

# Detection

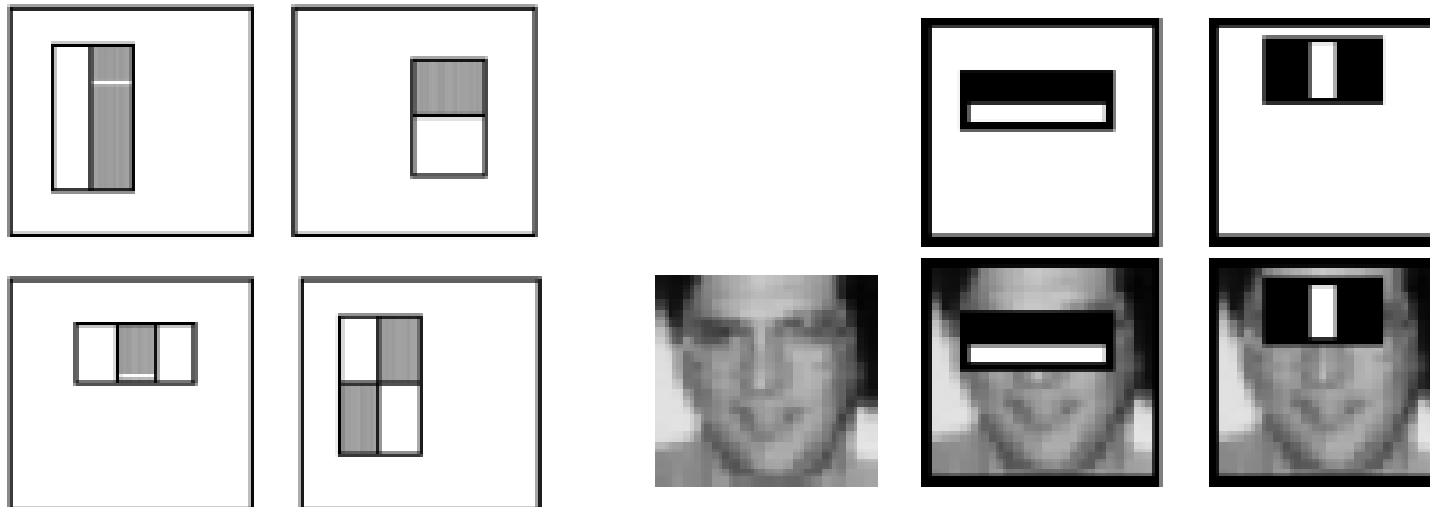
Sung Soo Hwang

# Introduction

- How can we make computers detect objects?
  - Training stage
    - Collect a large amount of object images and non-object images
    - Find features suitable to represent objects
    - Design a classifier(or threshold) to classify objects
  - Test stage
    - Extract features from the input image
    - Detect objects using the trained classifier

