



# Computer Vision

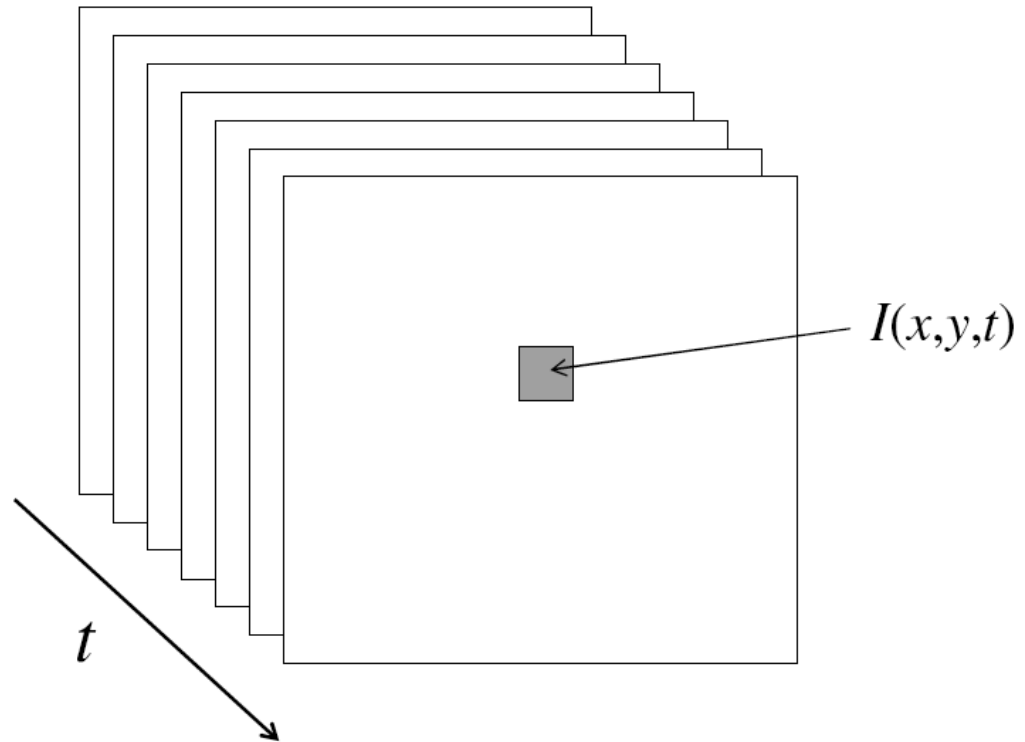
(Course Code: 4047)

