Game Is all Biellhood, feecuse
Sandord is not named Alabama
to with of User at 1719 to South
Beech, including wins at No. 3
Oregon and a potential remark
against U.S. for the Pac-12 title.





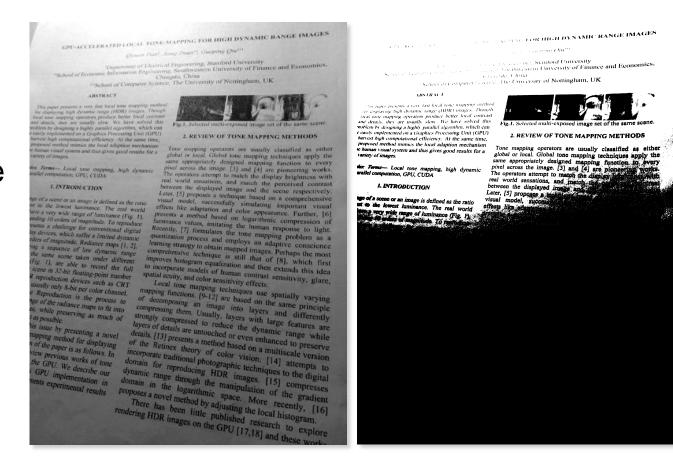






Sometimes, a global threshold does not work

Original image

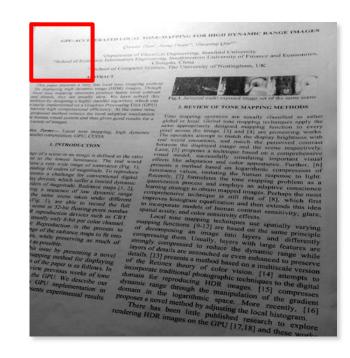


Thresholded with Otsu's Method



Locally adaptive thresholding

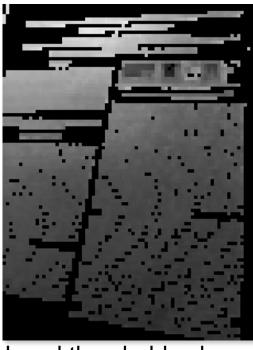
- Slide a window over the image
- For each window position, decide whether to perform thresholding
 - Thresholding should not be performed in uniform areas
 - Use variance or other suitable criterion
- Non-uniform areas: apply Otsu's method (based on local histogram)
- Uniform areas: classify the entire area as foreground or background based on mean value



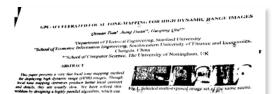
Locally adaptive thresholding (example)



Non-uniform areas



Local threshold values



in Ferms- Local tone mapping, high dynamic

I. INTRODUCTION

the of sector and many is defined as the ratio of a the lowest faminises. The real world was the first faminises. The real world was the sector as the lowest faminises. The reproductive faminises of the lowest families families for the lowest families families families families families families families families for the lowest families fam

is issue by presenting a novel mapping method for displaying no the paper is as follows. In view previous works of tone the GPU, We describe our of GPU implementation in sents experimental.

To response operators are manify classified as either to fee global on been classified to making the company per changes capity the same appropriately designed mapping insultion to every per common the many [2] and [4] are phoneering works.

The operators accessified an attach the perceived contrast in the operators accessified match the perceived contrast in the operators accessified in the scene respectively. Lace, [5] proposes a technique better comprehensive wond makel; successfully immulating comprehensive wond makel; successfully immulating comprehensive wind makel; successfully immulating comprehensive for the comprehensive consideration and color appearance. Further, [6] parameters are method based on logarithmic comprehensive changes are made of the comprehensive changes in the c

spenal acuty, and color sensitivity effects.