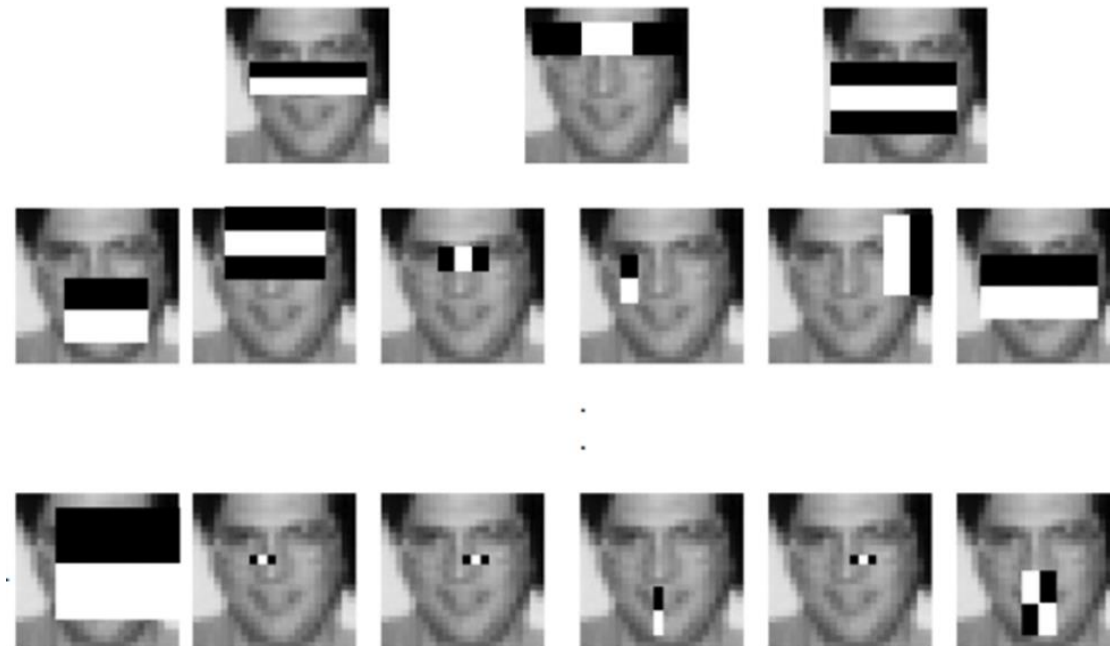
# Face Detection

- Feature
  - Harr-like feature is used in openCV
    - It can be defined as the difference of the sum of pixels of areas inside the rectangle
    - sum of pixels under black rectangle – sum of pixels under white rectangle
    - It can be at any position and scale within the original image



