

CMSC426: Project 3

Rotobrush: Hands-On

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Introducing **SnapCut** (Rotobrush in Adobe After Effects)

Rotobrush Output



Input Video
(courtesy of Artbeats)



Cutout Result

Segmenting the object of interest in the first frame

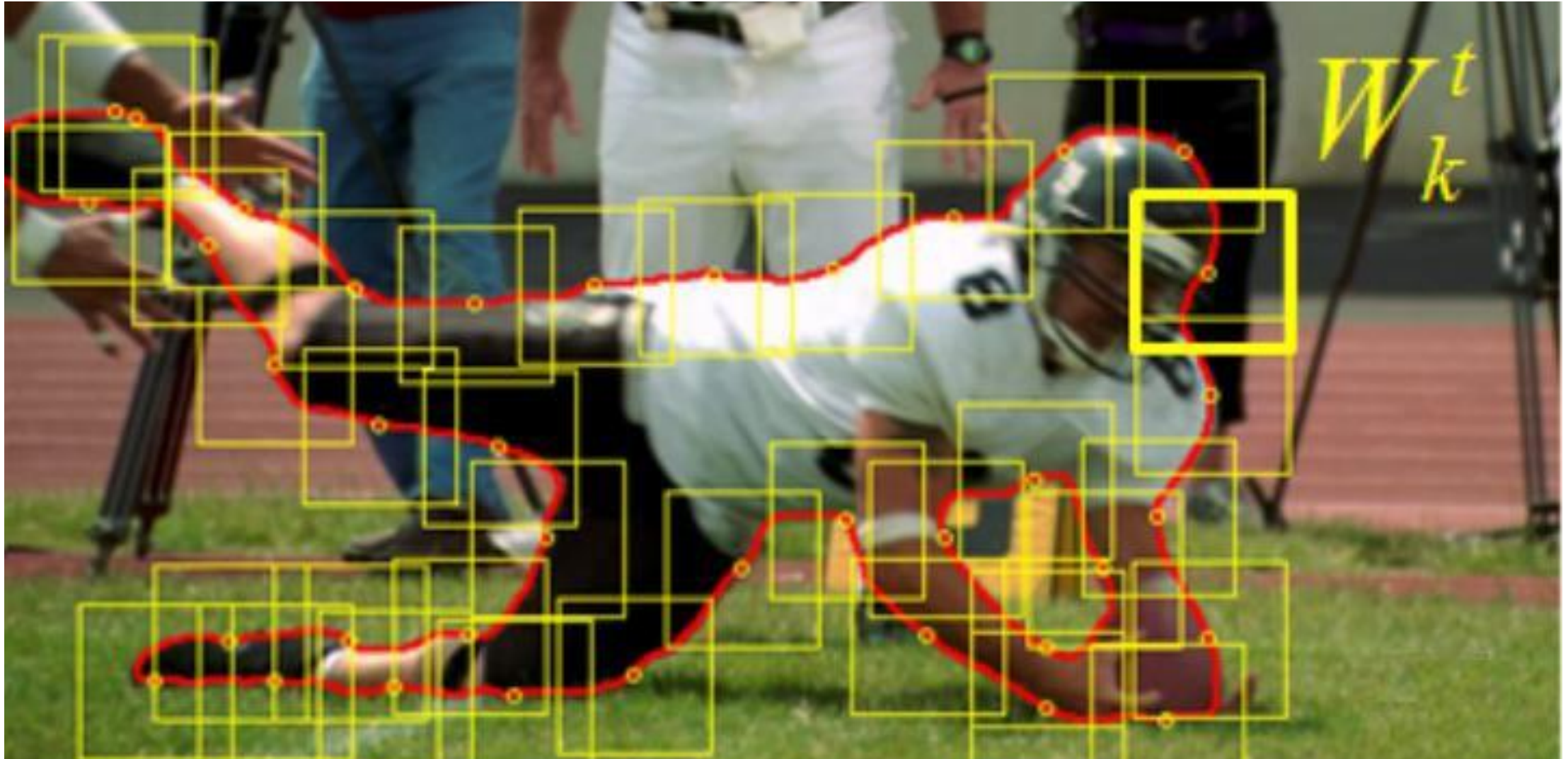
- Let's call it foreground or \mathcal{F} for ease.
- Rest everything is background or \mathcal{B}
- Use **roipoly** in MATLAB.