CRI Local fore mapping functions [9-12] are based on the same principle of the foreign and the store of compressing them. [9-12] are based on the same principle of the store of compressing them. [9-12] are based on the same principle of the store of compressing them. [9-12] are based on the same principle of the store of

Locally thresholded result



Maximally stable extremal regions

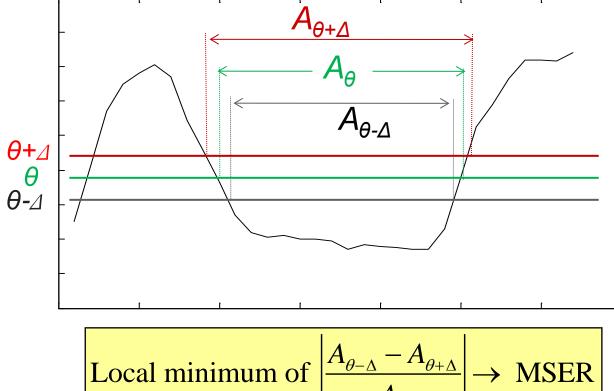
- Extremal region: any connected region in an image with all pixel values above (or below) a threshold
- Observations:
 - Nested extremal regions result when the threshold is successively raised (or lowered).
 - The nested extremal regions form a "component tree."
- Key idea: choose thresholds θ such that the resulting bright (or dark) extremal regions are nearly constant when these thresholds are perturbed by $\pm 1/2$

→ "maximally stable" extremal regions (MSER)

[Matas, Chum, Urba, Pajdla, 2002]

MSERs: illustration

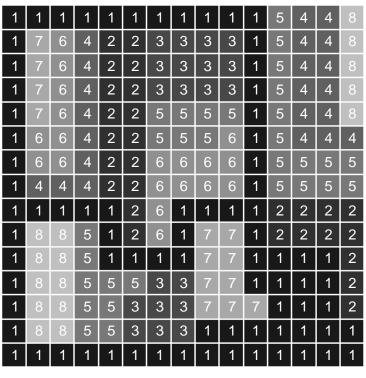






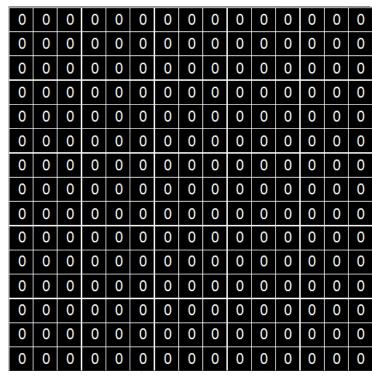
[Matas, Chum, Urba, Pajdla, 2002]

Level sets of an image



f[x, y]

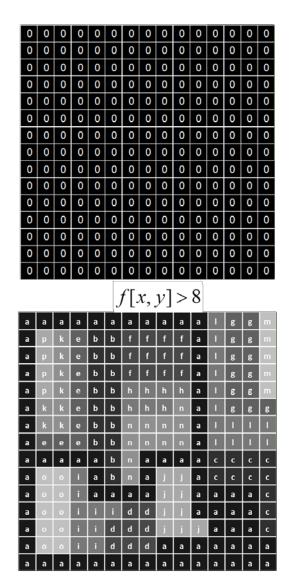
Image

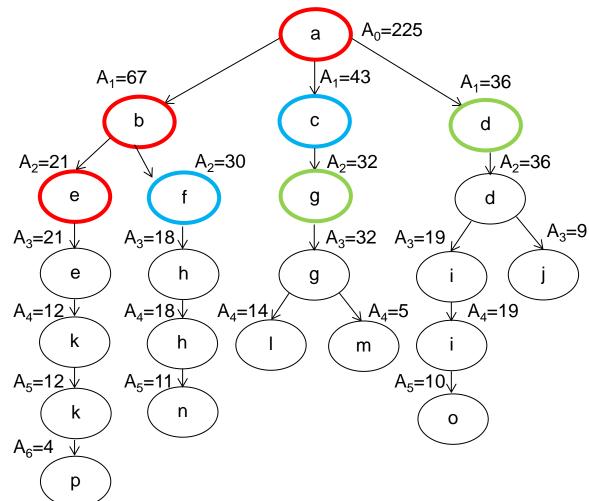


f[x, y] > 8

Level Set

Component tree of an image

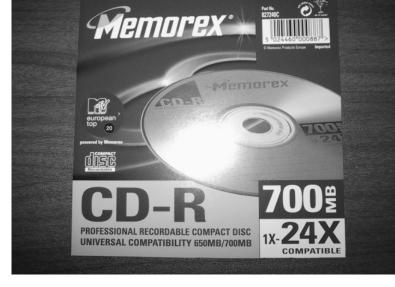




Local minima of sequence
$$\left| \frac{A_{\theta-\Delta} - A_{\theta+\Delta}}{A_{\theta}} \right|$$
 $\theta \vdash \Delta, \ \Delta + 1, \ \mathsf{K} \ \rightarrow \ \mathsf{MSERs}$