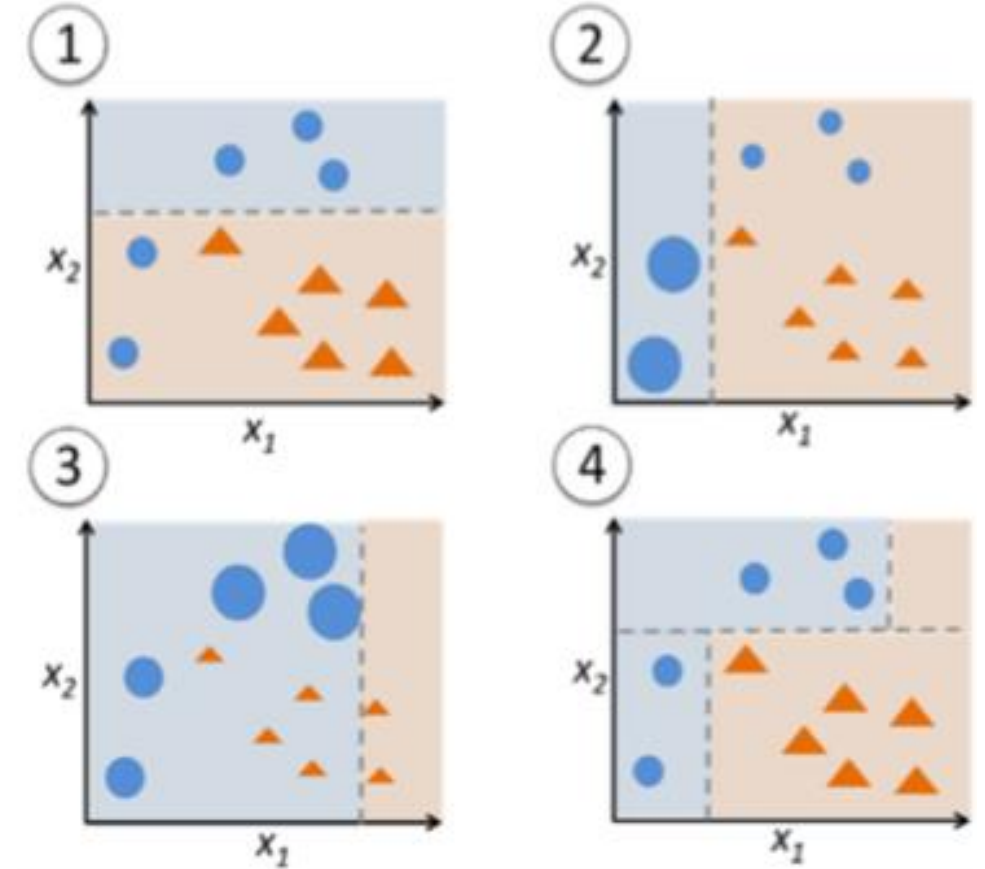
# Face Detection

- Training
  - By changing size and location, lots of features can be generated
  - Among them, choose features that classify human faces



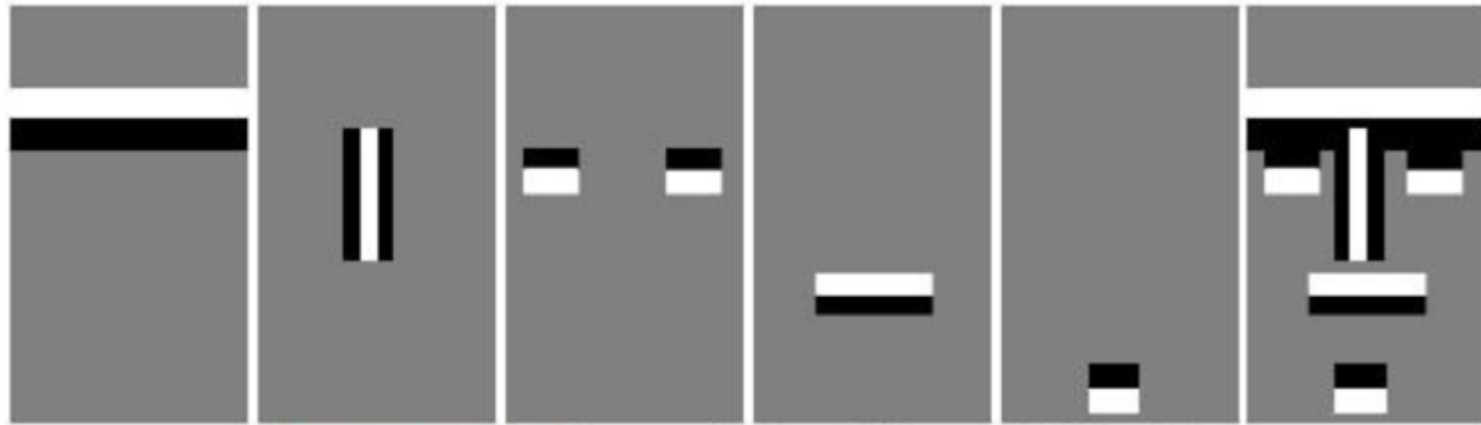
# Face Detection

- Training
  - In openCV, Adaboost(Adaptive Boosting) is used
  - Boosting: A set of weak-learner generates a strong-learner
  - Adaptive: Weight of each sample is adjusted depending on the accuracy of already-trained weak learners