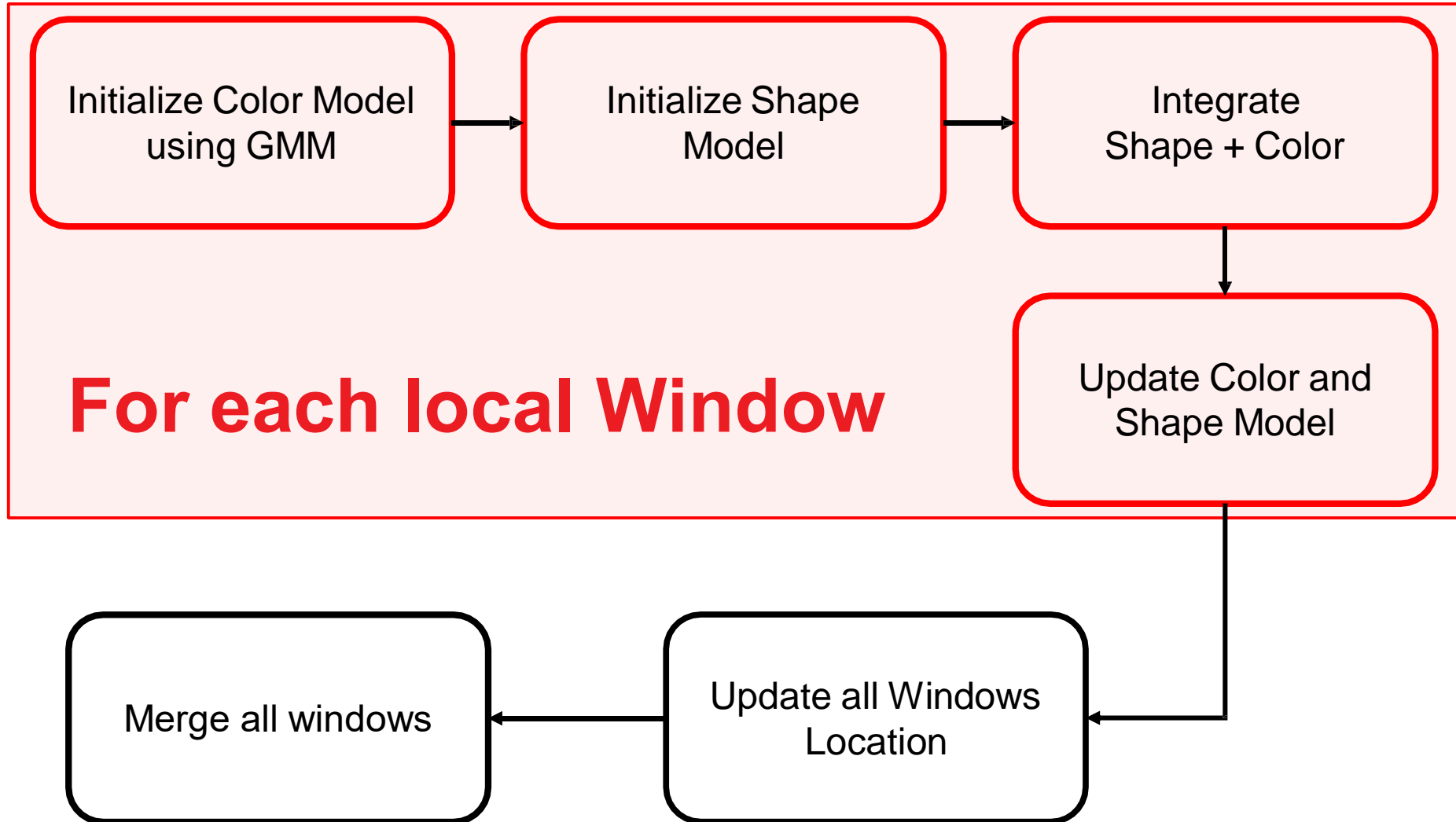


Create Local Classifiers

Create Local Windows



An Overview



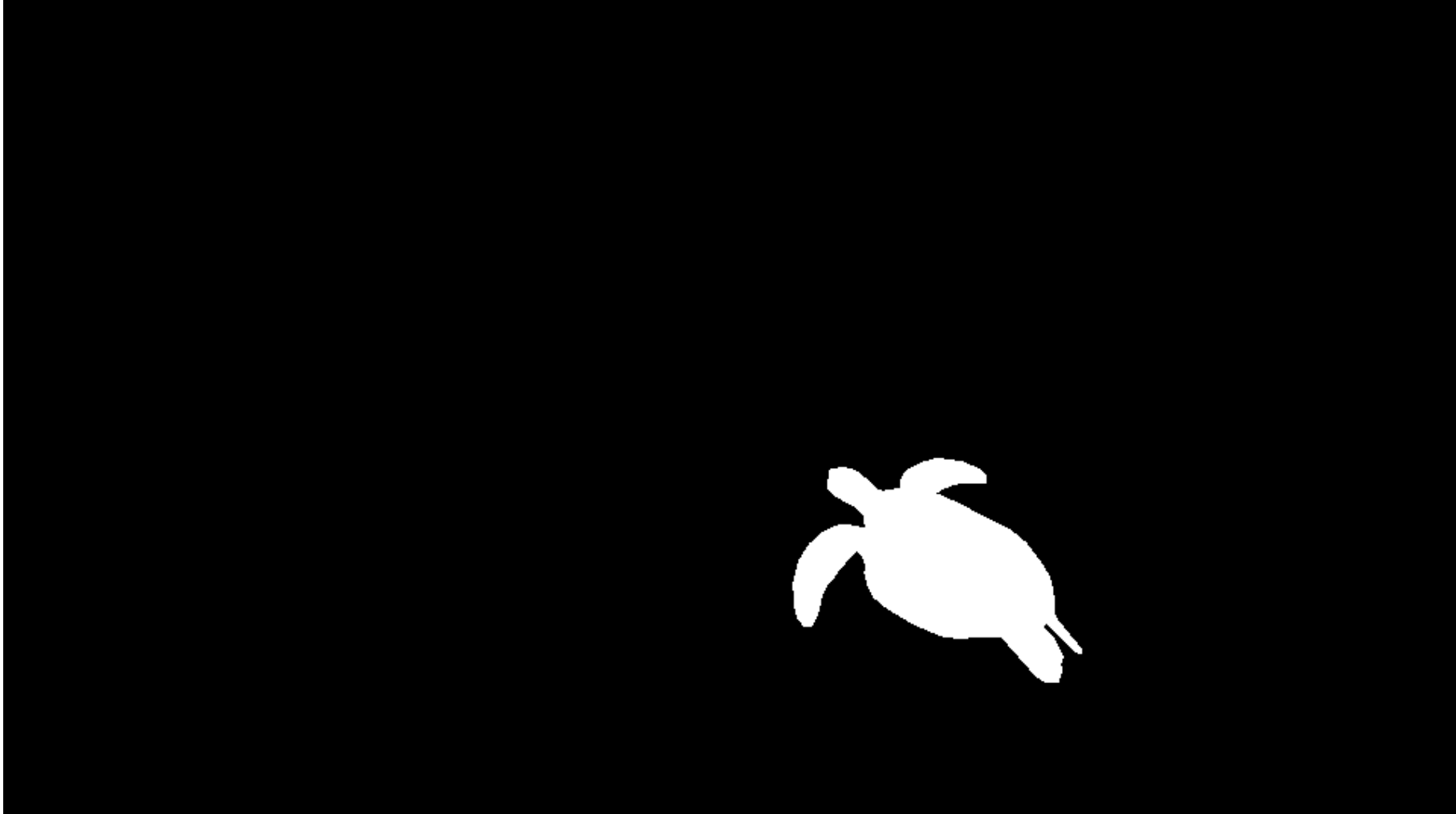


Let's dig deep

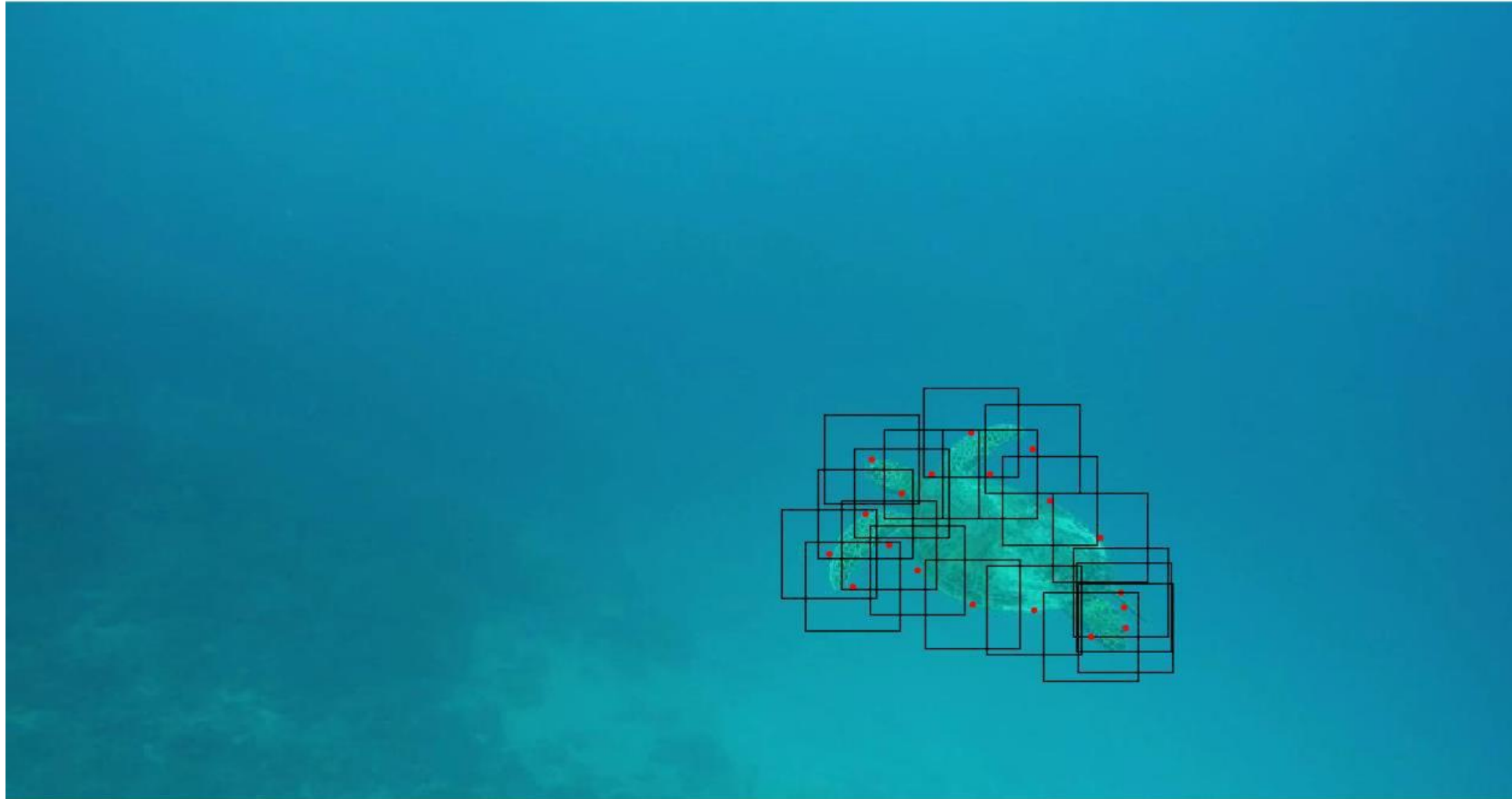
Input



Create Mask



Local Window (`initLocalWindows.m`)



Local Window (`initLocalWindows.m`)

```
[mask_outline, LocalWindows] = initLocalWindows(images{i}, mask,...  