## Module-3:Lecture-1: Background Subtraction and Modeling

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# Video

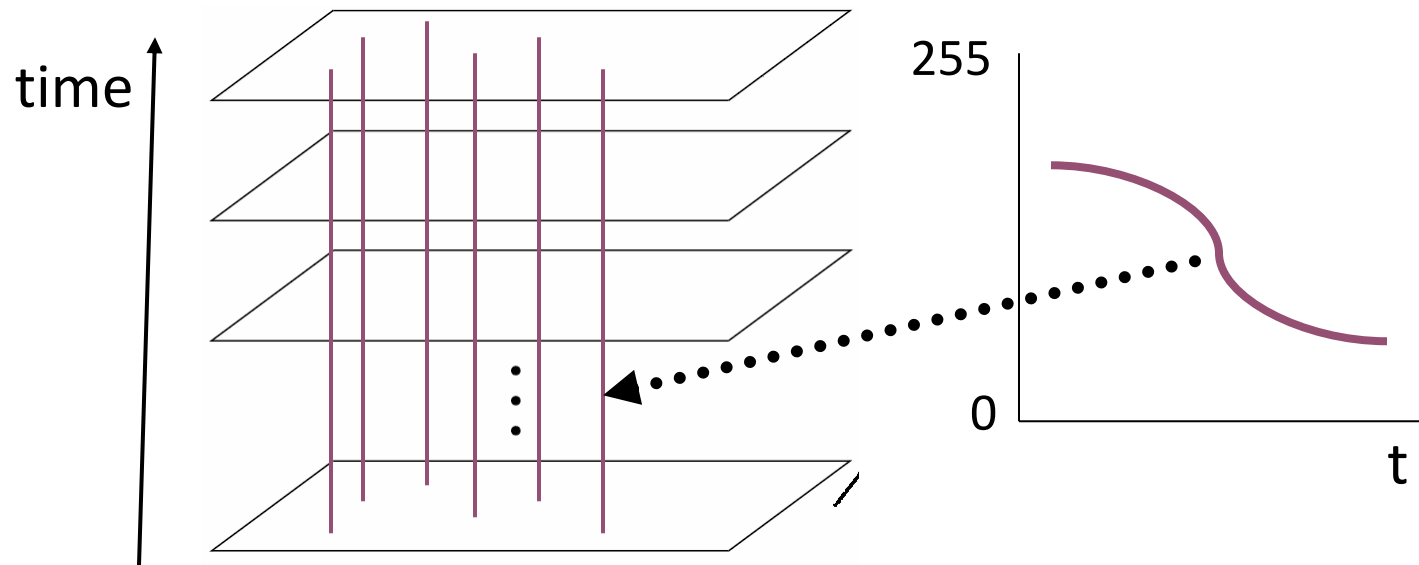
- A video is a sequence of frames captured over time
- Now our image data is a function of space (x, y) and time (t)



# Video as an “Image Stack”

❖ Can look at video data as a spatio-temporal volume

➤ If camera is stationary, each line through time corresponds to a single ray in space



# Background Subtraction

❖ Problem: separate the (static) background from the (dynamic) foreground.

❖ Applications:

- surveillance
- tracking of moving objects
- classification

❖ General Approach:

➤ For each frame  $i$  let

$x[n_1, n_2, i]$     intensity

$b[n_1, n_2, i]$     background

➤ Then     $|x[n_1, n_2, i] - b[n_1, n_2, i]| > threshold \quad \Rightarrow (n_1, n_2) \in object$

# Issues with the Background

## ❖ Illumination Changes

- Gradual Daily Changes (day/night)
- sudden (clouds)

## ❖ Motion

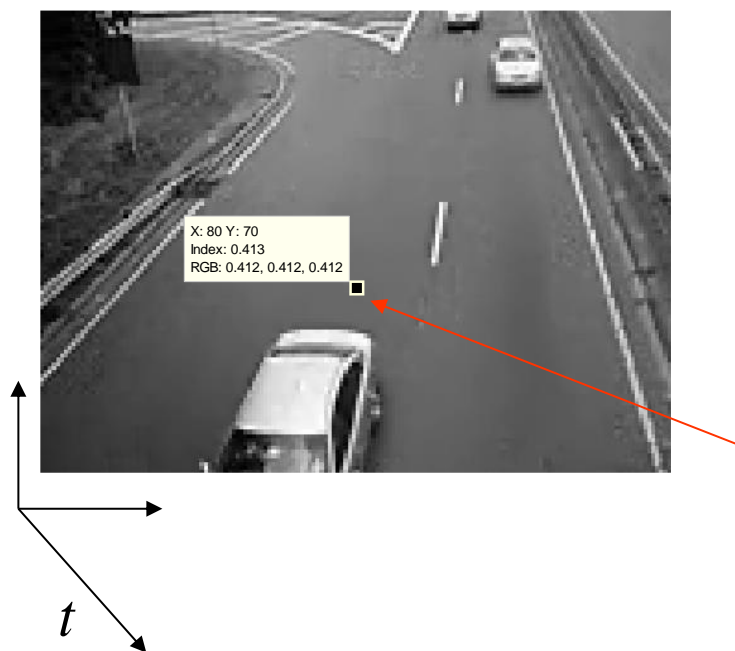
- Camera oscillations;
- small motions of trees, leaves, ...