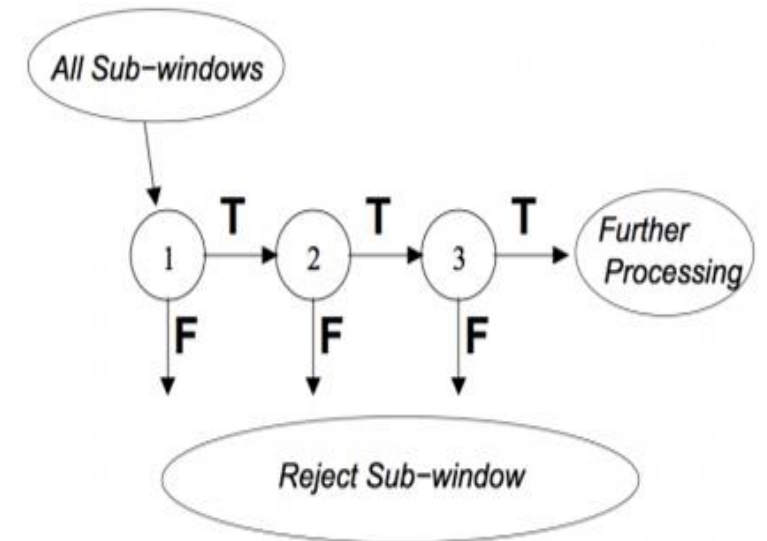
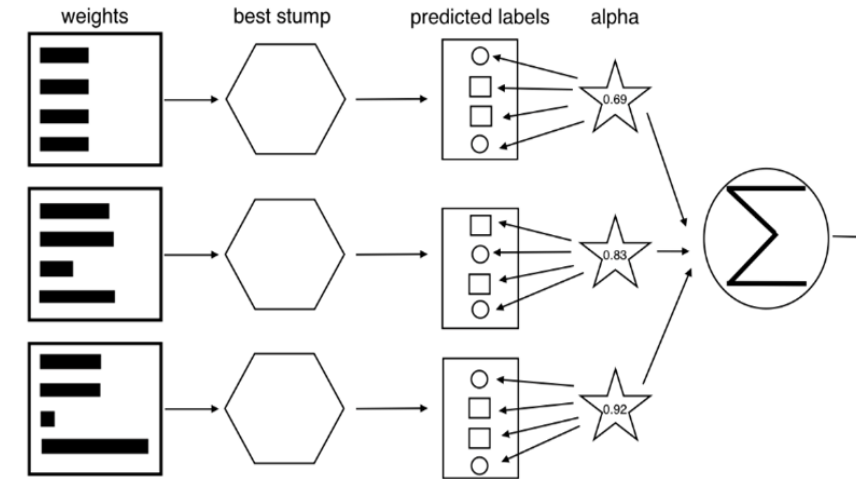
# Face Detection

- Training
  - Face representation using Harr-like features



# Face Detection

- Cascade classifier
  - Generate a strong learner using multiple weak learners
  
- Each strong learner is connected in cascade
  - Number of weak learners in each strong learner:  $3 > 2 > 1$
  - Lots of non-face regions are easily eliminated



# Face Detection

- Fast computation of Harr-like features(sum of pixels under a rectangle) can be conducted by using integral image

1	2	2	4	1
3	4	1	5	2
2	3	3	2	4
4	1	5	4	6
6	3	2	1	3

input image

0	0	0	0	0	0
0	1	3	5	9	10
0	4	10	13	22	25
0	6	15	21	32	39
0	10	20	31	46	59
0	16	29	42	58	74