                                                NumWindows, WindowWidth, flag)
```

For each boundary point:

```
[WindowCentersX, WindowCentersY] = ...  
    equidistantPointsOnPerimeter(Boundary(:,2), Boundary(:,1),  
    PointsForBoundary(i));
```

```
AllWindowCentersX =[AllWindowCentersX; WindowCentersX];
```

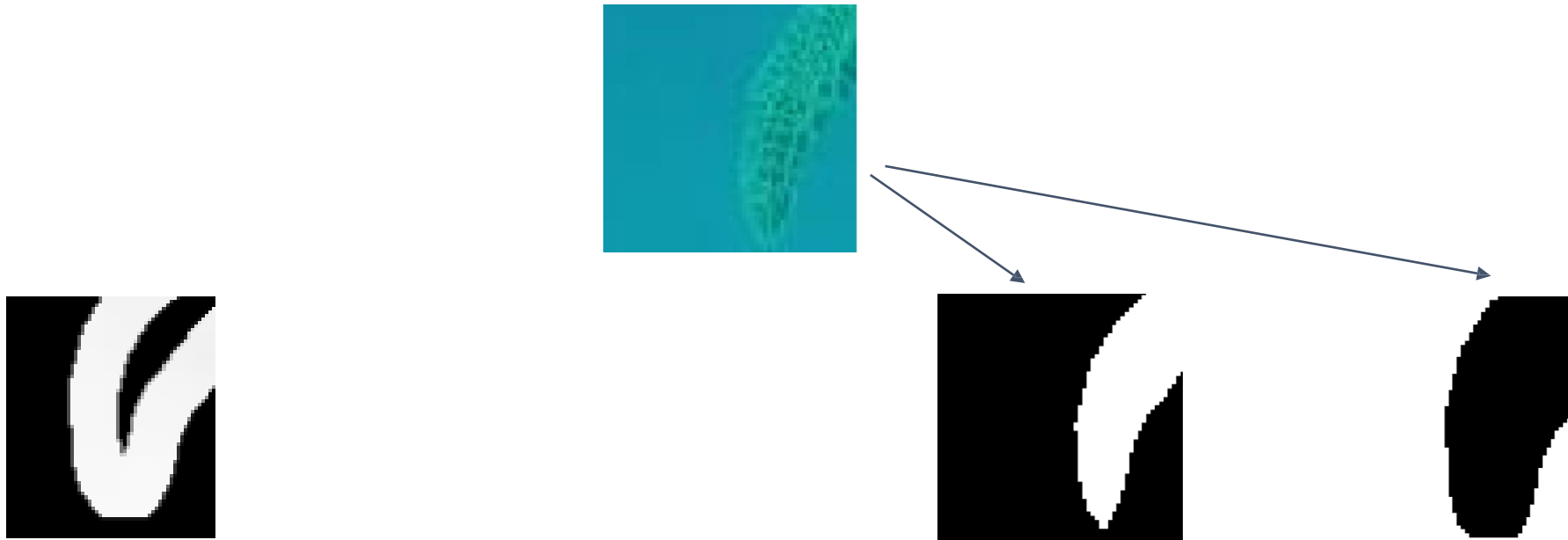
```
AllWindowCentersY =[AllWindowCentersY; WindowCentersY];
```

