



# Neural Networks – Term Project

## **Neural Network based Video Surveillance System for Intruder Detection from Complex Background**

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# Motivation

- ◆ Automated video surveillance technologies as well as intrusion detection systems are important research areas for it widespread applications in diverse disciplines.
- ◆ These system isolates the events of potential objects from a large volume of redundant image data, since human observers can easily get distracted from these tasks.
- ◆ Challenges: They should have robustness against
  - Illumination changes
  - Color changes
  - Shadows
  - Avoid false alarm generation.
- ◆ Since EBPN stores information in terms of weights, we go for Neural Network based Intruder Detection System.

# Problem Definition

- ◆ A basic moving object recognition algorithm takes the image sequence as input, detects frame having significant changes from the previous frames or background and identifies the changed region.
- ◆ Here we will take one reference image as background and check in subsequence images whether they have an intruder or not.
- ◆ The system will be implemented on Error Back Propagation algorithm initially trained with some sample training frames with/without intruders.
- ◆ Test Images will be checked for Intruders and set Alarm.
- ◆ Tools: VC++, MTES(Multimedia Technology Education System)

# Sample Scenario



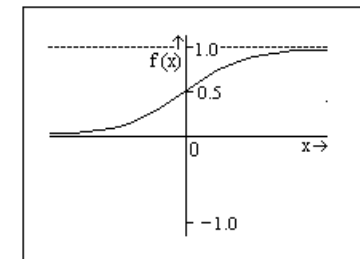
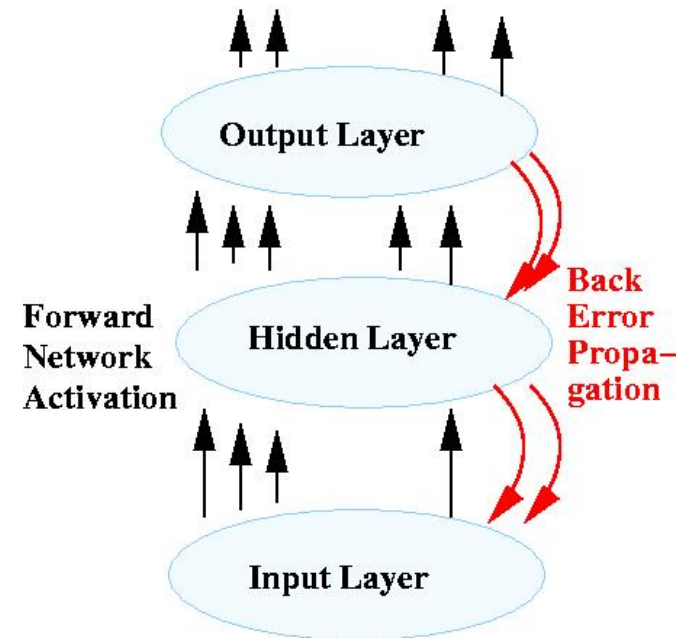
- Video Surveillance System captured image of the area being monitored.
- Safe as long as no one else is there.



- Video Surveillance System captured image of something else than the usual captured scene.
- Potential threat to security.
- Generate Alarm

# Error Back Propagation

- ◆ **Training Phase:** Training a network involves 3 stages:
  - Feed forward of the input training patterns.
  - Back propagation of associated error computed from known outputs of training patterns.
  - Adjustment of the weights.
- ◆ **Sigmoid function used as activation function**
  - Its output range (0 to 1) is perfect for learning to output Boolean values.

