



Let us assume the prior probabilities for background and foreground be $P(Bg)$, $P(Fg)$ respectively.

Let $I_{x,y}$ be the intensity value of cell (x, y) .

If $I_{x,y} > 0$, $I_{x,y}$ is a background pixel.

Else

$I_{x,y}$ is a foreground pixel.

$$P(I_{x,y} | \text{background}) =$$

$$P(\text{background} | I_{x,y}) = P(I_{x,y} | \text{background}) P(\text{background})$$

$$= P(Bg) \times \frac{1}{\sqrt{2\pi} \sigma_1} \exp \left(- \frac{(I_{x,y} - \mu_1)^2}{2\sigma_1^2} \right)$$

Similarly,

$$P(\text{foreground} | I_{x,y}) = P(I_{x,y} | \text{foreground}) P(\text{foreground})$$

$$= P(Fg) \times \frac{1}{\sqrt{2\pi} \sigma_2} \exp \left(- \frac{(I_{x,y} - \mu_2)^2}{2\sigma_2^2} \right)$$

Let θ = value of intensity for which
 probability of density of background
 = probability of density of foreground.

$$\Rightarrow \frac{P(B_g)}{\sqrt{2\pi} \sigma_1} \exp\left(-\frac{(\theta - \mu_1)^2}{2\sigma_1^2}\right) = \frac{P(F_g)}{\sqrt{2\pi} \sigma_2} \exp\left(-\frac{(\theta - \mu_2)^2}{2\sigma_2^2}\right)$$

$$\Rightarrow \log\left(\frac{P(B_g)}{\sigma_1}\right) + \log\left(\exp\left(-\frac{(\theta - \mu_1)^2}{2\sigma_1^2}\right)\right) = \log\left(\frac{P(F_g)}{\sigma_2}\right) + \log\left(\exp\left(-\frac{(\theta - \mu_2)^2}{2\sigma_2^2}\right)\right)$$

$$\Rightarrow \log(P(B_g)) - \log(\sigma_1) - \frac{(\theta - \mu_1)^2}{2\sigma_1^2} = \log(P(F_g)) - \log(\sigma_2) - \frac{(\theta - \mu_2)^2}{2\sigma_2^2}$$

$$\Rightarrow \log(P(B_g)) - \log(\sigma_1) - \frac{(\theta - \mu_1)^2}{2\sigma_1^2}$$

$$= \log(P(F_g)) - \log(\sigma_2) - \frac{(\theta - \mu_2)^2}{2\sigma_2^2}$$

[\because Given,
 $P(B_g) = P(F_g)$
 $\& \sigma_1 = \sigma_2$]

$$\Rightarrow \frac{(\theta - \mu_1)^2}{2\sigma_1^2} = \frac{(\theta - \mu_2)^2}{2\sigma_2^2}$$

$$\Rightarrow \theta - \mu_1 = \pm(\theta - \mu_2) \quad \left[\begin{array}{l} \theta - \mu_1 = \theta - \mu_2 \text{ is} \\ \text{invalid because} \\ \Rightarrow \mu_1 = \mu_2 \end{array} \right]$$

$$\Rightarrow \theta - \mu_1 = -\theta + \mu_2$$

$$\Rightarrow 2\theta = \mu_1 + \mu_2$$

$$\boxed{\theta = \frac{\mu_1 + \mu_2}{2}}$$

$$\Rightarrow \text{Putting } \theta = \frac{\mu_1 + \mu_2}{2}$$

$$\log(P(B_y)) - \log \sigma_1 - \frac{\left(\frac{(\mu_1 + \mu_2)}{2} - \mu_1\right)^2}{2\sigma_1^2}$$

$$= \log(P(F_y)) - \log \sigma_2 - \frac{\left(\frac{(\mu_1 + \mu_2)}{2} - \mu_2\right)^2}{2\sigma_2^2}$$

$$\Rightarrow \log\left(\frac{P(B_y)}{P(F_y)}\right) - \log\left(\frac{\sigma_1}{\sigma_2}\right) =$$

$$= \frac{\left(\frac{\mu_2 - \mu_1}{2}\right)^2}{2\sigma_1^2} - \frac{\left(\frac{\mu_1 - \mu_2}{2}\right)^2}{2\sigma_2^2}$$

$$\Rightarrow \log \left(\frac{P(B_1)}{P(F_1)} \right) - \log \left(\frac{\sigma_1}{\sigma_2} \right)$$

$$= \frac{(u_2 - u_1)^2}{8\sigma_1^2} - \frac{(u_2 - u_1)^2}{8\sigma_2^2}$$

$$= \frac{(u_2 - u_1)^2}{8} \left(\frac{1}{\sigma_1^2} - \frac{1}{\sigma_2^2} \right)$$

$$\Rightarrow \log \left(\frac{P(B_1)}{P(F_1)} \right) - \log \left(\frac{\sigma_1}{\sigma_2} \right) = \frac{(u_2 - u_1)^2}{8} \left(\frac{1}{\sigma_1^2} - \frac{1}{\sigma_2^2} \right)$$

$$c) \log(4 \times P(F_1)) - \log(\cancel{\sigma_1}) - \frac{(0 - 100)^2}{2\sigma_1^2}$$

$$= \log(P(F_1)) - \log(\cancel{\sigma_1}) - \frac{(0 - 200)^2}{2\sigma_2^2}$$

$$\Rightarrow \log(4) + \log(\cancel{P(F_1)}) - \frac{(0 - 100)^2}{2\sigma_1^2}$$

$$= \log(\cancel{P(F_1)}) - \frac{(0 - 200)^2}{2\sigma_2^2}$$

$$\Rightarrow \log(4) = \frac{(0^2 - 2000 + 10000)}{2\sigma_1^2} - \frac{(0^2 - 4000 + 40000)}{2\sigma_2^2}$$

$$\Rightarrow 2\sigma_1^2 \log(4) = 2000 - 30000$$

$$\Rightarrow 0 = \frac{2 \log(4) \sigma_1^2 + 30000}{200} = \frac{2.773 \sigma_1^2 + 30000}{200}$$