MSER: examples







Dark MSERs, ∆=15

Original image

Bright MSERs, Δ =15

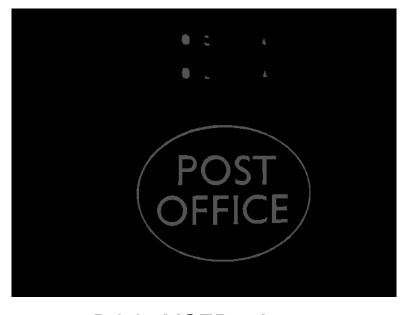
MSER: examples



Dark MSERs, Δ=15

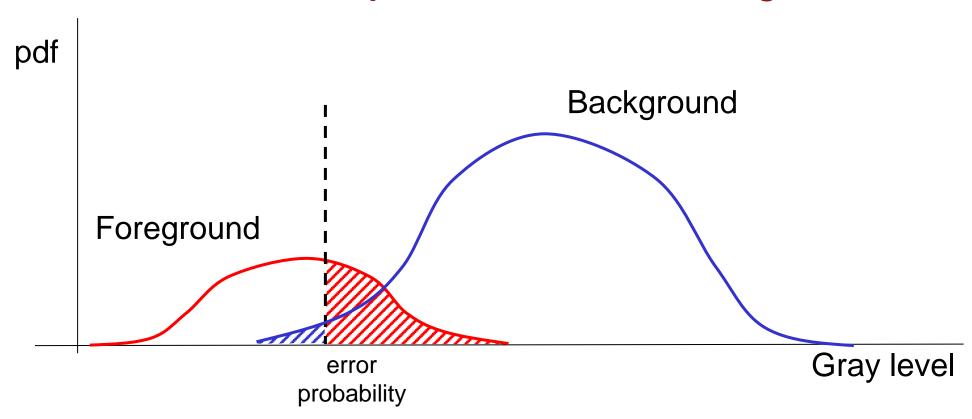


Original image

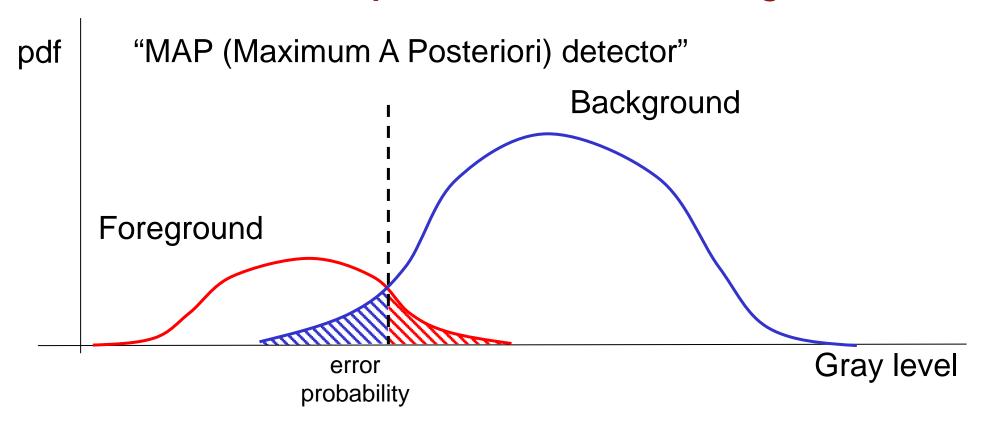


Bright MSERs, Δ =15

Supervised thresholding



Supervised thresholding

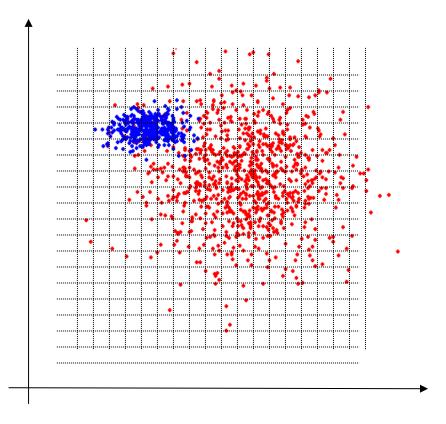


If errors BG→FG and FG→BG are associated with different costs: "Bayes minimum risk detector" is optimal.