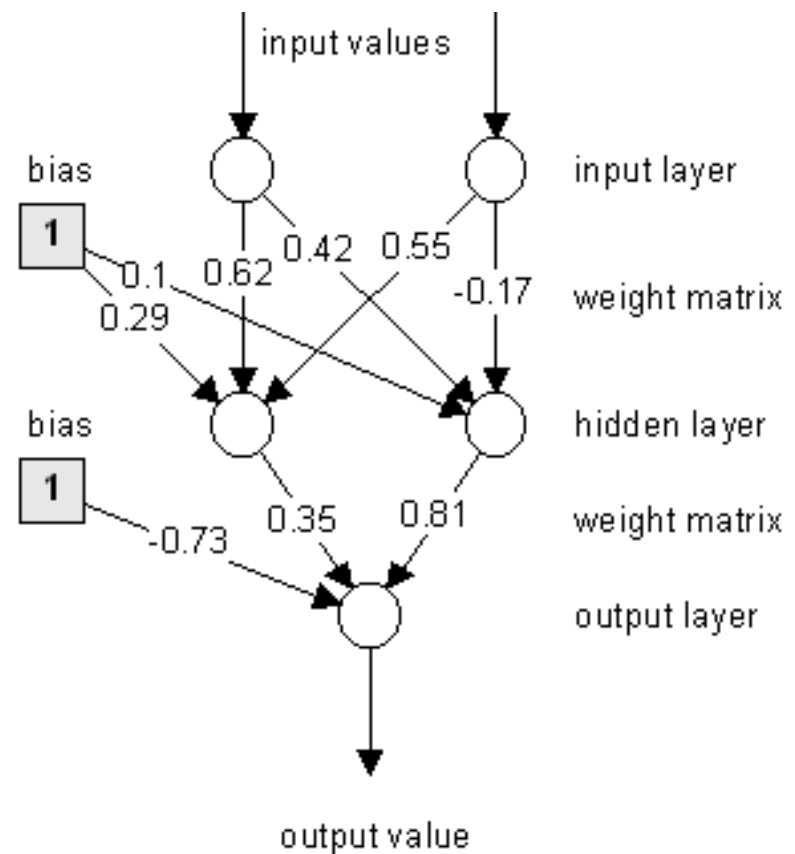


# Error Back Propagation Contd..

- ◆ We might also have some bias, which acts like weight on connections from units whose output is always 1.
- ◆ Modification of weights

$$\delta_j \text{ For Output layer } (d_j - y_j^o)$$

$$\delta_j \text{ For Hidden layer } \sum_k \delta_k w_{jk}$$



# System Specification

## ◆ Learning Phase

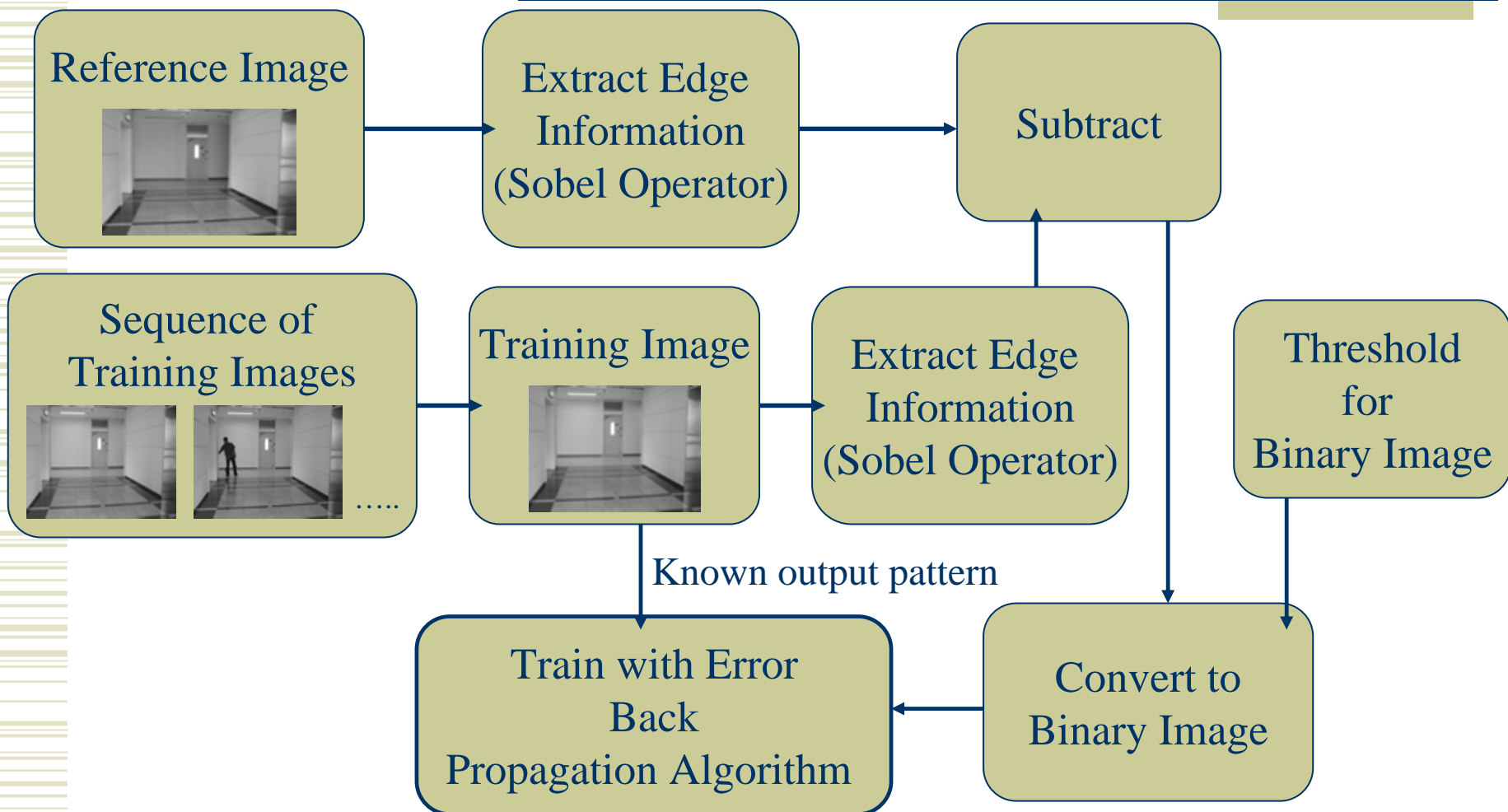
- Input: One reference image and six training images with known outputs: 3 of them are with intruder and 3 without intruder
- Images are of size  $100 \times 75$  pixels. So input layer of EBPN has 7500 nodes.
- Hidden layer is chosen to have 100 hidden nodes (if Multi-layer Perceptron network).
- Since we want to know whether intruder is present or not, output layer has only 1 output node. If present then 1 otherwise 0 for absence of intruder.