integral image



# Tracking

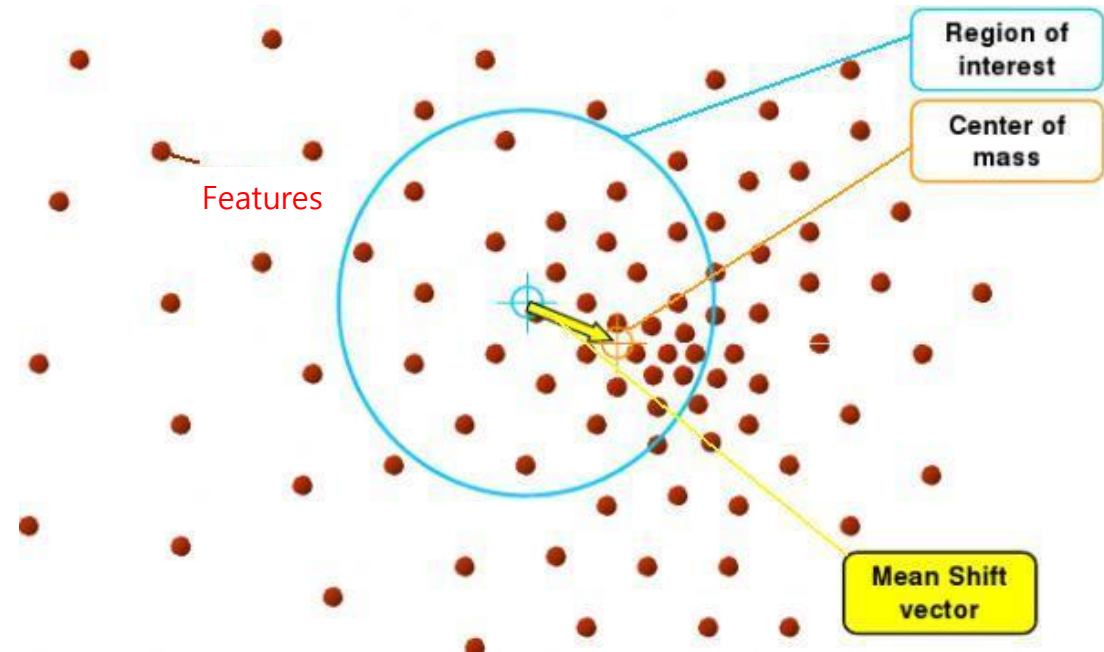
Sung Soo Hwang

# Introduction

- Basic concept
  - First, a ROI is selected by user-interaction or detection
  - Represent the ROI with histograms or features
  - Find the best matching patch to the ROI at the next frame

