

Project 3 – Rotobrush

Segmenting Deformable Objects in a Video

Segmentation

- Given a point/pixel $x_{i,j}$ in an image, where does it belong?
- Why do we need segmentation?
 - Localization
 - Object Detection
 - Tracking

Deformable Objects

- Objects that change shape
- Changing shape → Can't use static Mask

SnapCut

1. Segmenting object of interest in first frame
 - Call it foreground, F
 - Everything else is the background B
2. Create Local Classifiers
 - Local windows along edge of initial mask W_k^t
 - Must have *some* overlap between windows (20-30%)
3. For each local window
 1. Initialize Color Model for GMM
 2. Initialize Shape Model
 3. Combine Shape + Color
 4. Update Color + Shape Model

Digging Deeper

- Use `roipoly()` to select mask
- Select k points around object
 - $k \sim 60 \times 60$ windows \rightarrow Must have overlap
- **Color Model**
 - Per window

$$p_c(x) = \frac{p_c(x|F)}{(p_c(x|F) + p_c(x|B))}$$

- Fit GMM model \rightarrow can use MATLAB functions
 - Find probabilities of foreground
- **Color Confidence**
 - Are foreground and background separable?

$$f_c = 1 - \frac{\int_{W_c} |L^t(x) - p_c(x)| \cdot \omega_c(x) dx}{\int_{W_k} \omega_c(x) dx}$$

d = euclidean distance between x and foreground boundary

$$\omega_c = e^{\frac{-d^2(x)}{\sigma_c^2}}$$

- f_c is a single value
 - $\omega_c \rightarrow$ Weight function
 - $|L^t(x) - p_c(x)|$ = Subtract probabilities from mask
- After doing this for every window \rightarrow Will have border around object
- **Shape Model**

$$f_s(x) = 1 - e^{\frac{-d^2(x)}{\sigma_s^2}}$$

σ_s is a parameter

- Shape confidence decreases as color confidence increases
- **Local Window Propagation**
 - Use SIFT features \rightarrow track k points across frames
 - Update window based on matched point in next frame
 - Use optical flow (MATLAB functions) to update windows
 - Update Color and Shape Models with next frame data

$$p_F^k(x) = f_s(x)L^{t+1}(x) + (1 - f_s(x))p_c(x)$$

$$p_F(x) = \frac{\sum_k p_F^k(x)(|x - c_k| + \epsilon)^{-1}}{\sum_k (|x - c_k| + \epsilon)^{-1}}$$

c_k is the center of the window \implies closer points to center weighted more heavily

ϵ is a small value to regulate points next to or on center

- Visualization functions given