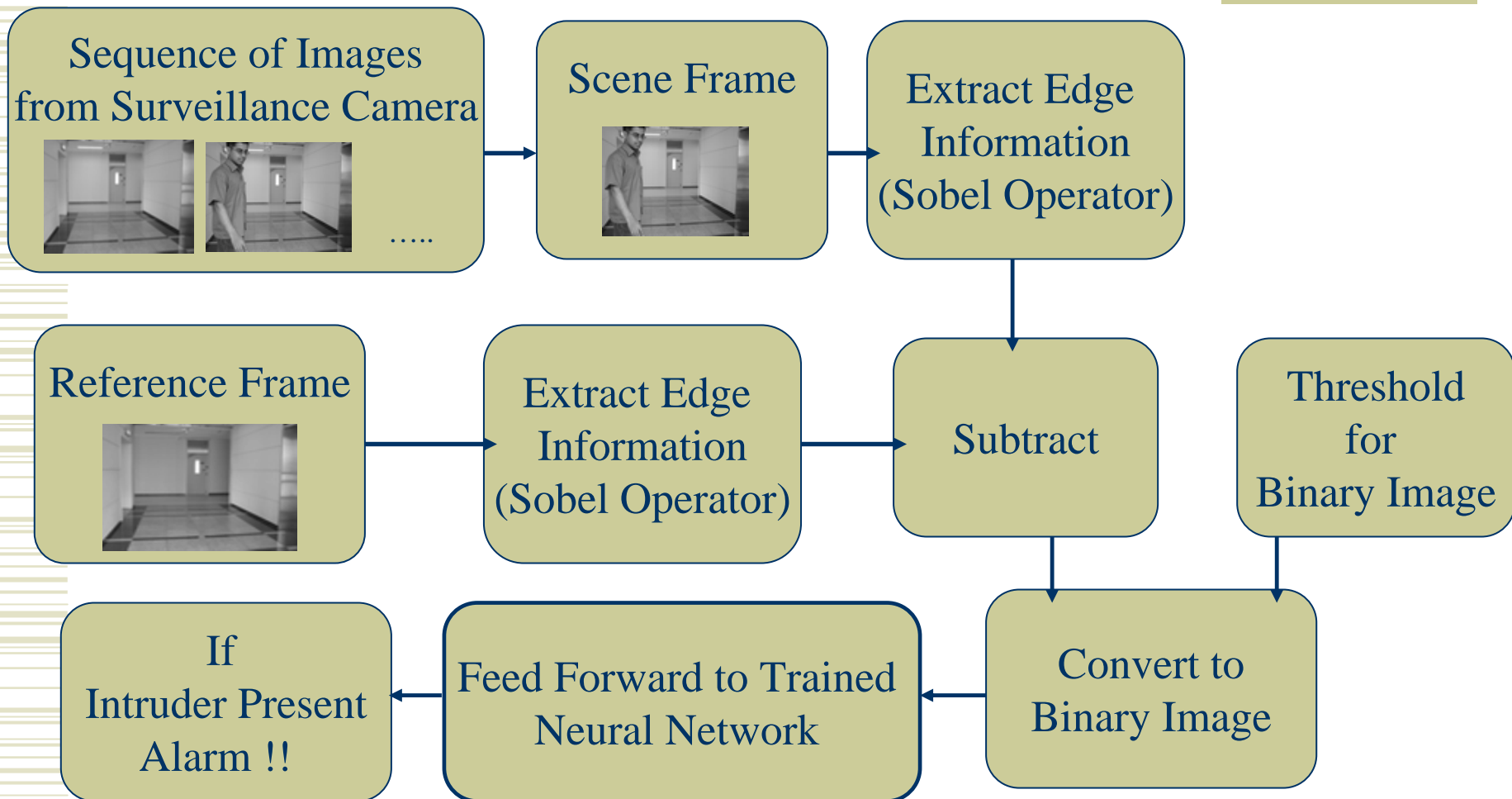
# System Specification Contd..

