

# **Video Segmentation**

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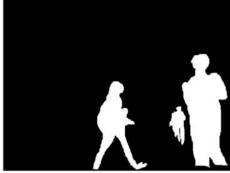
#### Introduction



- What is image/video segmentation?
  - Process of partitioning a digital image into multiple regions
  - Application
    - Chroma-keying
    - Surveillance camera

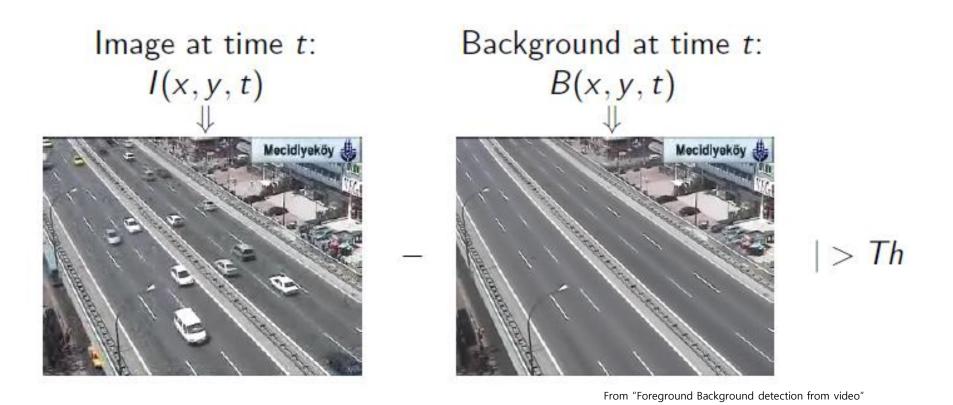








- Given a video, identify the foreground objects in that video
  - In most cases, objects are of interest, not the scene





- Concept
  - Assume we have two image frames f(x, y, t) and B(x, y, t)
  - We detect changes between two images pixel by pixel

$$d(x,y,t) = \begin{cases} 1 \text{ or } 255 & \text{if } |f(x,y,t_i) - B(x,y,t)| > T \\ 0 & \text{otherwise} \end{cases}$$

- The difference with value 1 are considered the result of object motion
- Assumption
  - Two images are registered spatially
  - Illumination is relatively constant



- Key to successful background subtraction
  - We should handle sudden or gradual illumination changes
  - Repetitive motion
    - Tree leaves
    - waves
  - Long-term scene change
    - unattended bag
    - parked car
  - → Estimating good background is the key!



- Background estimation
  - Mean filter
    - Background is the mean of the previous n frames

• 
$$B(x, y, t) = \frac{1}{n} \sum_{i=0}^{n-1} f(x, y, t - i)$$
 or

• 
$$B(x, y, t) = \frac{1}{n} \sum_{i=0}^{n-1} f(x, y, i)$$







n = 10

n = 20

n = 50



- Background estimation
  - Median filter
    - Background is more likely to appear in a scene
    - B(x, y, t) = median(f(x, y, t i)) or
    - B(x, y, t) = median(f(x, y, i))







n = 10

n = 20

n = 50



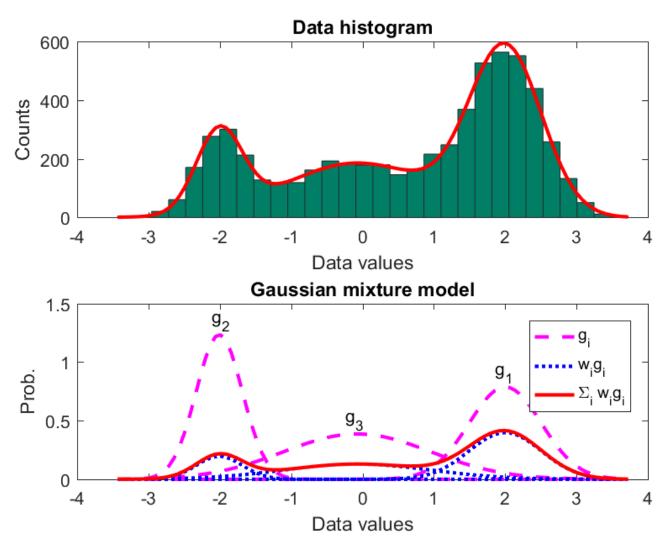
- Conditional probabilities
  - $p(A \cap B) = p(A|B)p(B) = p(B|A)p(A)$
- Bayes rule

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

- Assume A is pixel value, and B is background
  - If we can figure out(or estimate) p(A|B), then we can find out the probability of a certain pixel value being background  $\rightarrow p(B|A)$



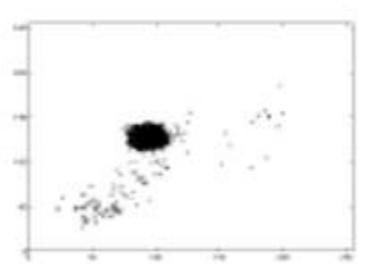
Gaussian mixture model(or mixture of Gaussian)