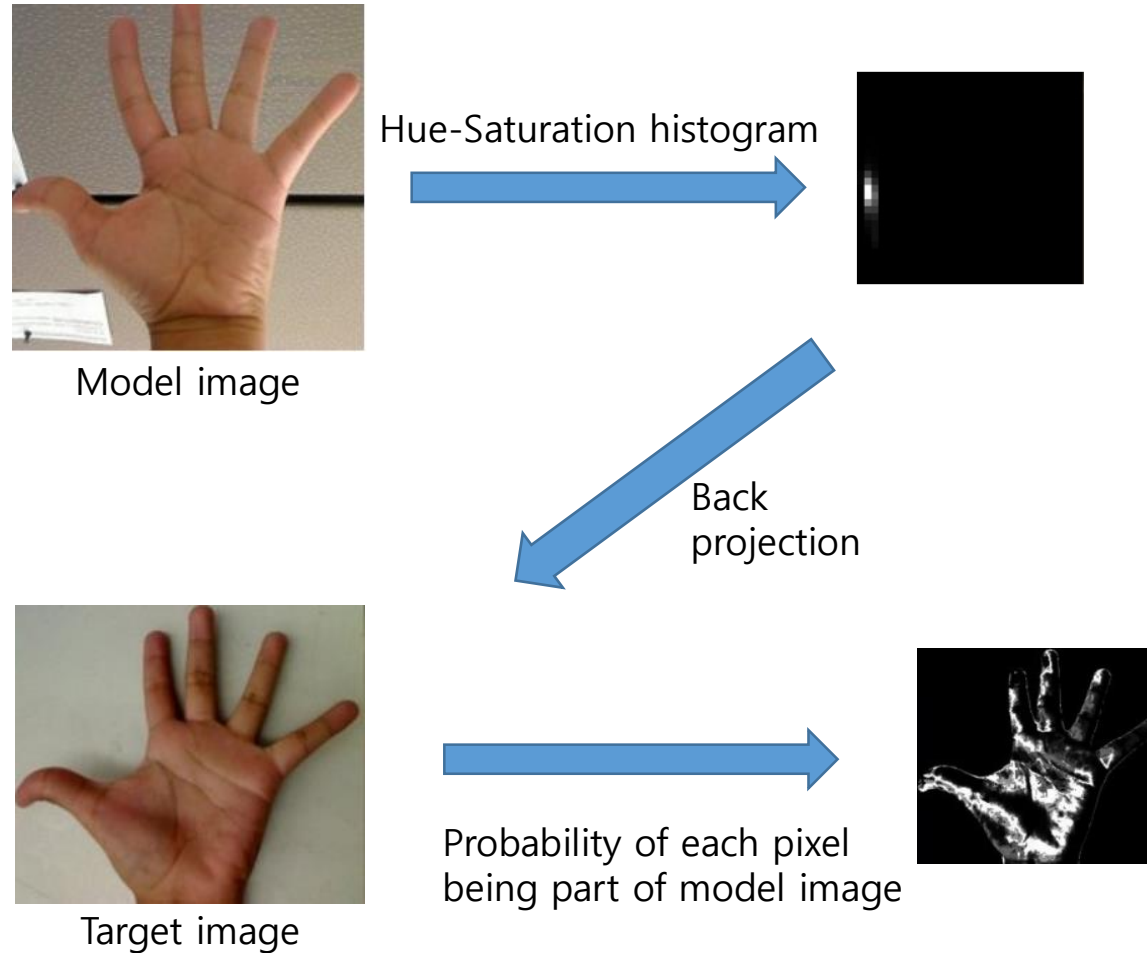
# Meanshift

- It is a procedure for locating the maxima of a density function given discrete data sampled from that function
- It is an iterative method



