

# LOCAL-GLOBAL BACKGROUND MODELING FOR ANOMALY DETECTION IN HYPERSPPECTRAL IMAGES

Eyal Madar, Oleg Kuybeda,  
David Malah, and Meir Barzohar

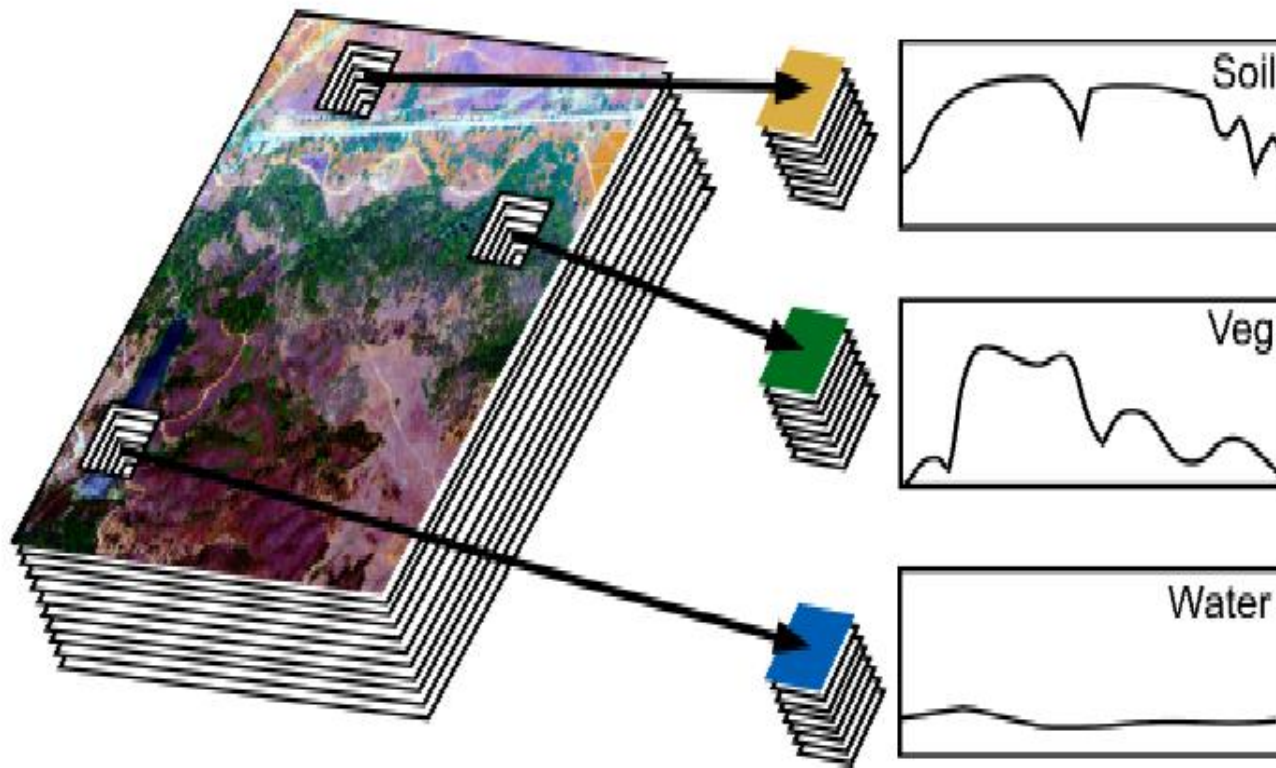
Department of Electrical Engineering  
Technion - Israel Institute of Technology

# Outline

- Hyperspectral Imaging
- Statistical Background Modeling
  - ▣ Local Approach
  - ▣ Global Approach
- Combined Local-Global Approach
- Proposed Algorithm
  - ▣ Local Part
  - ▣ Global Part
- Experimental Result

# Hyperspectral Imaging

## Image Acquisition



# Detection in Hyperspectral Imaging

## Unsupervised Detection

- Prior anomaly signatures are unknown
  
- Anomaly detection methods:
  - ▣ Model the background
  - ▣ Detect anomalies by finding pixels that are not well-described by the background model
  
- *Statistical Background Modeling*
  - Local approach
  - Global approach

# Local Approach

## □ **Principle**

Background is estimated in a local neighborhood of a tested pixel.