## ❖ Changes in Background

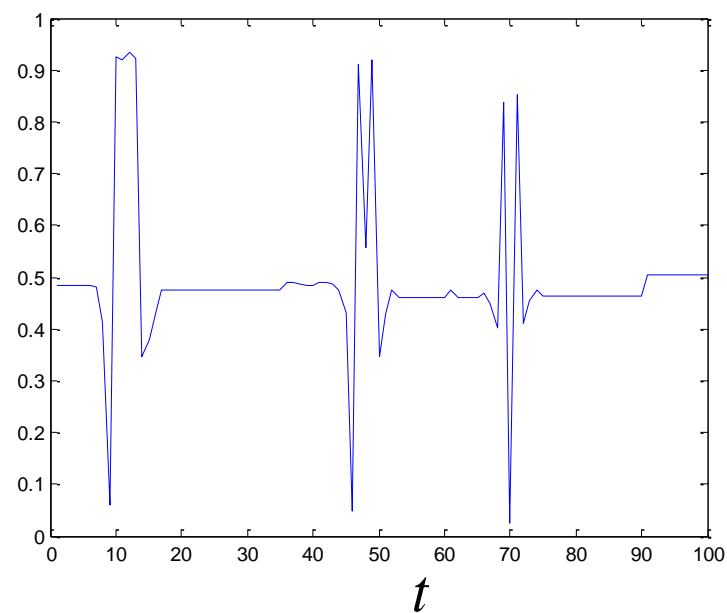
- Parked Cars and others

# Approach

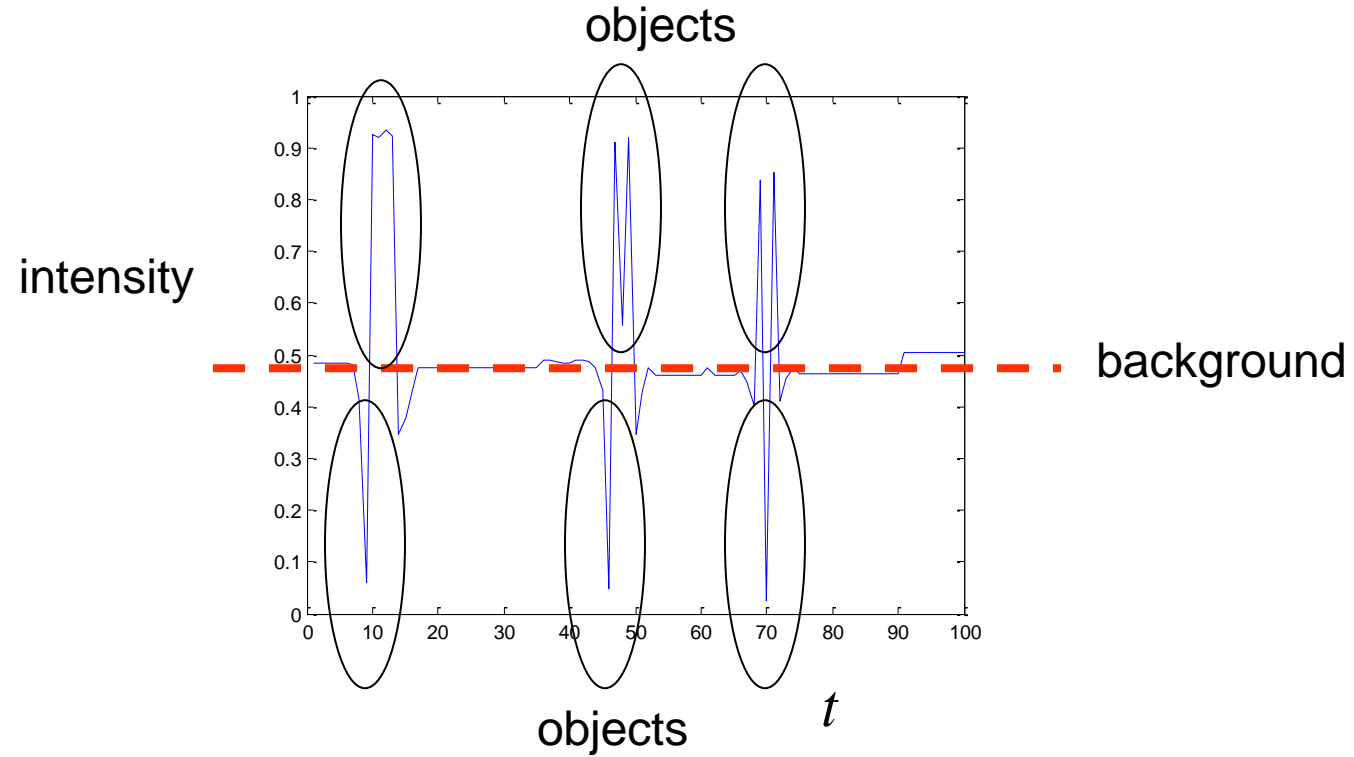
- ❖ Most approaches based on the time domain.
- ❖ Take one point and see how it changes in time:



intensity



Here it is easy to separate the background from the objects:

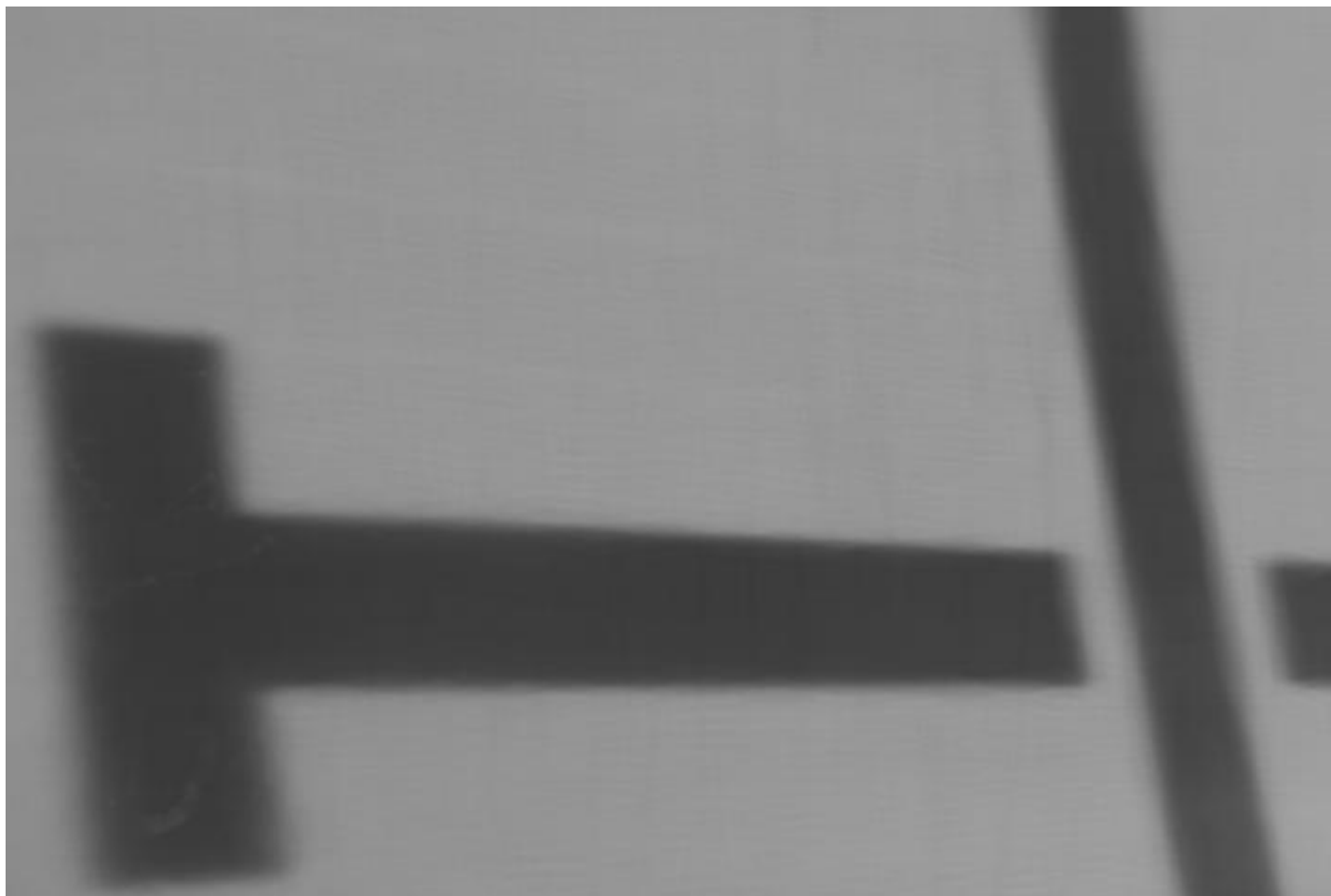


# Input Video





## Average Image



# Background subtraction

- ▶ Given an image (mostly likely to be a video frame), we want to identify the **foreground objects** in that image!



## Motivation

- ▶ In most cases, objects are of interest, not the scene.
- ▶ Makes our life easier: less processing costs, and less room for error.

# Background subtraction

- ❖ Simple techniques can do ok with static camera

- ❖ ...But hard to do perfectly

- ❖ Widely used:

- Traffic monitoring (counting vehicles, detecting & tracking vehicles, pedestrians),
- Human action recognition (run, walk, jump, squat),
- Human-computer interaction
- Object tracking

# Simple Approach

Image at time  $t$ :

$$I(x, y, t)$$



Background at time  $t$ :

$$B(x, y, t)$$



$$| > Th$$

1. Estimate the background for time  $t$ .
2. Subtract the estimated background from the input frame.
3. Apply a threshold,  $Th$ , to the absolute difference to get the **foreground mask**.

But, how can we estimate the background?

# Frame Differencing

- ▶ Background is estimated to be the previous frame.  
Background subtraction equation then becomes:

$$B(x, y, t) = I(x, y, t - 1)$$



$$|I(x, y, t) - I(x, y, t - 1)| > Th$$

- ▶ Depending on the object structure, speed, frame rate and global threshold, this approach may or may **not** be useful (usually **not**).



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| > Th

# Frame Differencing

$Th = 25$