- Background estimation using GMM
  - Determine the number of mode of GMM
  - At the training stage, estimate mean and variance of each Gaussian model with the training data
    - $\rightarrow$  estimate p(A|B)
  - Each pixel is classified into background/foreground by calculating p(B|A)



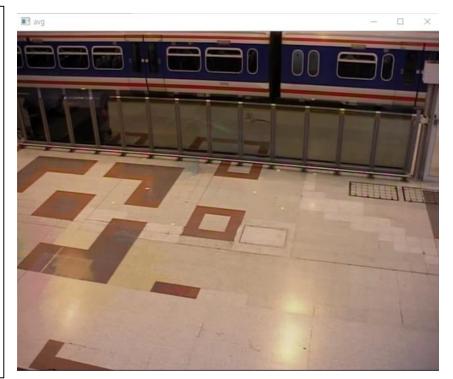


From "Foreground Background detection from video"



- Example code
  - Generating average image

```
int main() {
     VideoCapture capture("background.mp4");
     Mat image, sum, avg;
     int cnt = 2;
     capture >> avg;
     while (true) {
           if (!capture.read(image)) break;
           add(image / cnt, avg*(cnt - 1) / cnt, avg);
           imshow("avg", avg);
           cnt++;
           waitKey(33);
```





Example code(using absdiff)

```
int main() {
      VideoCapture capture("background.mp4");
      Mat background, image, gray, result, foregroundMask, foregroundImg;
      //set the first frame as background
      capture >> background;
      cvtColor(background, background, CV BGR2GRAY);
      while (true) {
            if (capture.grab() == 0) break;
            capture.retrieve(image);
            cvtColor(image, gray, CV BGR2GRAY);
            absdiff(background, gray, foregroundMask);
            threshold(foregroundMask, foregroundMask, 50, 255, CV_THRESH_BINARY);
            foregroundMask.copyTo(foregroundImg);
            gray.copyTo(foregroundImg, foregroundMask);
            imshow("foregroundImg", foregroundImg);
            imshow("foregroundMask", foregroundMask);
            imshow("background", background);
            waitKey(33);
```



Example code(using absdiff)





- Example code
  - openCV MoG2

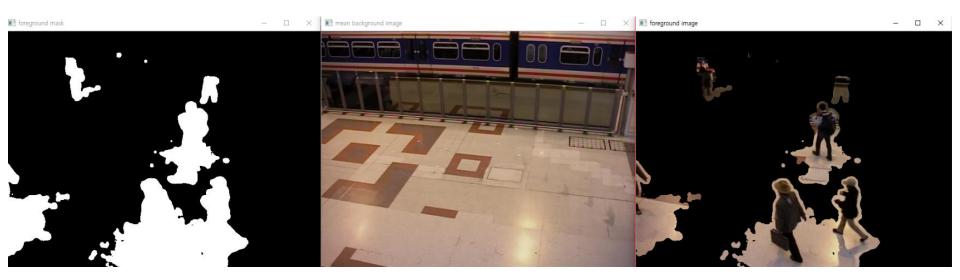
```
int main() {
     Ptr < BackgroundSubtractor > bg_model = createBackgroundSubtractorMOG2();
     Mat image, foregroundMask, backgroundImg, foregroundImg;
     VideoCapture cap("background.mp4");
     while (true) {
          cap >> image;
          resize(image, image, Size(640, 480));
          if (foregroundMask.empty())
                foregroundMask.create(image.size(), image.type());
          bg_model->apply(image, foregroundMask);
          GaussianBlur(foregroundMask, foregroundMask, Size(11, 11), 3.5, 3.5);
          threshold(foregroundMask, foregroundMask, 10, 255, THRESH_BINARY);
          foregroundImg = Scalar::all(0);
          image.copyTo(foregroundImg, foregroundMask);
           bg_model->getBackgroundImage(backgroundImg);
```



- Example code
  - openCV MoG2

```
imshow("foreground mask", foregroundMask);
imshow("foreground image", foregroundImg);

if (!backgroundImg.empty()) {
    imshow("mean background image", backgroundImg);
  }
  waitKey(33);
}
```





#### Counting the number of objects

```
int main() {
    Mat gray = imread("contours.jpg", 0);
    Mat result;
    threshold(gray, result, 180, 255, THRESH_BINARY_INV);
    vector < vector < Point > > contours;
    vector < Vec4i > hierarchy;
    findContours(result, contours, hierarchy, CV_RETR_EXTERNAL, CV_CHAIN_APPROX_SIMPLE);

    putText(result, format("contour count: %d", contours.size()), Point(50, 80),
    FONT_HERSHEY_SIMPLEX, 1, Scalar(128), 4);

    imshow("contours", result);
    waitKey(0);
}
```