Where

`EquidistantPointsOnPerimeter` finds the equally spaced points along the perimeter of a polygon. Check implementation from MATLAB blog.

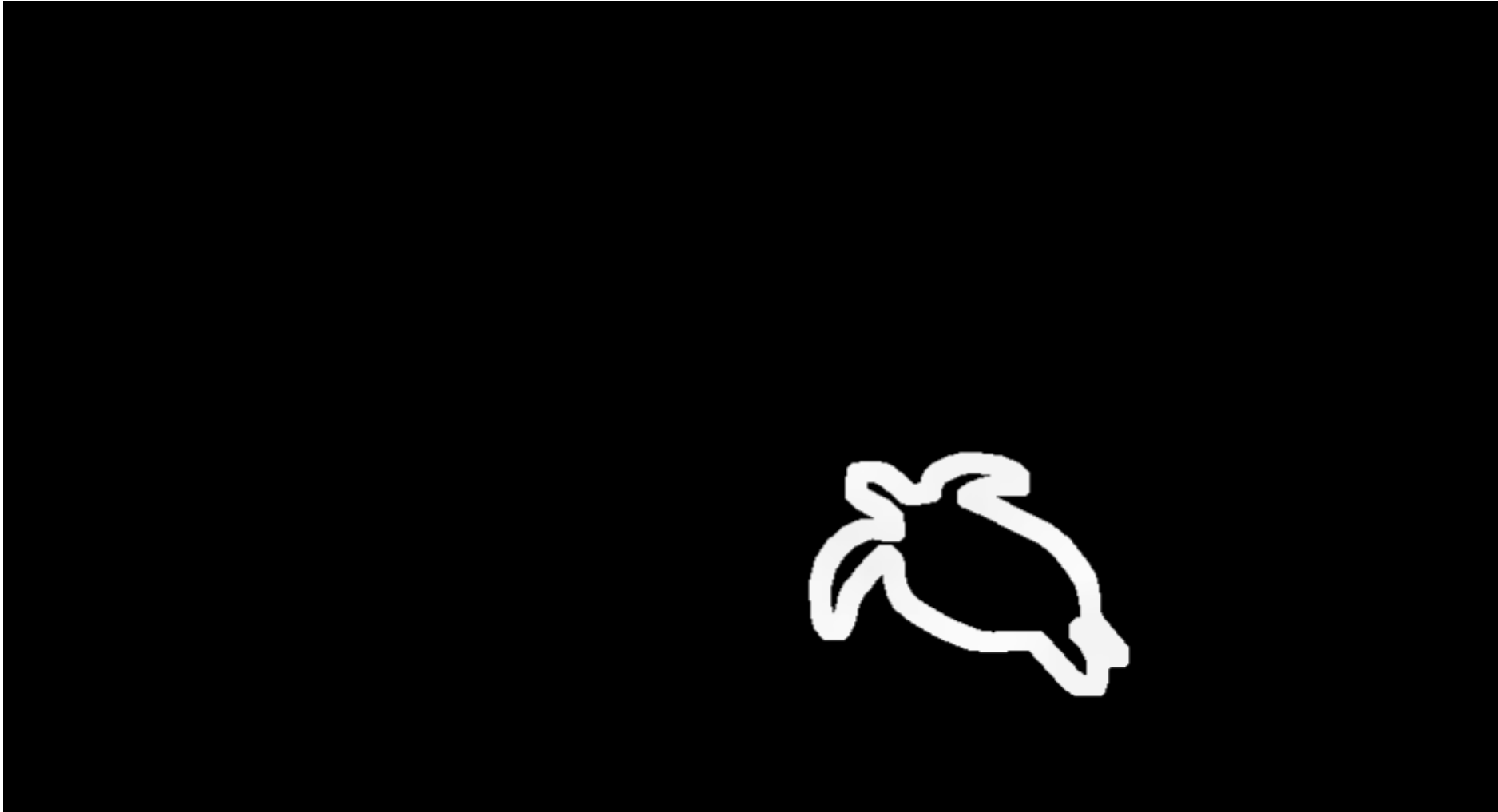
Source: <https://blogs.mathworks.com/steve/2012/07/06/walking-along-a-path/#5d684ab6-0edb-4c52-8df2-4bce4a79ee80>

Color model (`initColorModels.m`)



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

Color model (`initColorModels.m`)



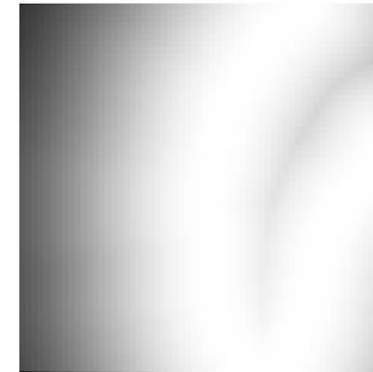
Color model (`initColorModels.m`)