$Th = 50$



$Th = 100$



$Th = 200$





# Mean Filter

- In this case the background is the mean of the previous  $n$  frames:

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

$\Downarrow$

$$|I(x, y, t) - \frac{1}{n} \sum_{i=0}^{n-1} I(x, y, t - i)| > Th$$

- For  $n = 10$ :

Estimated Background



Foreground Mask



# Median Filter

- ▶ Assuming that the background is more likely to appear in a scene, we can use the median of the previous  $n$  frames as the background model:

$$B(x, y, t) = \text{median}\{I(x, y, t - i)\}$$

$\Downarrow$

$$|I(x, y, t) - \text{median}\{I(x, y, t - i)\}| > Th \text{ where} \\ i \in \{0, \dots, n - 1\}.$$

- ▶ For  $n = 10$ :

Estimated Background



Foreground Mask





# Average/Median Image



Previous N frame image



Average/median of these images  
(Background of the image)

# Background Subtraction



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# Pros and cons

## Advantages:

- ❖ Extremely easy to implement and use!
- ❖ All pretty fast.
- ❖ Corresponding background models need not be constant, they change over time.

## Disadvantages:

- ❖ Accuracy of frame differencing depends on object speed and frame rate
- ❖ Median background model: relatively high memory requirements.
- ❖ Setting global threshold  $Th...$

*When will this basic approach fail?*

# References

❖ Background Subtraction

❖ [https://www.cs.utexas.edu/~grauman/courses/fall2009/slides/lecture9\\_background.pdf](https://www.cs.utexas.edu/~grauman/courses/fall2009/slides/lecture9_background.pdf)