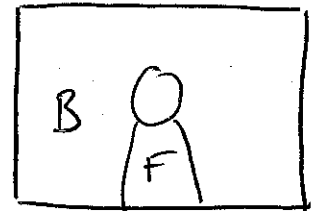


Matting

9/11/07 ①

$$I = \alpha F + (1 - \alpha) B$$

↑ image you capture ↑ foreground "object" ↑ alpha matte ↑ background image



Once you have α , F , B you can generate new image I_{new} with new background B_{new} :

$$\alpha F + (1 - \alpha) B_{new} = I_{new}$$

So how do you solve for α , F , B for a single I ?

Solution 1:

Use a constant-color background whose color is not found in F .

~~$F_i = [R_i \ G_i \ 0]$~~
 $F_i = [R_i \ G_i \ 0]$
 $B_i = [0 \ 0 \ B_B]$

↑ use this

↪ blue screen
green screen

$$I_B = \frac{\alpha 0}{0} + (1 - \alpha) B_B$$

$$\alpha = 1 - \frac{I_B}{B_B} \leftarrow \text{compute } \alpha$$

$$\text{Then } F = \frac{I - (1 - \alpha) B}{\alpha}$$

This is a huge assumption, restricting foreground colors to the RG plane!

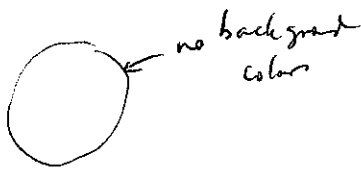
Solution 2:

Assume that frequent color is gray, constant-color background

$$F = \{k, k, k\} \quad \leftarrow \text{unknown}$$

$$I_r = k\alpha$$

$$I_g = k\alpha$$



$$I_b = \underbrace{k\alpha}_{\text{known}} + (1-\alpha)B_b \quad \leftarrow \text{known}$$

$$I_r = I_g$$

$$\text{so } \frac{(I_b - I_r)}{B_b} = 1 - \alpha$$

$$\alpha = 1 - \frac{I_b - I_r}{B_b}$$

$$k = \frac{I_r}{\alpha}$$

$$\therefore \alpha, F, B \ll$$

Solution 3: Triangulation using 2 different background colors, foreground is unknown (3)

$$F = \{F_r, F_g, F_b\}$$

$$B_1 = \{0, 0, B_{b1}\}$$

$$B_2 = \{0, 0, B_{b2}\}$$

$$I_{b1} = \alpha F_b + (1-\alpha) B_{b1}$$

$$I_{b2} = \alpha F_b + (1-\alpha) B_{b2}$$

$$I_{b1} - I_{b2} = (1-\alpha) B_{b1} - (1-\alpha) B_{b2}$$

$$I_{b1} - I_{b2} = (1-\alpha)(B_{b1} - B_{b2})$$

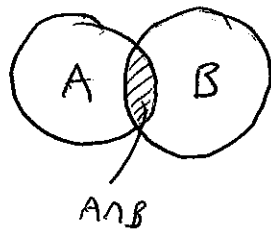
$$\alpha = 1 - \frac{I_{b1} - I_{b2}}{B_{b1} - B_{b2}}$$

Bayesian Matting

Bayes Theorem:
$$P(A|B) = \frac{P(B|A) P(A)}{P(B)}$$

Labels:
 - $P(A|B)$: posterior
 - $P(B|A)$: likelihood
 - $P(A)$: prior
 - $P(B)$: normalizing constant

$P(A|B)$



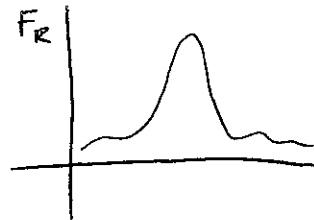
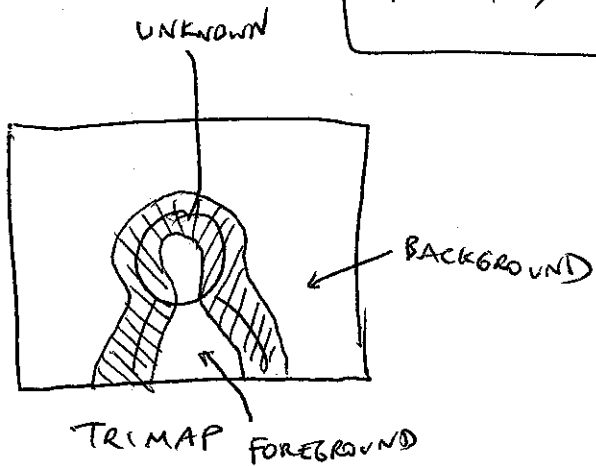
$$P(A|B) = \frac{P(A \cap B)}{P(B)}$$

$$P(B|A) = \frac{P(A \cap B)}{P(A)}$$

equating the two equations:

$$P(B) P(A|B) = P(B|A) P(A)$$

$$P(A|B) = \frac{P(B|A) P(A)}{P(B)}$$



Maximum a posteriori (MAP): $\arg \max_{F, B, \alpha} P(F, B, \alpha | I)$

Apply Bayes rule

$$= \arg \max_{F, B, \alpha} P(I | F, B, \alpha) P(F) P(B) P(\alpha) / P(I)$$

← constant with respect to optimization

Take log $P() = L()$

$$= \arg \max_{F, B, \alpha} \underbrace{L(I|F, B, \alpha)}_1 + \underbrace{L(F)}_2 + \underbrace{L(B)}_3 + \underbrace{L(\alpha)}_4$$

Think of this like this:
 Find F, B, α that result in the image as close as possible to I while observing the probability distributions of $F, B,$ and α !

Now model each term.

1. $L(I|F, B, \alpha)$

eg. Probability distribution of I given F, B, α

well $I = \alpha F + (1-\alpha) B$
 $I_{THEORY} \leftarrow$ center of the distribution

$$L(I|F, B, \alpha) = - \frac{((I_{meas} - I_{THEORY})^2)}{\sigma_I^2}$$

where σ_I standard deviation
 image you are computing
 You want $I_{THEORY} \rightarrow I_{meas}$

2, 3. $L(F)$ measured in a window around the pixel that is in the known foreground
 Same for background

4. $L(\alpha)$ assumed constant.