L_t is the known segmentation label

p_c is foreground probability

$d(x)$ is the spatial distance between x and the foreground boundary

σ_c is fixed as half of the window size



$$\omega_c(x) = \exp(-d^2(x) / \sigma_c^2)$$

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

Color model (`initColorModels.m`)

```
ColorModels = initColorModels(images{1}, mask, mask_outline,...,  
    LocalWindows, BoundaryWidth, WindowWidth, DistTransMaskOutline);
```

where `DistTransMaskOutline = bwdist(mask_outline); %Find global distance transform` $\longrightarrow d$

```
F_GMM_Model = fitgmdist(ForegroundPixelsRGB, k, 'RegularizationValue',  
0.001);
```

```
B_GMM_Model = fitgmdist(BackgroundPixelsRGB, k, 'RegularizationValue',  
0.001);
```

For all x and y :

- Compute p_{df} of both the models above: P_f and P_b
- Compute f_c

Shape model (`initShapeConfidences.m`)

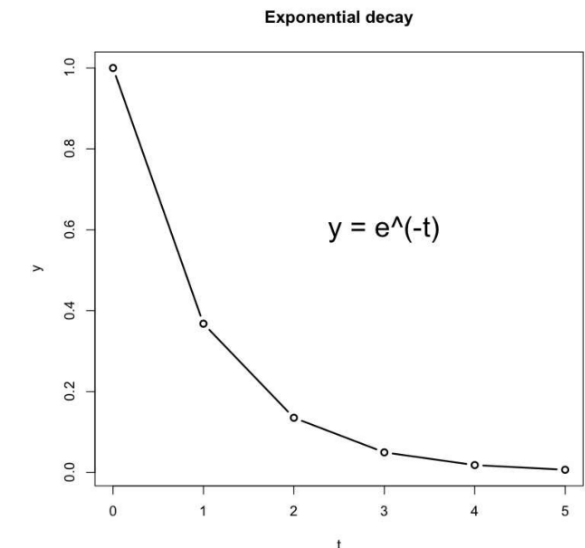


$$f_s(x) = 1 - \exp(-d^2(x) / \sigma_s^2)$$

f_s is shape confidence mask

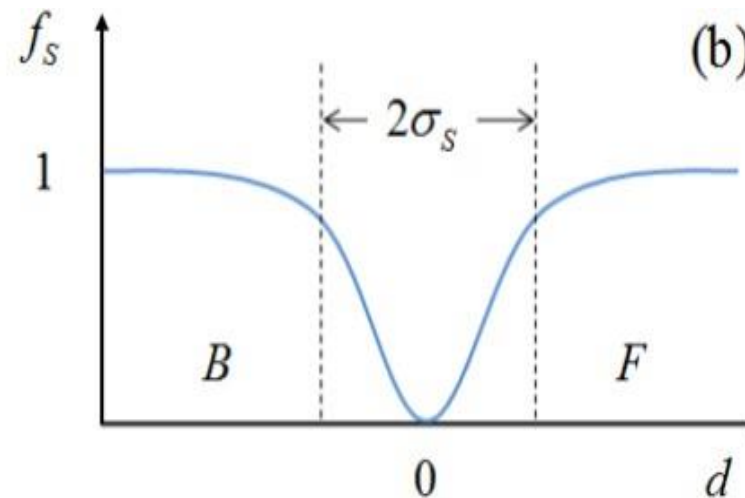
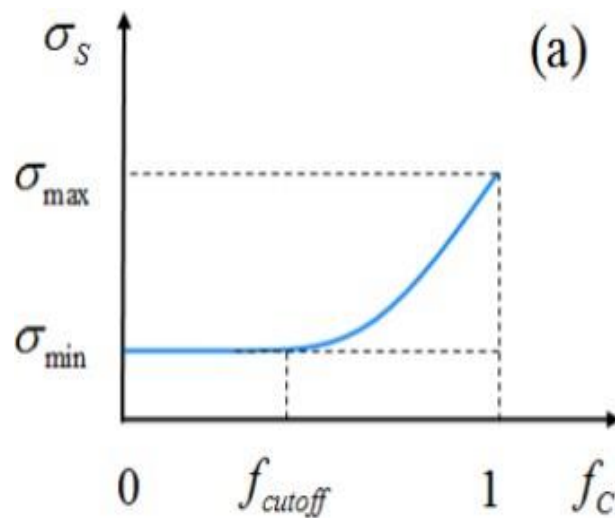
$d(x)$ is the spatial distance between x and the foreground boundary