- ◆ Instead of using the whole image, we feed edge information.
- ◆ **Preprocessing:** Most information lies on the boundaries of different regions. Extraction of edges from an image
  - Significantly reduce the amount of data
  - Filter out the useless information.
  - Less sensitive to noise.
  - Consistent more that illumination changes.
- ◆ We use Sobel operator to get the gradient information of image. Actually gives us edge information.
- ◆ As we want to know the change in scene, we take image difference.
- ◆ Finally convert image into binary image with a predetermined threshold prior to Neural Network feeding.

# System Overview (Learning)



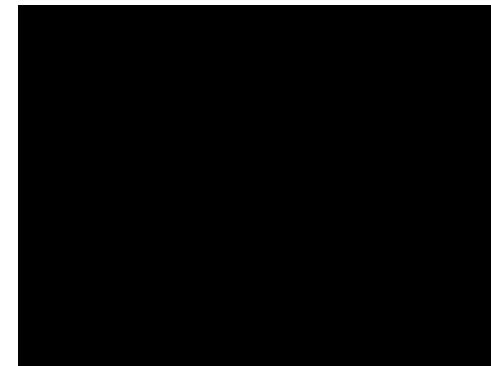
# System Overview (Application)



# Experimental Results



Reference  
Image



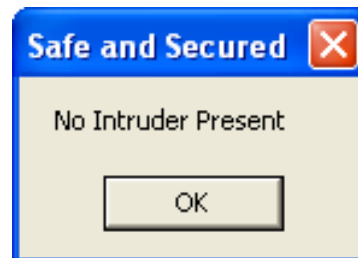
Edge(Gradient) Information  
after Subtraction.

[Ideal case with exact match]

# Experimental Results Contd..



Another scene frame



Edge(Gradient) Information after Subtraction.



# Experimental Results Contd..



Another scene frame



Edge(Gradient) Information after Subtraction.





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# Discussion

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- ◆ Since Training sample was not large enough (Only Six training patterns were taken for this System), Some false alarm generates or Intruder is not detected.
- ◆ Fast operating and Requires less human interaction

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# Thank You

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MTES can be downloaded from  
<http://vision.khu.ac.kr/hellovision/download/MTES.zip>