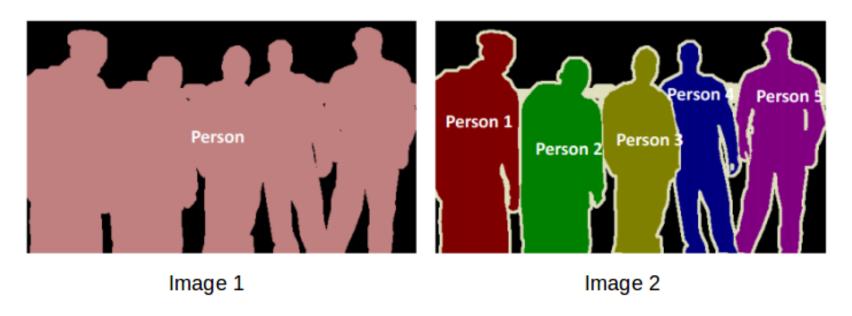
Image Segmentation

What is Image Segmentation?

- We can divide or partition the image into various parts called segments
- It's not a great idea to process the entire image at the same time as there will be regions in the image which do not contain any information
- By dividing the image into segments, we can make use of the important segments for processing the image
- Image segmentation creates a pixel-wise mask for each object in the image
 - This technique gives us a granular understanding of the object(s) in the image

Different types of image segmentation



- Image 1: Semantic segmentation
 - Every pixel belongs to a particular class (either background or person)
- Image 2: Instance segmentation
 - Different objects of the same class have different colors (Person 1 as red, Person 2 as green, background as black, etc.)

Threshold segmentation

- One simple way to segment different objects could be to use their pixel values
 - The pixel values will be different for the objects and the image's background if there's a sharp contrast between them.
- Threshold segmentation: the pixel values falling below or above the threshold can be classified accordingly (as an object or the background)

Threshold segmentation

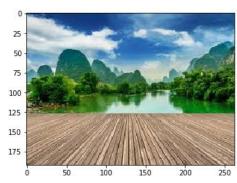
• Global threshold: If we want to divide the image into two regions (object and background), we define a single threshold value

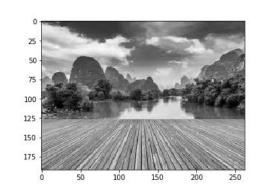
• Local threshold: If we have multiple objects along with the background, we must define multiple thresholds

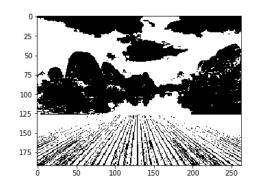
There are four different segments in the above image.

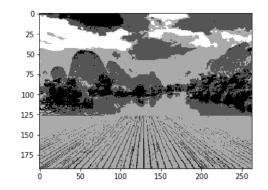
We can set different threshold values and check how the segments are made.



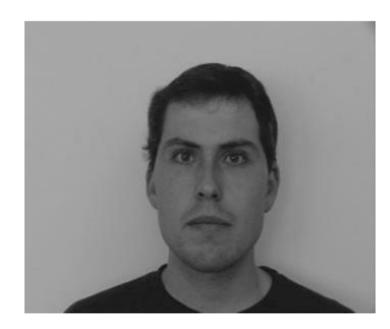








Gray-level thresholding

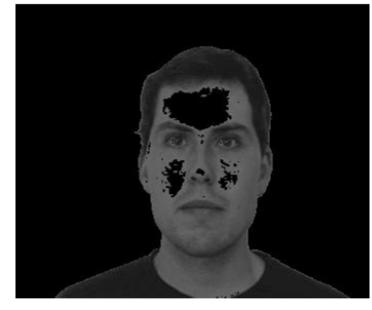


Original image Peter f[x,y]