σ_s is a parameter which depends on f_c

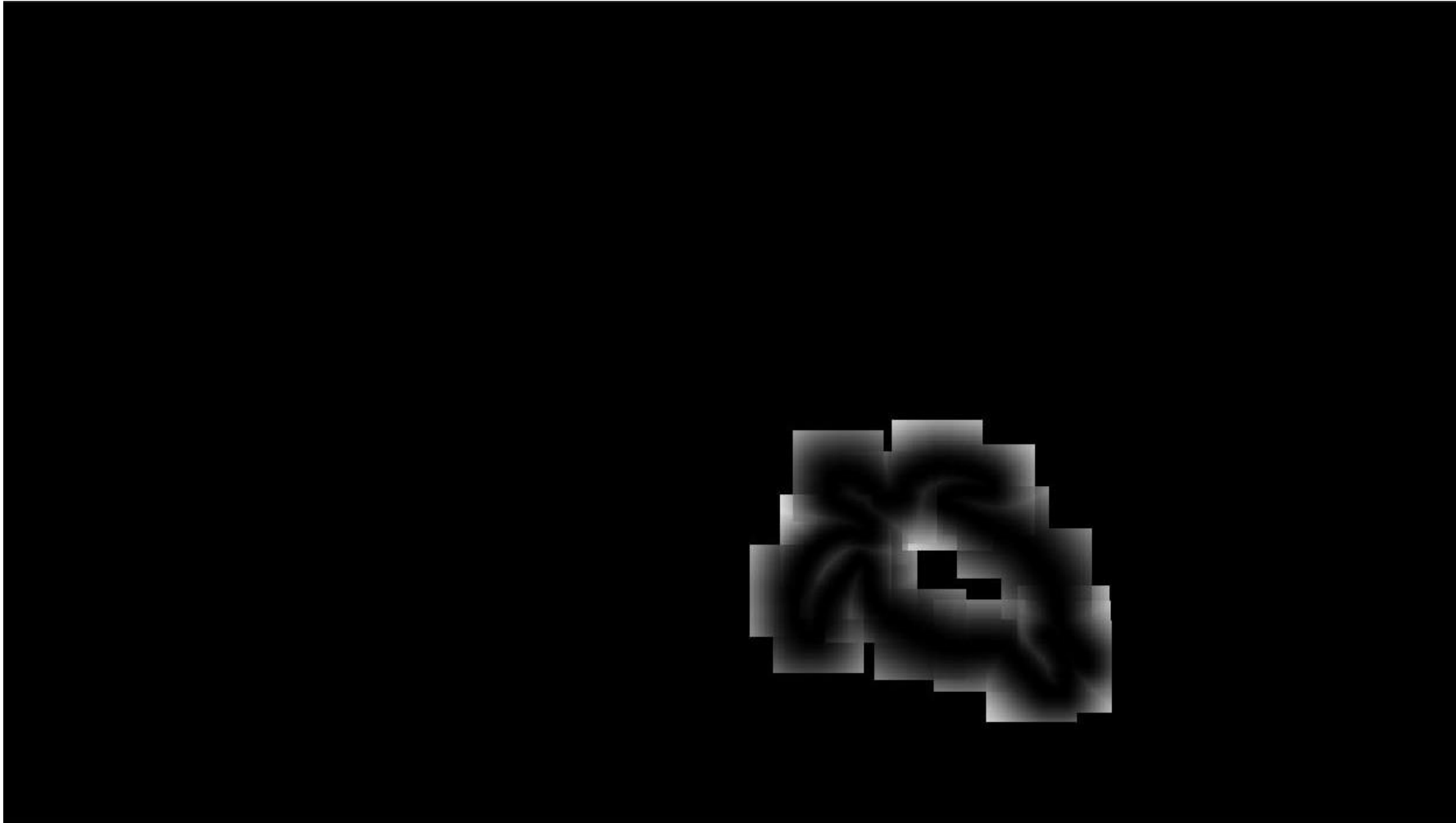


Shape model (`initShapeConfidences.m`)

$$\sigma_s = \begin{cases} \sigma_{min} + a(f_c - f_{cutoff})^r & f_{cutoff} < f_c \leq 1, \\ \sigma_{min} & 0 \leq f_c \leq f_{cutoff}, \end{cases}$$



Shape model (`initShapeConfidences.m`)



Shape model (`initShapeConfidences.m`)

```
ShapeModels = initShapeConfidences(LocalWindows, ColorModels,...  
                                   WindowWidth, SigmaMin, a, fcutoff, R,DistTransMaskOutline);
```

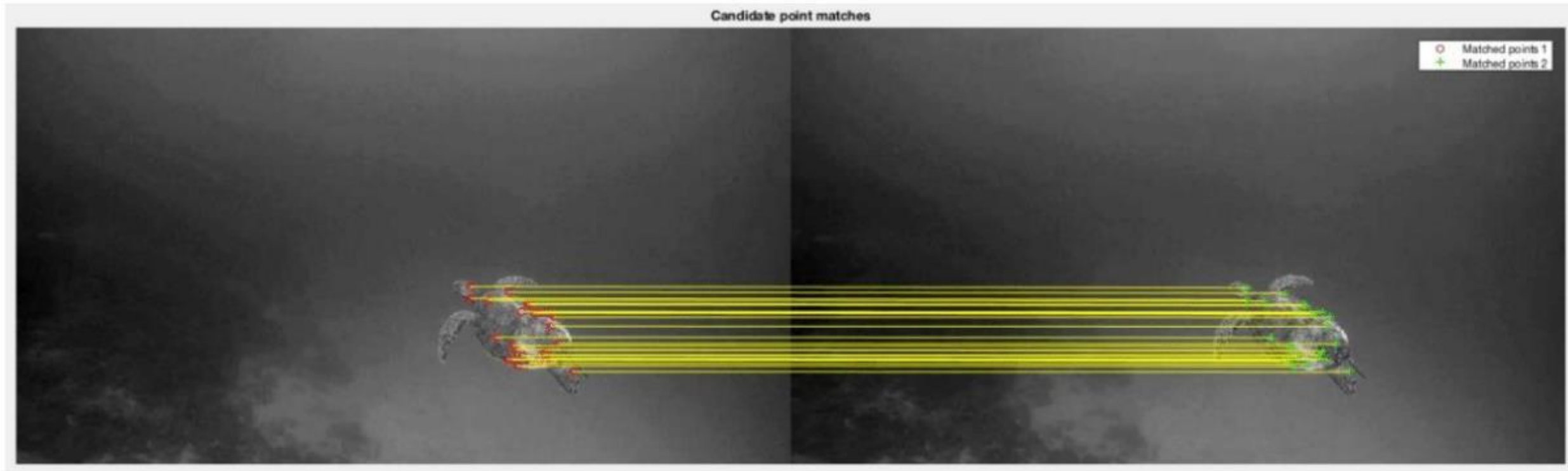
```
a = (sigmaMax - sigmaMin) / ((1-fcutoff)^R);
```

Define `sigmaS` for two different `fc` cases

For all `x` and `y`:

```
f_s = 1 - exp(-(d^2)/(sigmaS^2));
```

Local Window Propagation (`calculateGlobalAffine.m`)



Local Window Propagation (`calculateGlobalAffine.m`)

```
[warpedFrame, warpedMask, warpedMaskOutline, warpedLocalWindows] =  
    calculateGlobalAffine(images{prev}, images{curr}, ...,  
                          mask, LocalWindows);
```