## □ **Advantages**

Due to the many degrees of freedom, local background models can be tightly fitted to the background data.

## □ **Problem**

- ▣ Too high number of degrees of freedom may cause model overfitting.
- ▣ Insufficient data for parameters estimations of complex local models

# Global Approach

## □ **Principle**

Background modeling is based on the entire image.

## □ **Advantages**

Robust estimation and more resistant to the overfitting problem.

## □ **Problem**

- ▣ Limited ability to adapt to all nuances of the background process (underfitting problem)
- ▣ Difficult optimization process with a lot of local minimums

# Combined Local-Global Approach

## □ **Goal**

Significantly improve detector performance by a proper combination of the local and global background modeling principles.

## □ **Proposed Algorithm**

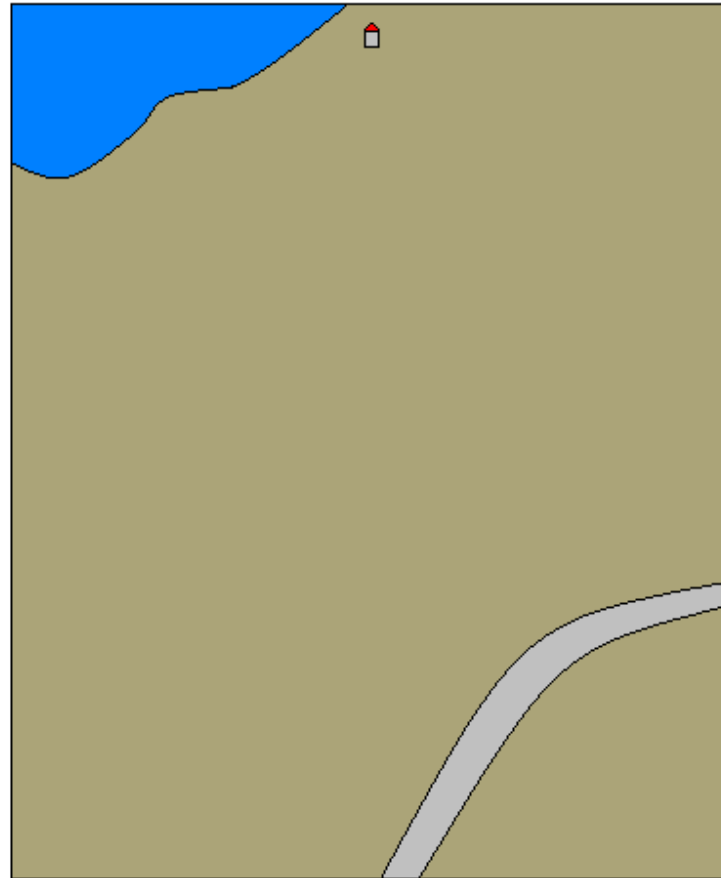
### Local part:

- Greedy sequential estimation process
- Robust Gaussian statistics estimation
- Background cluster hypothesis testing based on extreme value theory results