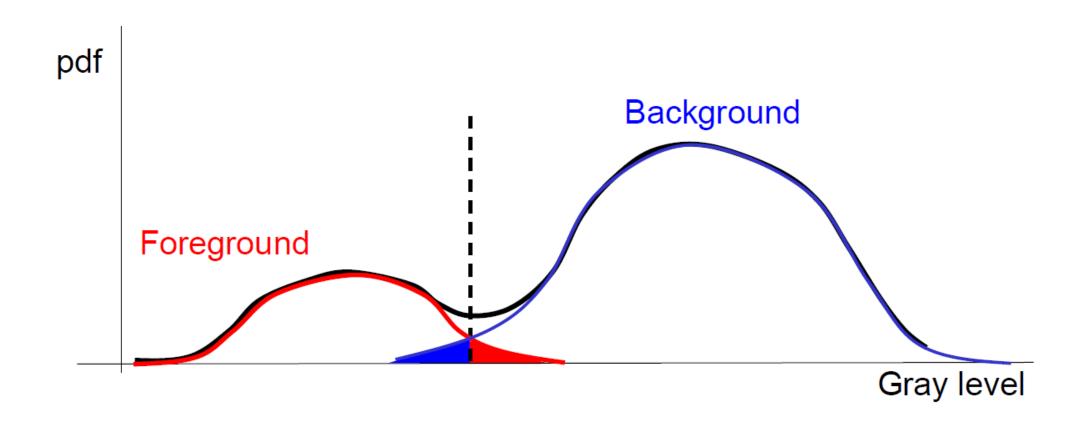
Thresholded Peter m[x,y]

How can holes be filled?



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

How to choose the threshold?



Otsu's Thresholding Method

- Based on a very simple idea: Find the threshold that minimizes the weighted within-class variance
- This turns out to be the same as maximizing the between-class variance
- Operates directly on the gray level histogram [e.g. 256 numbers, P(i)], so it's fast (once the histogram is computed)

Otsu: Assumptions

- Histogram (and the image) are bimodal.
- No use of spatial coherence, nor any other notion of object structure.
- Assumes stationary statistics, but can be modified to be locally adaptive
- Assumes uniform illumination (implicitly), so the bimodal brightness behavior arises from object appearance differences only

The weighted within-class variance is:

$$\sigma_w^2(t) = q_1(t)\sigma_1^2(t) + q_2(t)\sigma_2^2(t)$$

Where the class probabilities are estimated as:

$$q_1(t) = \sum_{i=1}^{t} P(i)$$
 $q_2(t) = \sum_{i=t+1}^{l} P(i)$

And the class means are given by:

$$\mu_1(t) = \sum_{i=1}^{t} \frac{iP(i)}{q_1(t)} \qquad \mu_2(t) = \sum_{i=t+1}^{t} \frac{iP(i)}{q_2(t)}$$

Finally, the individual class variances are:

$$\sigma_1^2(t) = \sum_{i=1}^t [i - \mu_1(t)]^2 \frac{P(i)}{q_1(t)}$$

$$\sigma_2^2(t) = \sum_{i=t+1}^{I} [i - \mu_2(t)]^2 \frac{P(i)}{q_2(t)}$$

Now, we could actually stop here. All we need to do is just run through the full range of t values [1,256] and pick the value that minimizes $\sigma_w^2(t)$.

But the relationship between the within-class and betweenclass variances can be exploited to generate a recursion relation that permits a much faster calculation.

Between/Within/Total Variance

- The basic idea is that the total variance does not depend on threshold (obviously)
- For any given threshold, the total variance is the sum of the withinclass variances (weighted) and the *between class variance*, which is the sum of weighted squared distances between the class means and the grand mean

After some algebra, we can express the total variance as...

$$\sigma^{2} = \sigma_{w}^{2}(t) + q_{1}(t)[1 - q_{1}(t)][\mu_{1}(t) - \mu_{2}(t)]^{2}$$
Within-class, from before

Between-class, $\sigma_{B}^{2}(t)$

Since the total is constant and independent of *t*, the effect of changing the threshold is merely to move the contributions of the two terms back and forth.

So, minimizing the within-class variance is the same as maximizing the between-class variance.

The nice thing about this is that we can compute the quantities in $\sigma_B^2(t)$ recursively as we run through the range of t values.

Finally...

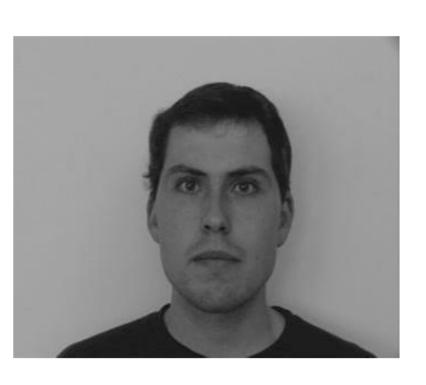
Initialization...
$$q_1(1) = P(1); \mu_1(0) = 0$$

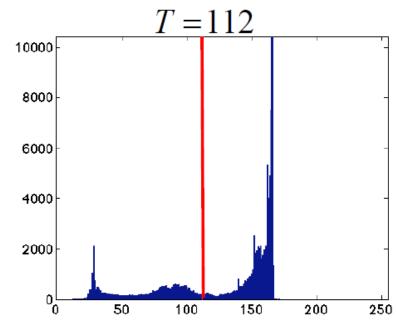
Recursion...