Multidimensional MAP detector

Training

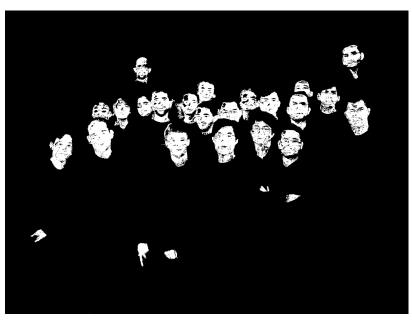
- Provide labelled set of training data
- Subdivide n-dimensional space into small bins
- Count frequency of occurrence for each bin and class in training set, label bin with most probable class
- (Propagate class labels to empty bins)
- For test data: identify bin, look up the most probable class



MAP detector in RGB-space



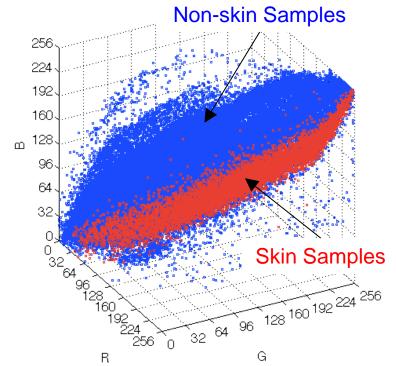
Original image



Skin color detector



Five training images



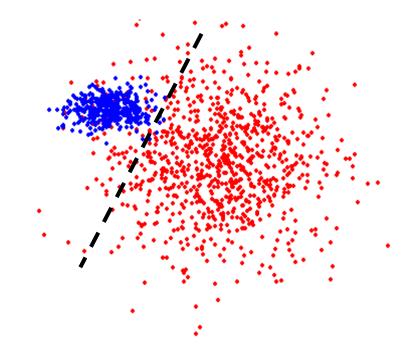


Linear discriminant function

■ To segment image with n components f_i , i=1,2,...,n into two classes, perform test

$$\sum_{i} w_i f_i + w_0 \ge 0 ?$$

- Categories are separated by hyperplane in n-space
- Numerous techniques to determine weights w_i , i=0,1,2,...,n, see, e.g., [Duda, Hart, Stork, 2001]
- Can be extended to the intersection of several linear discrimant functions
- Can be extended to multiple classes

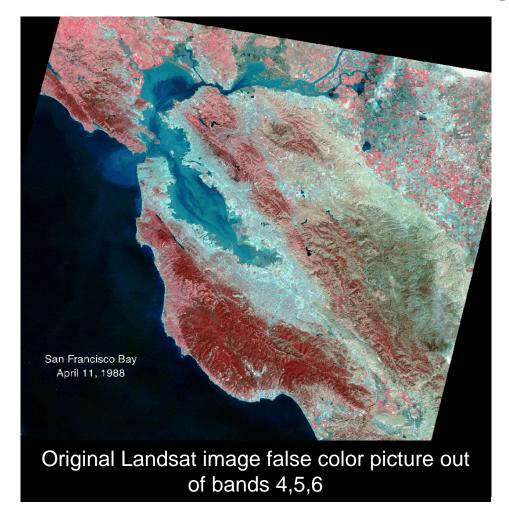


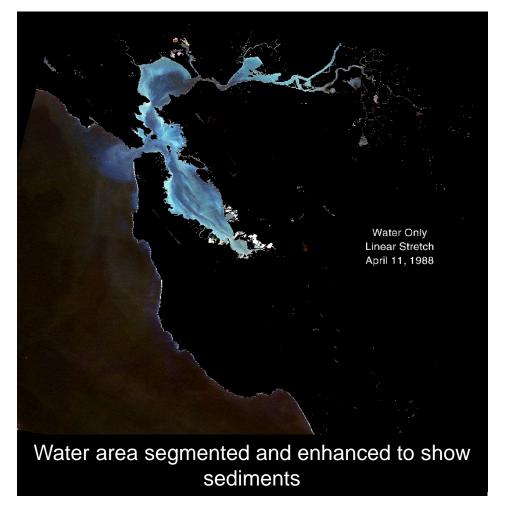
Chroma keying





Landsat image processing

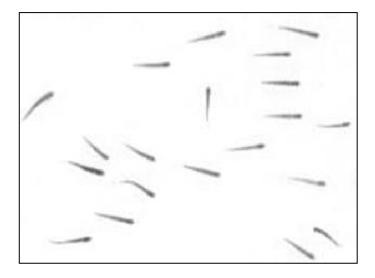




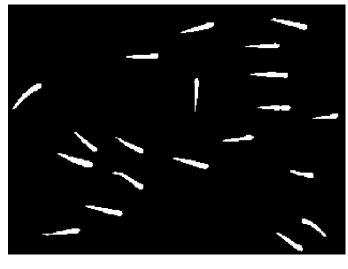
Source: US Geological Survey USGS, http://sfbay.wr.usgs.gov/