- Detect and match features between the two frames
- Estimate geometric transformation from matching point pairs. Use `estimateGeometricTransform` and get the `tform`
- Warp the image/Mask using `imwarp`
- Get WarpedMaskOutline using `bwperim`
- Now find the WarpedLocalWindows using the `invert` of the `tform`

Optical Flow Wrapping (`localFlowWrap.m`)



Optical Flow Wrapping (`localFlowWrap.m`)

```
[NewLocalWindows] = localFlowWarp(WarpedPrevFrame, CurrentFrame, ... ,  
                                  LocalWindows, Mask, WindowWidth)
```

- Compute Optical Flow: use `opticalFlowFarneback` and `estimateFlow`
- For all `localWindows`
 - Find the new Local Window location by something like:
`old_window_location + average_flow` for that windows

Updating the Shape and Color Models (`updateModels.m`)

$$p_{\mathcal{F}}^k(x) = f_s(x)L^{t+1}(x) + (1 - f_s(x)) p_c(x)$$

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

L^{t+1} is the warped
segmentation label from
previous frame
 f_s is the shape confidence mask

p_c is the foreground probability
 c_k is the center of window
 ϵ is a small constant

Updating the Shape and Color Models (`updateModels.m`)

```
ShapeModels = updateShapeConfidences(LocalWindows, ColorModels,...,  
                                     WindowWidth, SigmaMin, a, fcutoff, R, DistTransMaskOutline)
```

where `DistTransMaskOutline = Bwdist(warpedMaskOutline)`

```
a = (sigmaMax - sigmaMin) / ((1-fcutoff)^R);
```

- Define `sigmaS` for two different `fc` cases

- For all `x` and `y`:

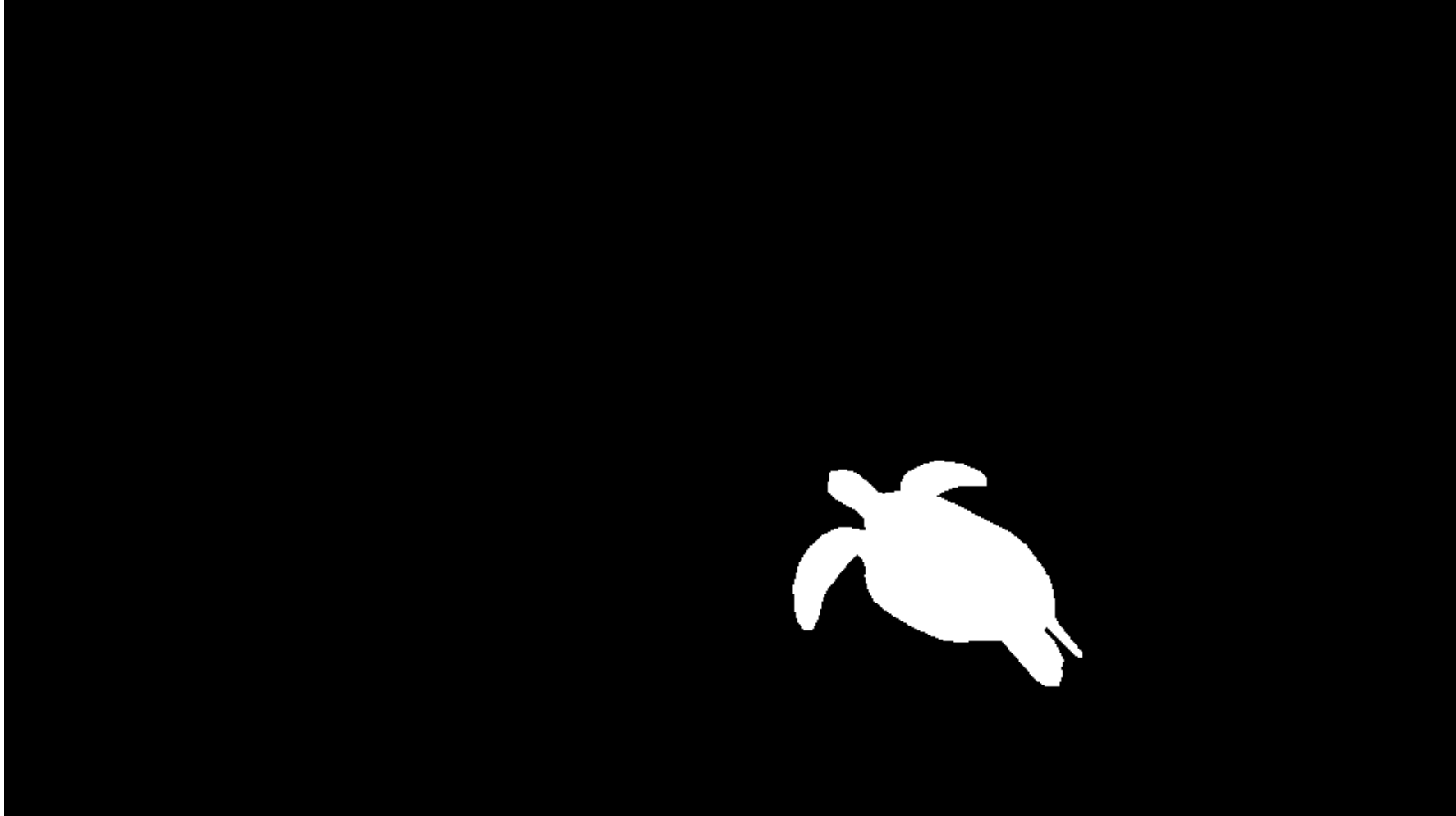
```
f_s = 1 - exp(-(d^2) / (sigmaS^2));
```

Then Combine Shape and Color models

- For each window:

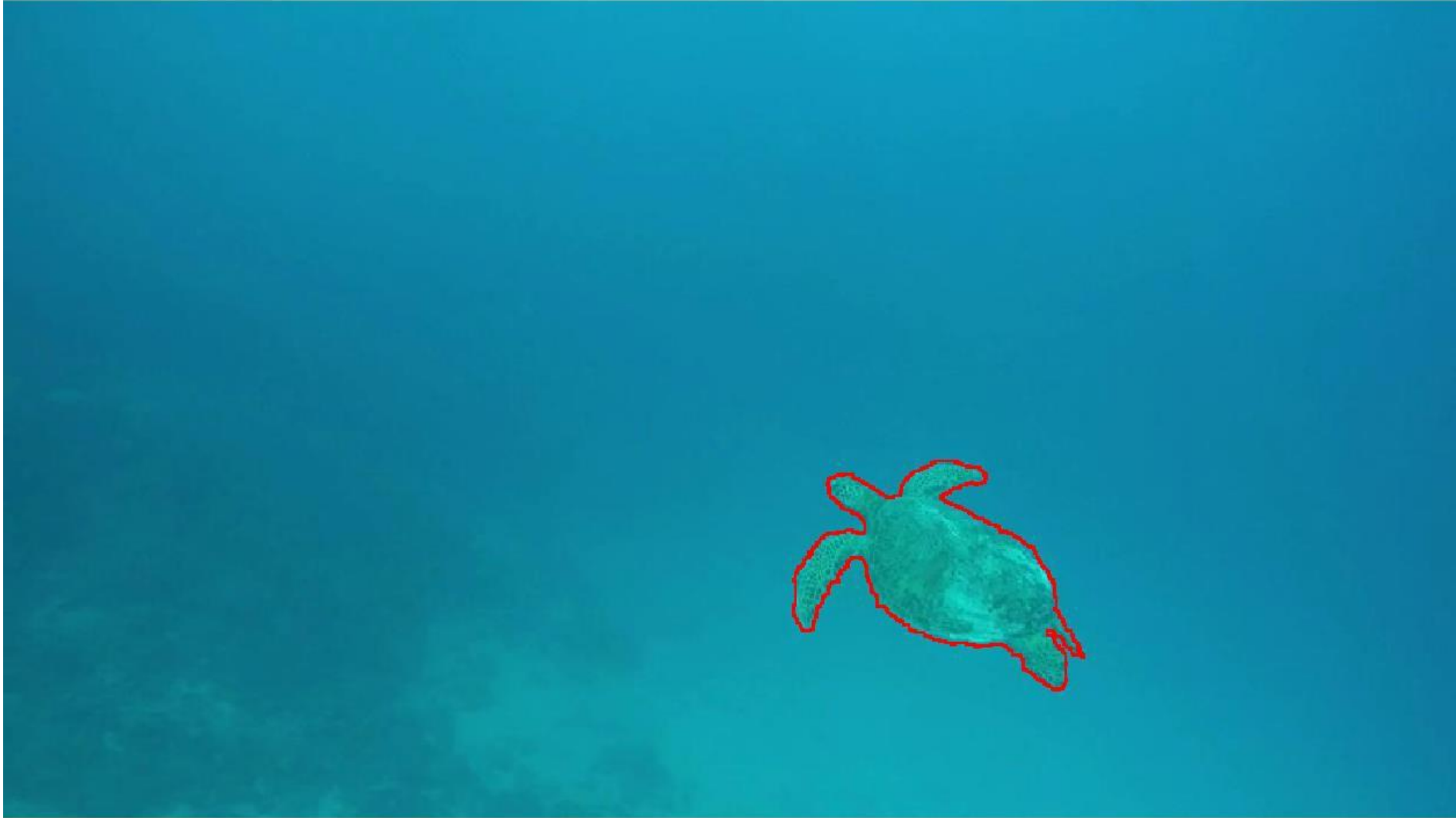
$$p_{\mathcal{F}}^k(x) = f_s(x)L^{t+1}(x) + (1 - f_s(x))p_c(x)$$

Updating the Shape and Color Models (`updateModels.m`)



Updating the Shape and Color Models (`updateModels.m`)

Hint: Update the color model based on the area of the new foreground probability mask



Pseudo-code (myRotobrush.m)

Algorithm 1 Rotobrush

```
1: procedure MYROTOBRUSH
2:   set parameters
3:   load images
4:   create mask
5:   initLocalWindows()           ▷ initialize local window
6:   initColorModels()           ▷ initialize Color model
7:   initShapeConfidences()       ▷ initialize Shape model
8:   for every image do
9:     calculateGlobalAffine()    ▷ transform between previous and current frames
10:    localFlowWarp()            ▷ local warping based on optical flow
11:    updateModels()             ▷ update color and shape model
12:  end for
13: end procedure
```

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Deadline: Nov 14 2018

(Midnight)

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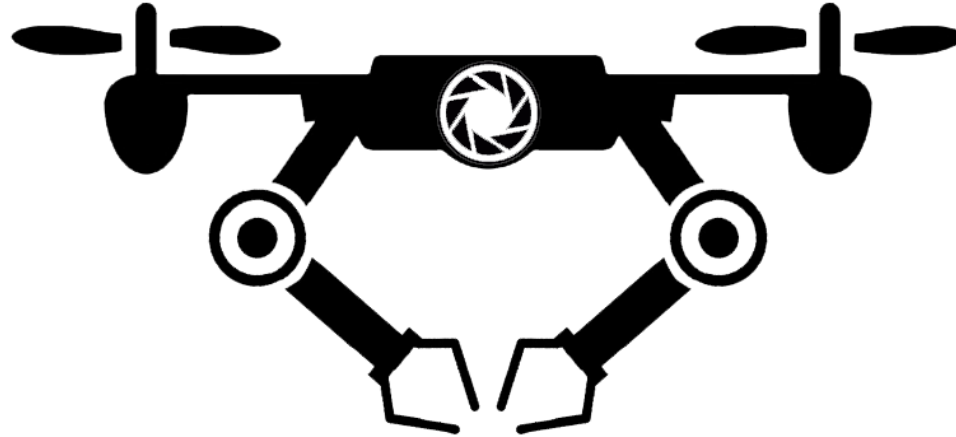
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Thank You!



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