### Global part:

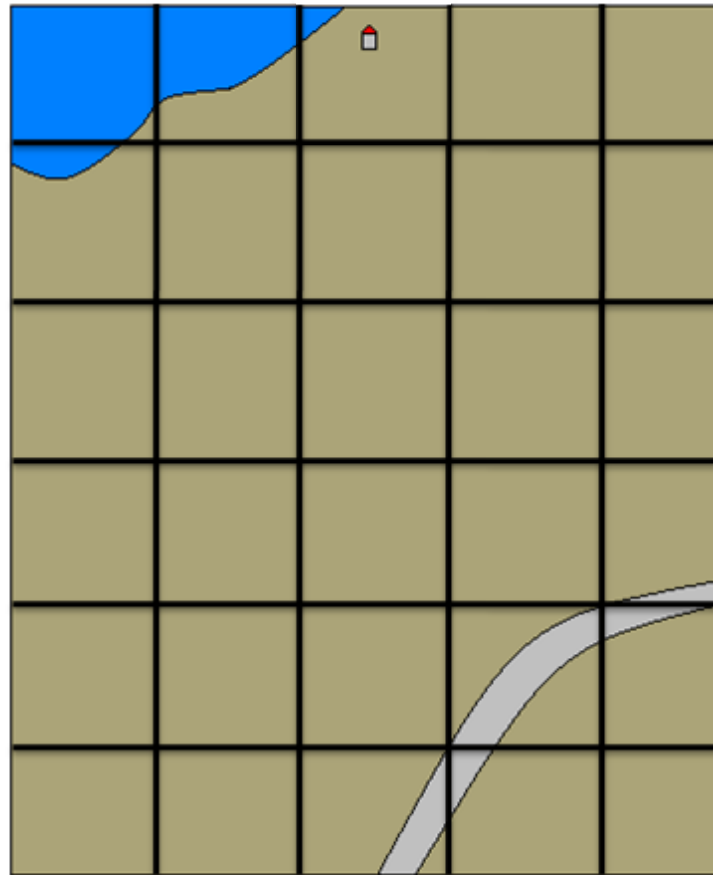
- Global filtering which reduces the number of degrees of freedom

# A combined local-global algorithm





# A combined local-global algorithm

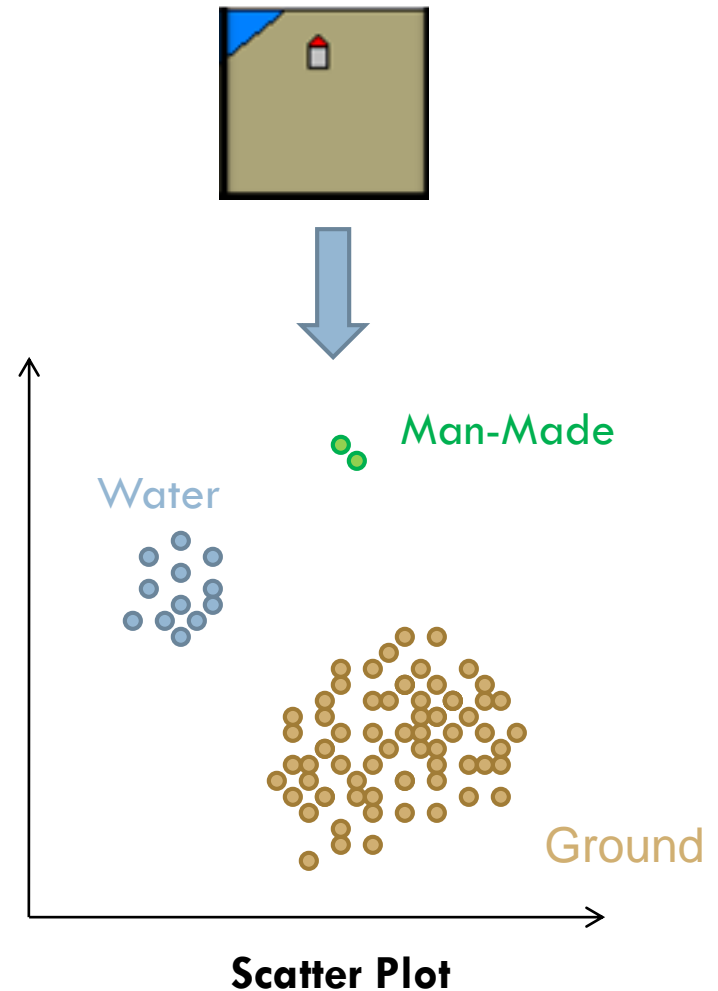


# Local Part Assumption

## □ Local assumption:

The local background model is composed of a small number of distinct clusters,  $L$ , (up to 3), ordered by size, each distributed as a separate Gaussian distribution.

