



Global Thresholding

Let $f(x,y)$ be the source image
and $b(x,y)$ the binary image.

$$b(x, y) = \begin{cases} 1, & f(x, y) \leq T \\ 0, & \text{otherwise} \end{cases}$$

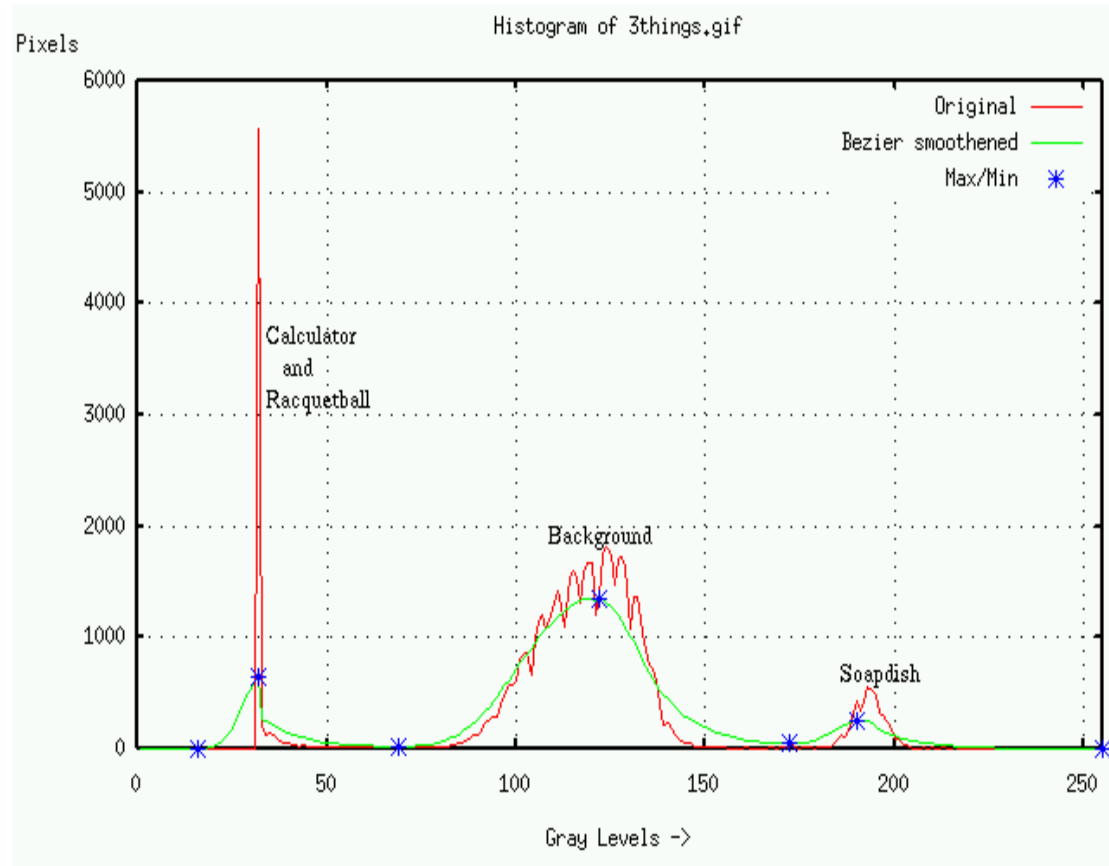
$\text{Object} \in [T_1, T_2]$

$$b(x, y) = \begin{cases} 1, & T_1 \leq f(x, y) \leq T_2 \\ 0, & \text{otherwise} \end{cases}$$

$\text{Object} \in Z$

where Z is a set of intensity values

$$b(x, y) = \begin{cases} 1, & f(x, y) \in Z \\ 0, & \text{otherwise} \end{cases}$$



Histogram analysis



Semithresholding

- Pixels whose values lie within a given threshold range retain their original values. Pixels with values lying outside of the threshold range are set to 0.

$$b(x, y) = \begin{cases} f(x, y) & \text{if } h \leq f(x, y) \leq k \\ 0 & \text{otherwise} \end{cases}$$

- Regions of high values can be isolated using:

$$b(x, y) = \begin{cases} f(x, y) & \text{if } f(x, y) \geq k \\ 0 & \text{otherwise} \end{cases}$$

- Regions of low values can be isolated using:

$$b(x, y) = \begin{cases} f(x, y) & \text{if } f(x, y) \leq k \\ 0 & \text{otherwise} \end{cases}$$



Example





Multilevel thresholding

- Multilevel thresholding allows for segmentation of pixels into multiple classes.
- For example, if the image histogram contains three peaks, then it is possible to segment the image using two thresholds. These thresholds divide the value set into three non-overlapping ranges, each of which can be associated with a unique value in the resulting image.
- Let $f(x,y)$ be the source image and let k_1, \dots, k_n be threshold values satisfying $k_1 > k_2 > \dots > k_n$. These values partition R into $n+1$ intervals which are associated with values v_1, \dots, v_{n+1} in the thresholded result image.
- The threshold image b is defined by:

$$b(x, y) = \begin{cases} v_1 & \text{if } f(x, y) < k_1 \\ v_i & \text{if } k_{i-1} \leq f(x, y) < k_i \\ v_{n+1} & \text{if } k_n \leq f(x, y) \end{cases}$$



Variable Thresholding

- Variable thresholding allows different threshold levels to be applied to different regions of an image.
- Let $f(x,y)$ be the source image and let $d(x,y)$ denote the local (region) threshold value associated with each point in the image, that is $d(x,y)$ is the threshold value associated with the region in which point (x,y) lies.
- The thresholded image $b(x,y)$ is defined by:

$$b(x, y) = \begin{cases} 1 & \text{if } f(x, y) \geq d(x, y) \\ 0 & \text{if } f(x, y) < d(x, y) \end{cases}$$



Threshold selection using mean and standard deviation

- Let $f(x,y)$, be an image
- The mean of $f(x,y)$ is: $\mu = \frac{1}{mn} \sum_{i=1}^m \sum_{j=1}^n f(i, j)$
- The standard deviation of the image a is:

$$\sigma = \sqrt{\frac{1}{mn} \sum_{i=1}^m \sum_{j=1}^n [f(i, j) - \mu]^2}$$

- The threshold level T is set at: **$T = k_1\mu + k_2\sigma$** where the constants k_1 and k_2 are image type dependent.
- For typical low-resolution IR images $k_1 = k_2 = 1$ seems to work fairly well.
- For higher resolutions $k_1 = 1$ or $k_1 = 1.5$ and $k_2 = 2$ may yield better results.