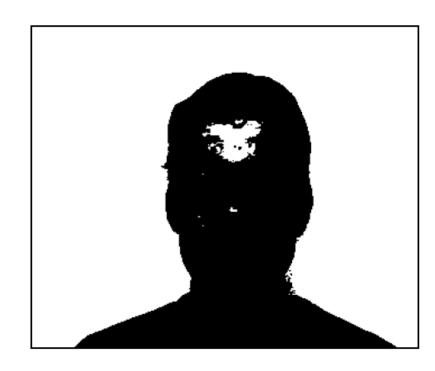
$$q_1(t+1) = q_1(t) + P(t+1)$$

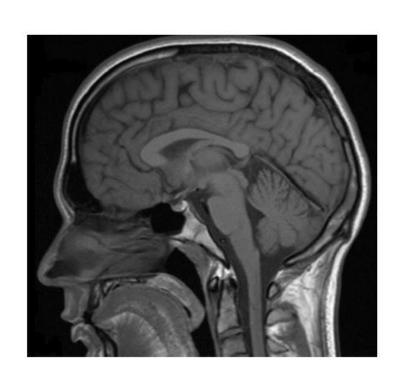
$$\mu_1(t+1) = \frac{q_1(t)\mu_1(t) + (t+1)P(t+1)}{q_1(t+1)}$$

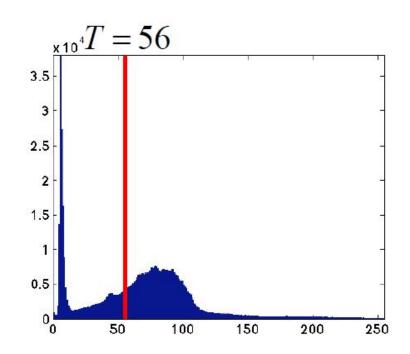
$$\mu_2(t+1) = \frac{\mu - q_1(t+1)\mu_1(t+1)}{1 - q_1(t+1)}$$

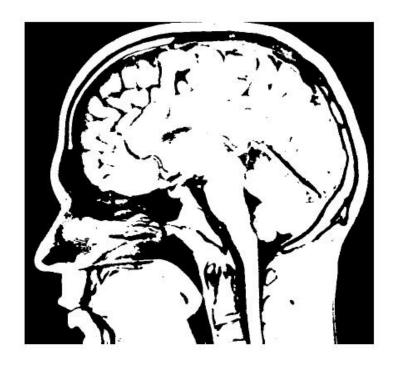












The Stanford Baily

The winding road ahead



nd defensive lineman Josh Mauro put the pressure on USC's deferred correlately short down Brokley and his traded wide receivers.

Handing out the USC game balls

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Stephen Theface it oil some and conductive the conductive of the action of the ball safe for and more against USC 16; backup center Cyron 16-44.

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but with Saturday night's win have to build off his fourth BCS ranking and an over USC, the Stanford quarter success against USC to bid to BCS game. Cardinal put itself in position to play at a higher level consistentschieve beyond the puth puved by number 12. You heard right: though there's plenty of work good enough to win a champi-laft to do, this 2012 Stanford tended to the champion of the wind on the second tended to the second that it is eagable of leading the champion of team showed that it is capable of playing at a national champi-

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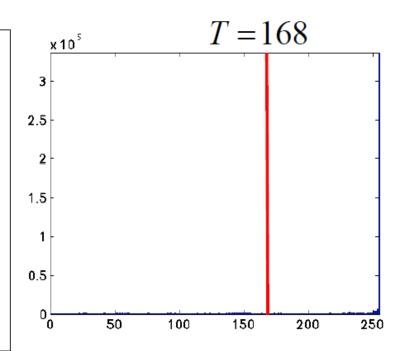
fire worst-case scenario. To get be in the position it is in now. No From Stanford's current vanfiere, Stanford almost certainly Toby Gerhart, no Jim Harbaugh From Smillford's Current var-lage point, there are three publis has to winn of Crego on Nos. 17, and no Andrews. Lack; pail Sob the rest of the season could take. Sone of the toughest tasks in all of Door Number One leads to The Promised Land, a berth in the If Smillford curr's beat the Proteined Lind, a berth in the IRS National Champsoning in Back a BCS board creates a Green Lind Relification Described in all Relificacy Described in an annual Adabasia and tempolary to indust the up-turbed of the Chamberl will have turbed or for the Chamberl will have turbed in Chamberl will have turbed or for the Chamberl will have turbed in the Chamberl wil or LSU, the Cardinal will have station of not having a by Number One not so orasy to town out to carm a trip to South to the south to carm a trip to South to the south carm a trip to South to carm a trip to South to the south carming a south to carm a trip to South to the south to carming the south to the south to contact south t