$$\begin{cases} x \in C_k, & 1 \leq k \leq L \\ C_k \sim N(\mu_k, \Gamma_k) \\ |C_1| \geq |C_2| \geq \dots \geq |C_L| \end{cases}$$



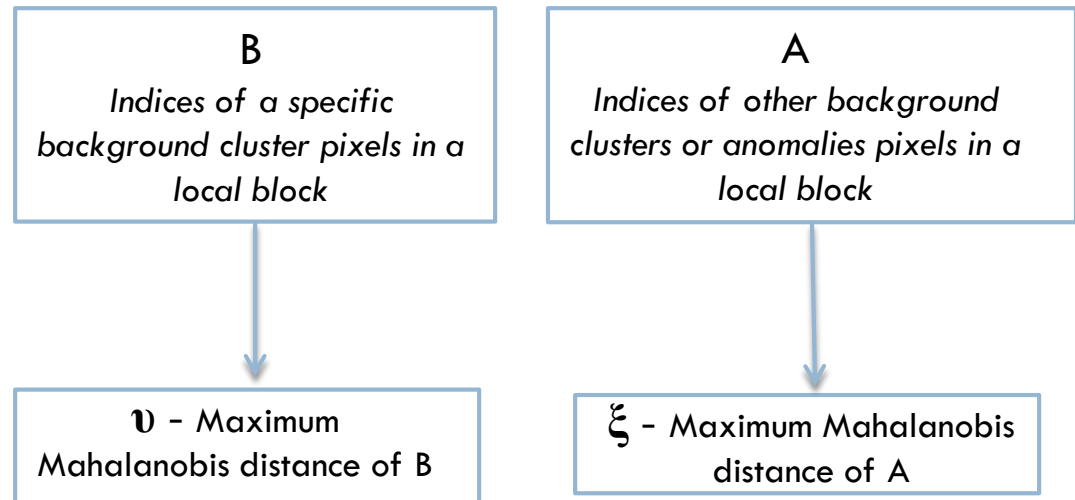
# Background cluster hypothesis test (1)

## □ Goal

An automatic test that isolates pixels belonging to a specific background cluster in a local image area.

## □ Assumption

The cluster statistics (Gaussian mean and covariance) are known



# Background cluster hypothesis test (2)

## □ **Distribution of $v$**

Given by extreme value statistics of maximum-norm Gaussian realizations:

$$P(v \leq x) = G(a_N(x - b_N))$$

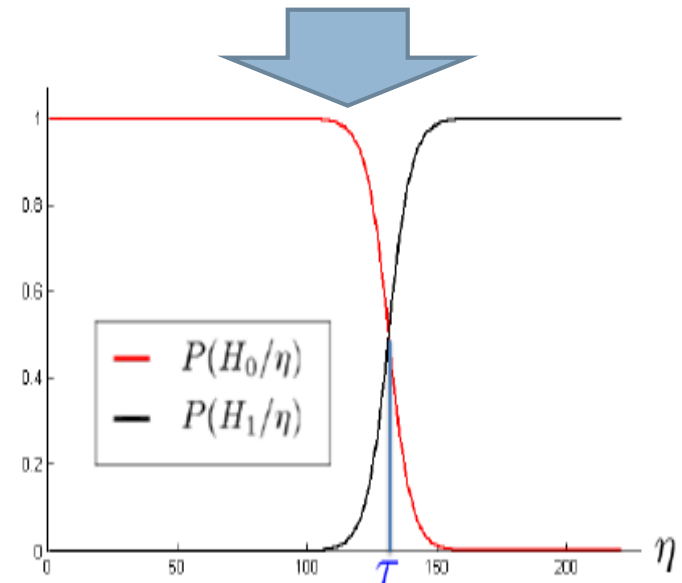
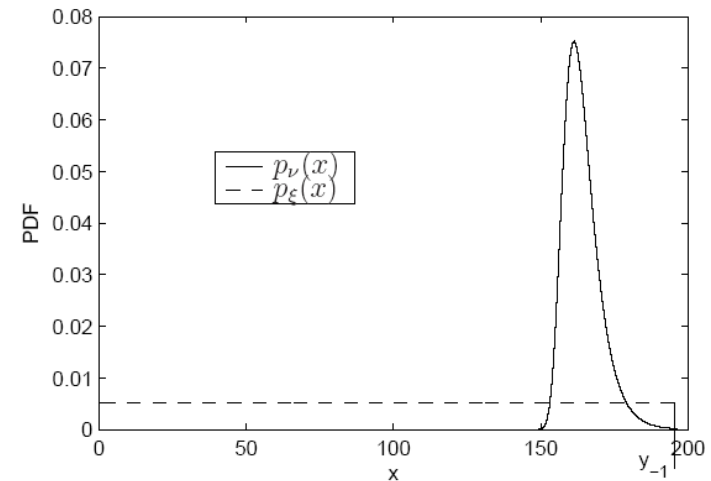
with

$$G(x) = e^{-e^{-x}} \text{ (Gumbel distribution)}$$

## □ **Distribution of $\xi$**

Assumed to be uniformly-distributed

$$\xi \sim U(0, \eta)$$



# Local Part Algorithm

## □ Main Loop :

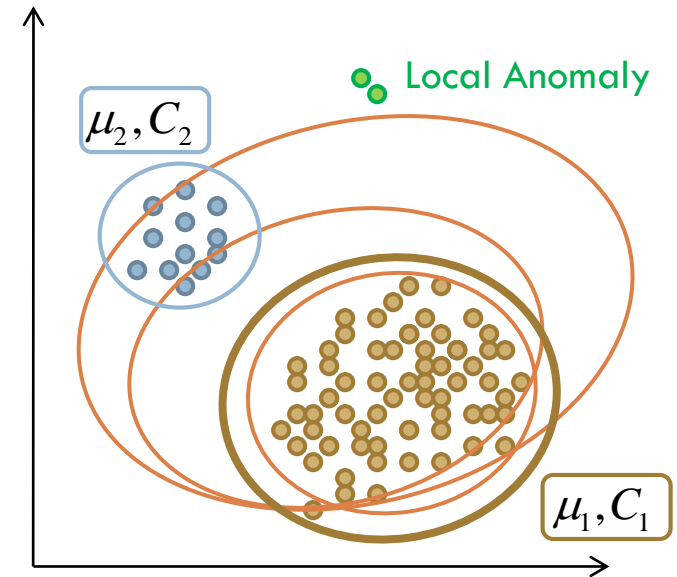
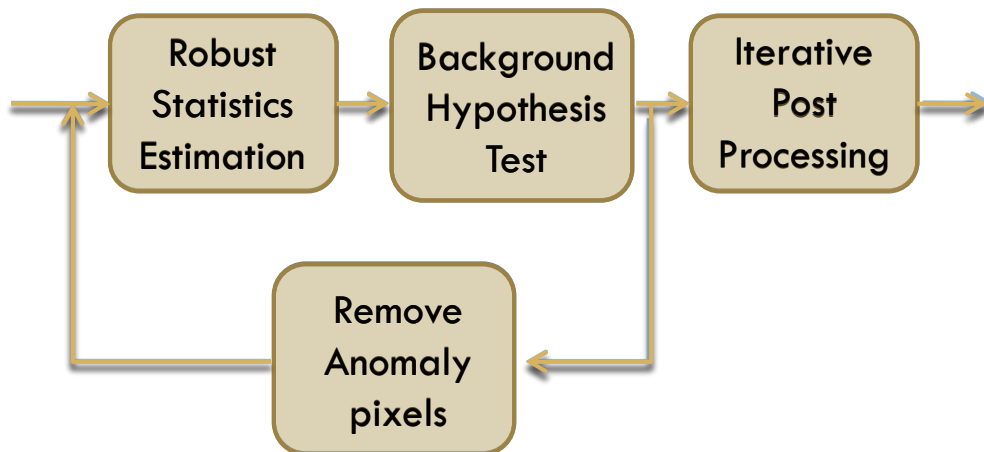
An intermediate set  $B_{tmp}$  exclusively composed of pixels of the dominant background cluster is obtained.