Region labeling and counting

How many fish in this picture?



Original Fish image



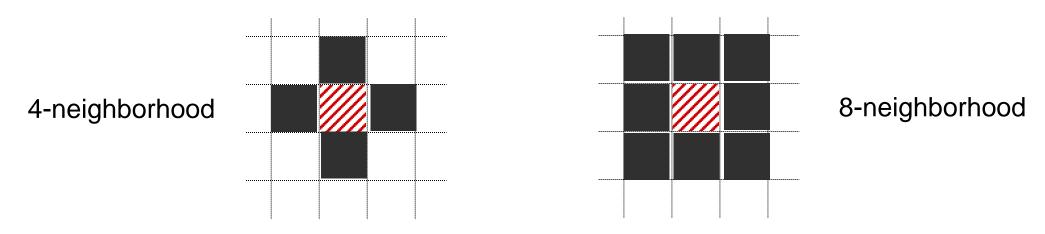
after thresholding

- Which pixels belong to the same object (region labeling)?
- How large is each object (region counting)?



4-connected and 8-connected neighborhoods

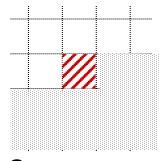
Definition: a region is a set of pixels, where each pixel can be reached from any other pixel in the region by a finite number of steps, with each step starting at a pixel and ending in the neighborhood of the pixel.



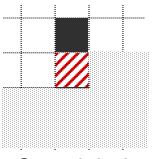
 Typically, either definition leads to the same regions, except when a region is only connected across diagonally adjacent pixels.

Region labeling algorithm (4-neighborhood)

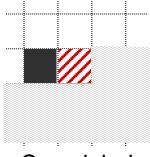
- Loop through all pixels f[x,y], left to right, top to bottom
- If f[x,y]=0, do nothing.
- If f(x,y)=1, distinguish 4 cases



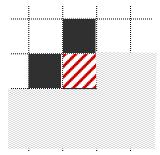
Generate new region label



Copy label from above



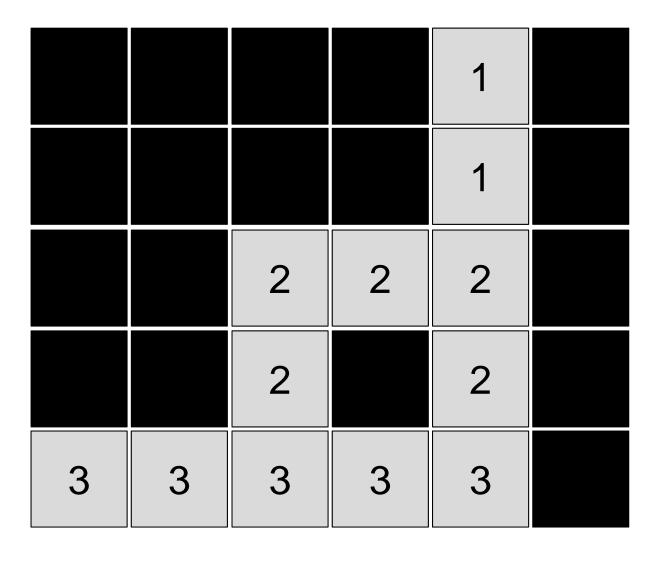
Copy label from the left

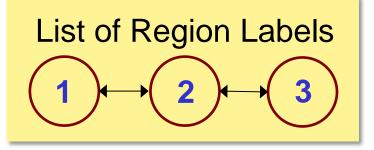


Copy label from the left. If labels above and to the left are different, store equivalence.

 Second pass through image to replace equivalent label by the same label.

