Image Processing and Computer Vision (IPCV)



Prof. J. Weickert Mathematical Image Analysis Group

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Example Solutions for Classroom Assignment 11 (C11)

Problem 1: (Otsu's Threshold Selection Method)

By using the identities $\mu_0(T) = \frac{\mu(T)}{\omega(T)}$ and $\mu_1(T) = \frac{\mu_{\text{tot}} - \mu(T)}{1 - \omega(T)}$, it suffices to take the definition of $\sigma_B^2(T)$ and to plug these values in.

$$\begin{split} \sigma_B^2(T) &= \omega(T)(\mu_0(T) - \mu_{\text{tot}})^2 + (1 - \omega(T))(\mu_1(T) - \mu_{\text{tot}})^2 \\ &= \omega(T) \left(\frac{\mu(T)}{\omega(T)} - \mu_{\text{tot}}\right)^2 + (1 - \omega(T)) \left(\frac{\mu_{\text{tot}} - \mu(T)}{1 - \omega(T)} - \mu_{\text{tot}}\right)^2 \\ &= \omega(T) \left(\frac{\mu(T) - \mu_{\text{tot}}\omega(T)}{\omega(T)}\right)^2 + (1 - \omega(T)) \left(\frac{\mu_{\text{tot}}\omega(T) - \mu(T)}{1 - \omega(T)}\right)^2 \\ &= \frac{(\mu(T) - \mu_{\text{tot}}\omega(T))^2}{\omega(T)} + \frac{(\mu_{\text{tot}}\omega(T) - \mu(T))^2}{1 - \omega(T)} \\ &= (\mu_{\text{tot}}\omega(T) - \mu(T))^2 \left(\frac{1}{\omega(T)} + \frac{1}{1 - \omega(T)}\right) \\ &= (\mu_{\text{tot}}\omega(T) - \mu(T))^2 \frac{1 - \omega(T) + \omega(T)}{\omega(T)(1 - \omega(T))} \\ &= \frac{(\mu_{\text{tot}}\omega(T) - \mu(T))^2}{\omega(T)(1 - \omega(T))} \end{split}$$

Problem 2: (Toboggan Watershed Algorithm)

First of all we compute the magnitude of the first derivative of the signal f by using backward differences:

$$\boldsymbol{f} = (6, 1, 5, 5, 3, 5, 6, 3, 4, 4, 2, 7, 7, 6, 3, 2)^{\top}$$
$$\boldsymbol{g} := |f_i - f_{i-1}|_{i=1\dots 16} = (0, 5, 4, 0, 2, 2, 1, 3, 1, 0, 2, 5, 0, 1, 3, 1)^{\top}.$$

In each pixel of g we follow the direction of the steepest descend until a local minimum (in bold font) is reached and keep track of all pixels on the way to this minimum. The tracked pixels are set to the value f_i of the original signal that corresponds the local minimum:

$$\begin{array}{rcl} \boldsymbol{f} & = & (6,1,5,5,3,5,6,3,4,4,2,7,7,6,3,2)^{\top} \\ \boldsymbol{g} & = & (\mathbf{0},5,\underline{4},\mathbf{0},2,\underline{2},\mathbf{1},3,\underline{1},\mathbf{0},2,\underline{5},\mathbf{0},1,3,\underline{1})^{\top} \\ \boldsymbol{f}_{\text{filtered}} & = & (6,6,5,5,5,6,6,6,4,4,4,7,7,7,7,2)^{\top} \end{array}.$$