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Tuesday, September 18, 2012 ◆ 13

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The Stanford Baile

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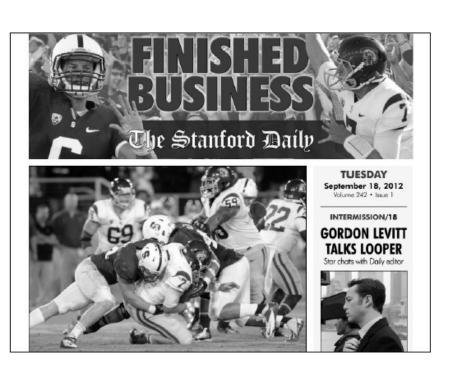
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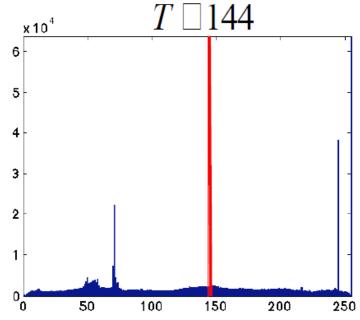
Handing out the USC game balls

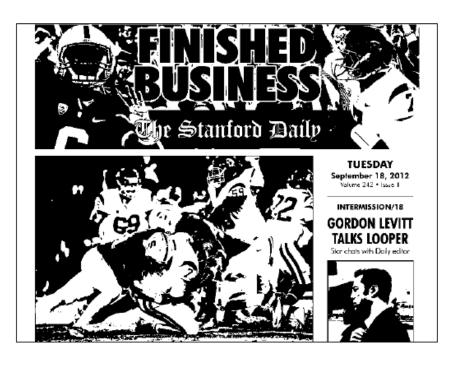
By SAM RISHER

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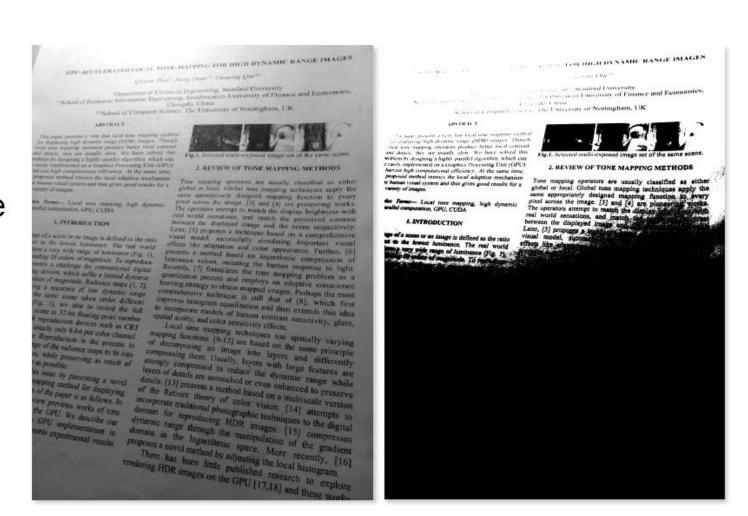






Sometimes, a global threshold does not work

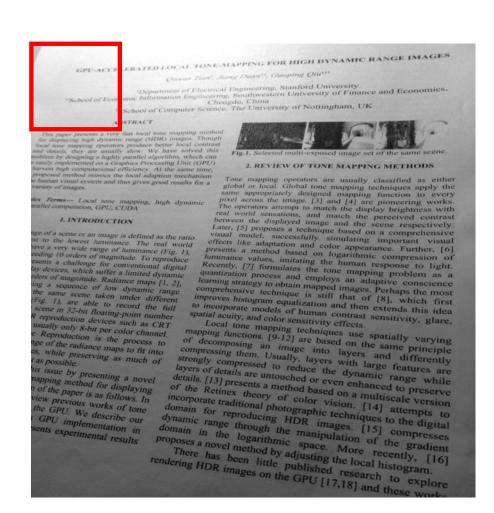
Original image



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



sents experimental results

readering HDR images on the GPU [17,18] and these worths. Locally thresholded result

proposes a novel method by adjusting the local histogram. There has been little published research to explore