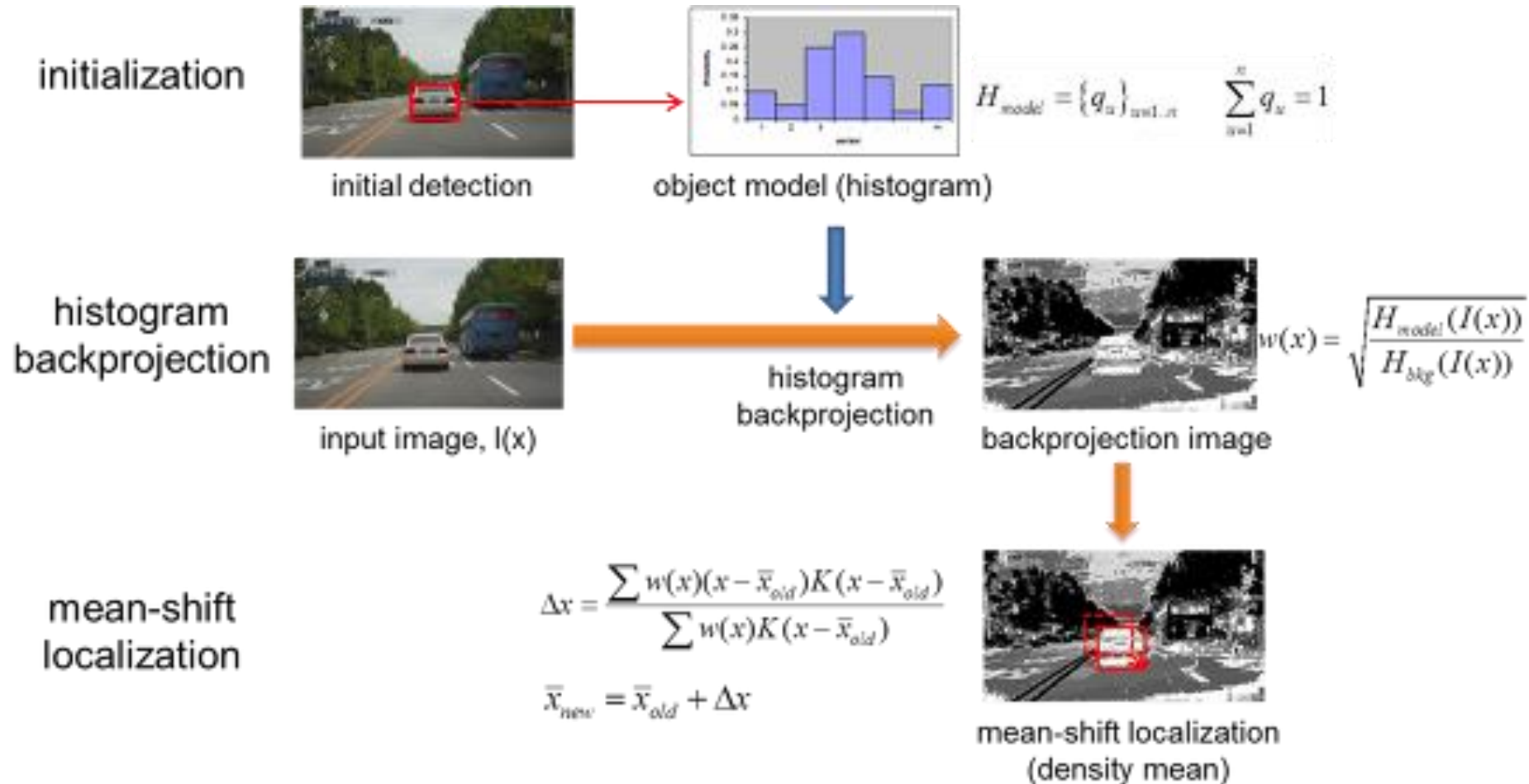
# Meanshift

- Histogram back-projection



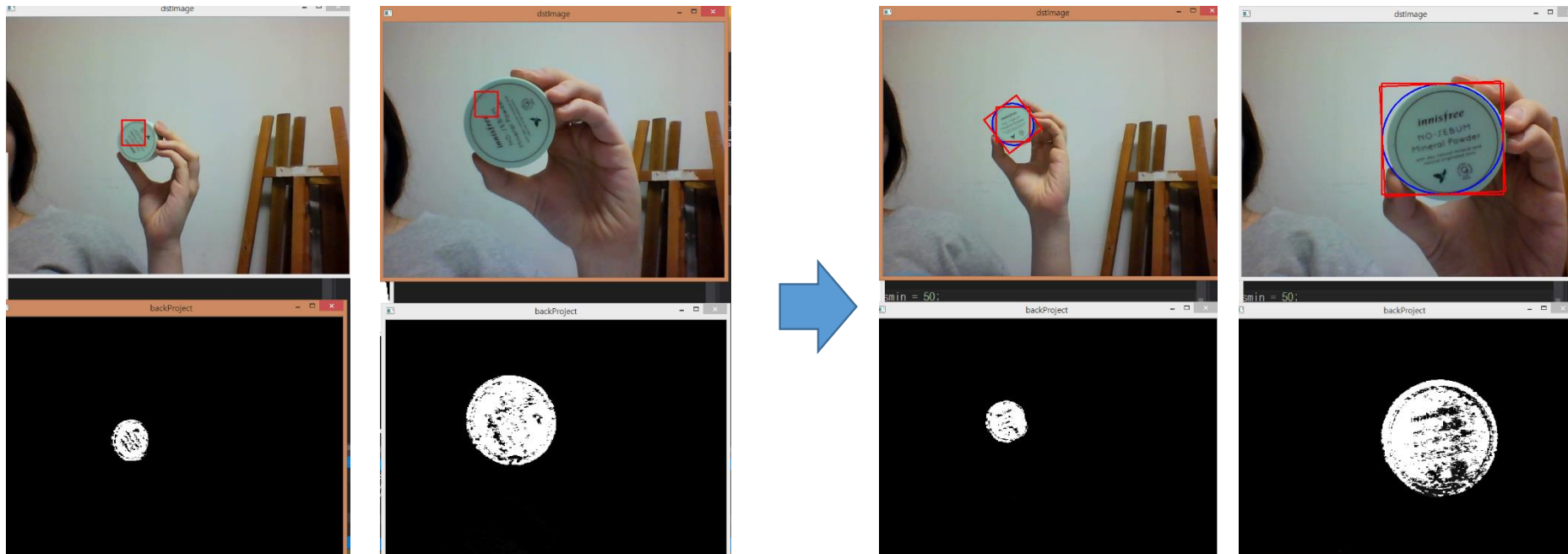
# Meanshift

- Tracking using mean shift



# Camshift

- Cam shift
  - Modified version of mean-shift
  - The size of search window can be changed