## □ Post Processing:

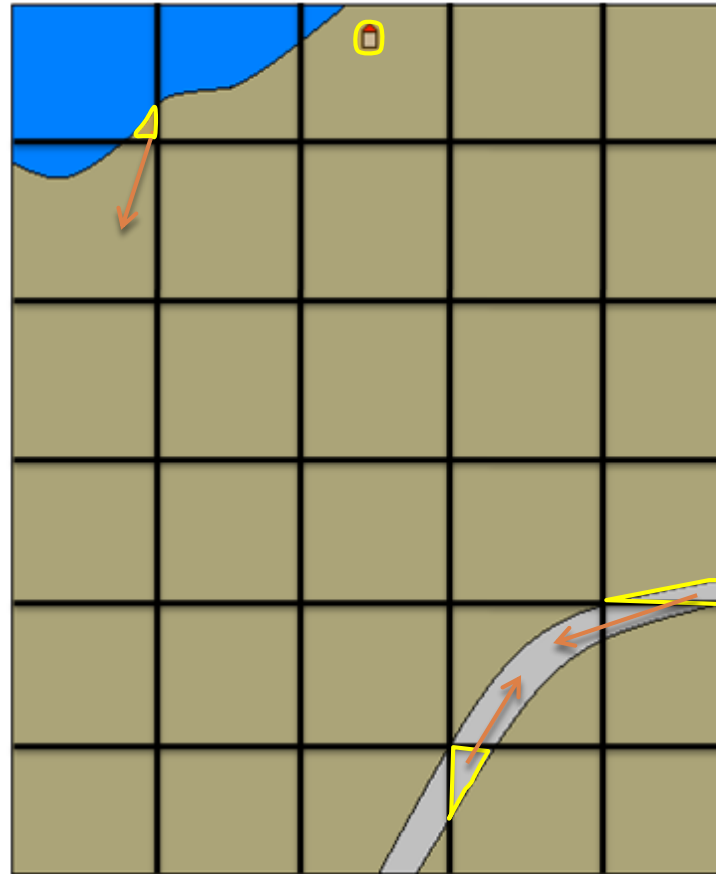
Re-introduce the excluded pixels back into  $B_{tmp}$

## □ Result:

- Clusters statistics estimation
- Pixel classification ( local anomalies)