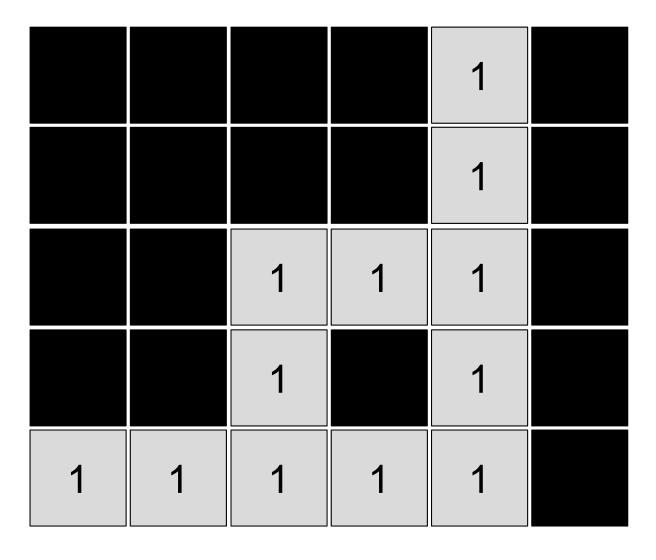
Region labeling example (4-neighborhood)





All three labels are equivalent, so merge into single label.

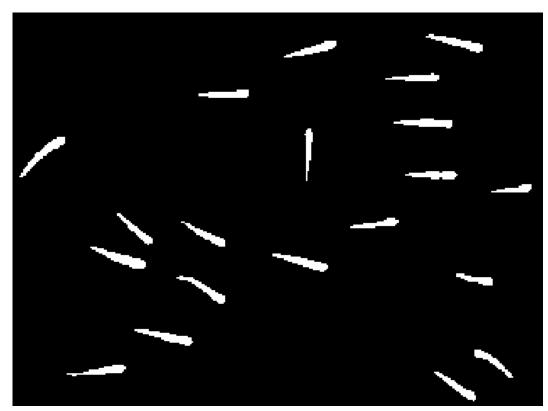
Region labeling example (4-neighborhood)



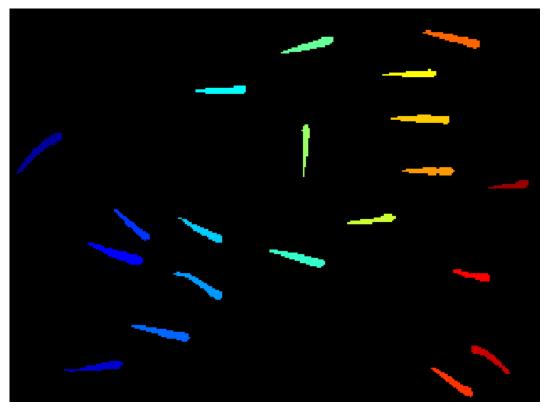
List of Region Labels

1

Example: region labeling



Thresholded image



20 labeled regions



Region counting algorithm

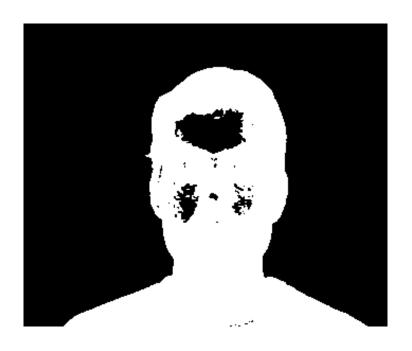
- Measures the size of each region
- Initialize counter[label]=0 for all label
- Loop through all pixels f[x,y], left to right, top to bottom
 - If f[x,y]=0, do nothing.
 - If f(x,y)=1, increment counter[label(x,y)]

Small region removal

- Loop through all pixels f[x,y], left to right, top to bottom
 - If f(x,y)=0, do nothing.
 - If f[x,y]=1 and counter[label[x,y]] < S, set f[x,y]=0
- Removes all regions smaller than S pixels

Hole filling as dual to small region removal

Mask with holes



After NOT operation, (background) region labeling, small region removal, and second NOT operation