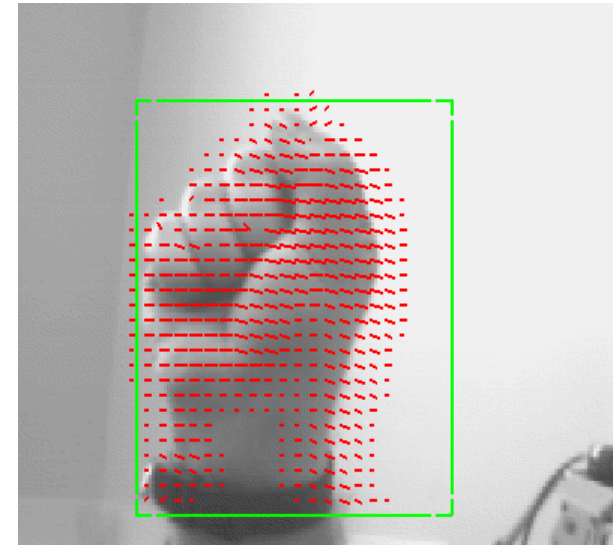
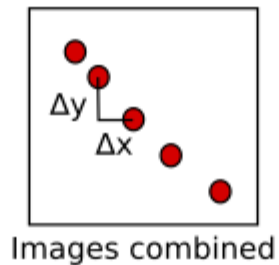
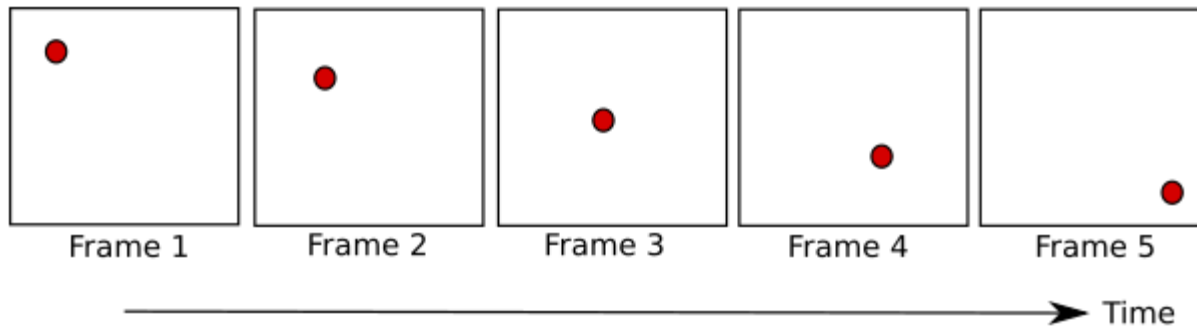


Mean-shift

Cam shift

# Optical Flow

- Optical flow is the apparent motion of brightness patterns in the image



# Optical Flow

- KLT algorithm
  - Assumption
    - Intensity of objects are not changed over consecutive frames
    - Movement of pixels are similar to that of adjacent pixels
  - ➔  $I(x, y, t) = I(x + \Delta x, y + \Delta y, t + \Delta t)$
  - By applying Taylor series
    - $I(x + \Delta x, y + \Delta y, t + \Delta t) = I(x, y, t) + \frac{\partial I}{\partial x} \Delta x + \frac{\partial I}{\partial y} \Delta y + \frac{\partial I}{\partial t} \Delta t$
    - ➔  $\frac{\partial I}{\partial x} \Delta x + \frac{\partial I}{\partial y} \Delta y + \frac{\partial I}{\partial t} \Delta t = 0$
- Extract features first and track the extracted features



# Optical Flow

- KLT algorithm with pyramids
  - Original KLT algorithm cannot handle large movement
  - To overcome this limitation, image pyramid is used