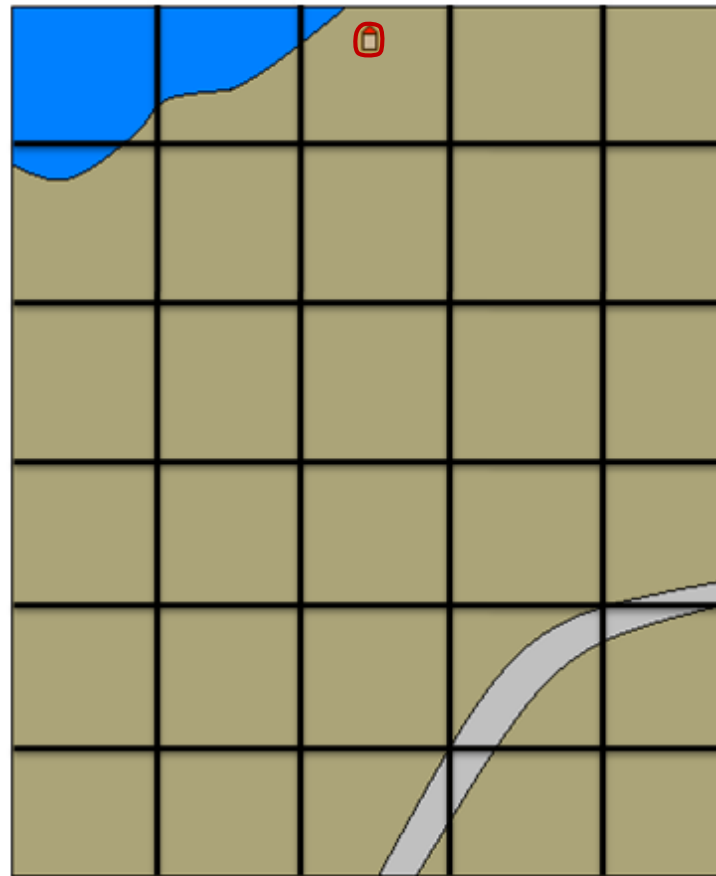


# Global Part



Anomaly classification to  
Local anomaly  
Determine  
back ground cluster in a  
block of a global area

# Global Part



— Global  
Anomalies

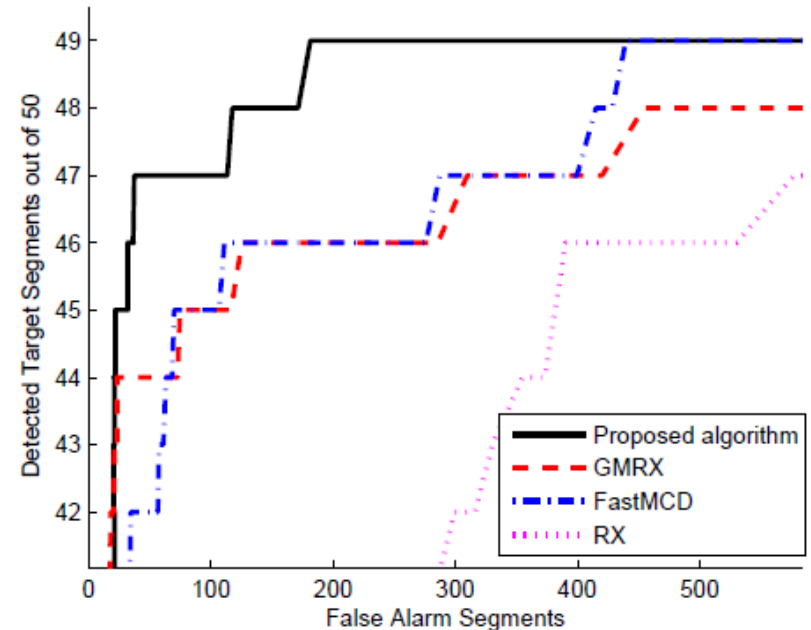
# Experimental Results

## □ Data:

- AISA airborne sensor
- 5 real hyperspectral image cubes (1.2km<sup>2</sup>)
- 65 spectral bands (400-1000nm)
- 50 anomalies (vehicles and small constructions)

## □ Algorithm:

- Local block size 40x40
- Global block size 400x400





# Summary

- ***Combined Local-Global proposed algorithm***
  - Local – Greedy sequential estimation process
  - Global – Filtering using large image area statistics
- ***Advantages of the new approach***
  - Reduces the vast number of degrees of freedom while retaining the ability to be locally adjusted to the background.
  - Outperforms both standard local and global algorithms.