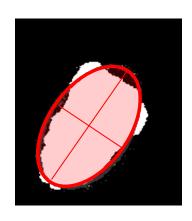
Region moments

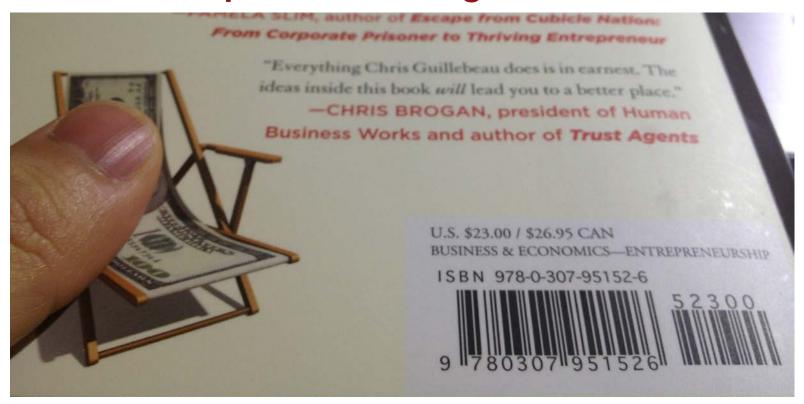
■ Raw moments
$$M_{pq} = \sum_{x,y \in \text{Region}} x^p y^q$$

Central moments

$$\mu_{pq} = \sum_{x,y \in \text{Region}} (x - \overline{x})^p (y - \overline{y})^q \quad \text{with } \overline{x} = \frac{M_{10}}{M_{00}} \text{ and } \overline{y} = \frac{M_{01}}{M_{00}}$$

Region orientation and eccentricity: calculate eigenvectors of covariance matrix $\begin{bmatrix} \mu_{20} & \mu_{11} \\ \mu_{11} & \mu_{02} \end{bmatrix}$

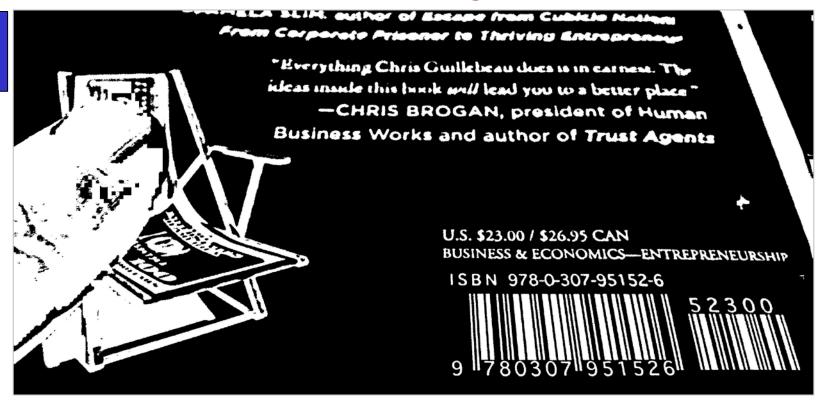




Original Image



Locally adaptive thresholding





Locally adaptive thresholding

Filtering by eccentricity

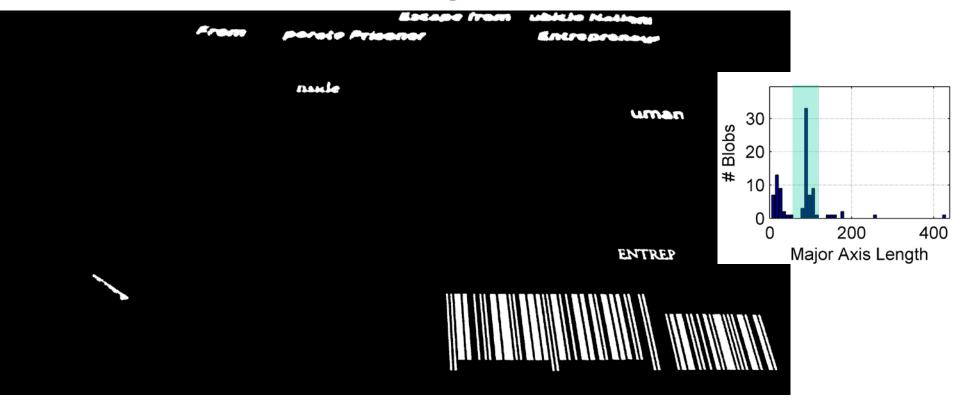




Locally adaptive thresholding

Filtering by eccentricity

Filtering by major axis length





Locally adaptive thresholding

Filtering by eccentricity

Filtering by major axis length

Filtering by orientation



