

# Camera Forensics

Jean-Baptiste Boin, Tiffany Jou  
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# Goal

- Classify real vs. fake images
- Localize tampered region in fake images



# Background

- Camera Forensics
  - Exploit periodicity of Color Filter Array
- Unknown sampling lattice + Unknown interpolation
  - Expectation-Maximization (EM)



# Algorithm

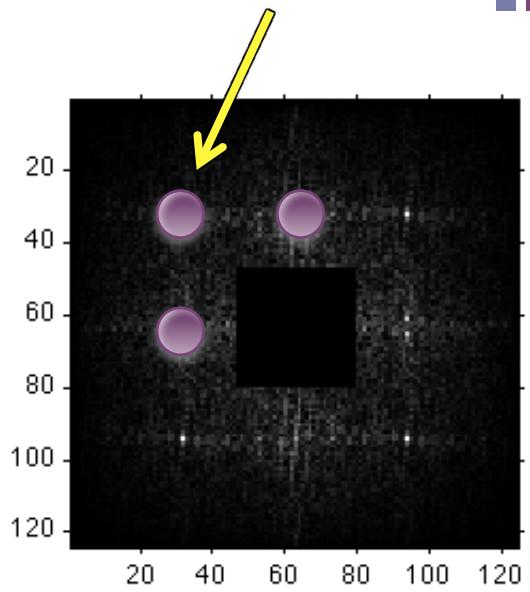
- Sliding blocks: 64x64
  - 32 pixels increments
- EM
- Localization: Clustering
- Classification



# Algorithm: EM

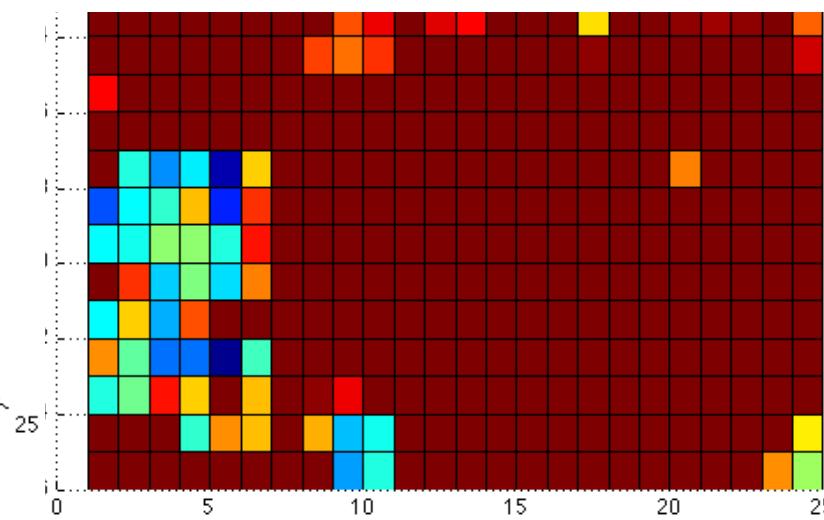
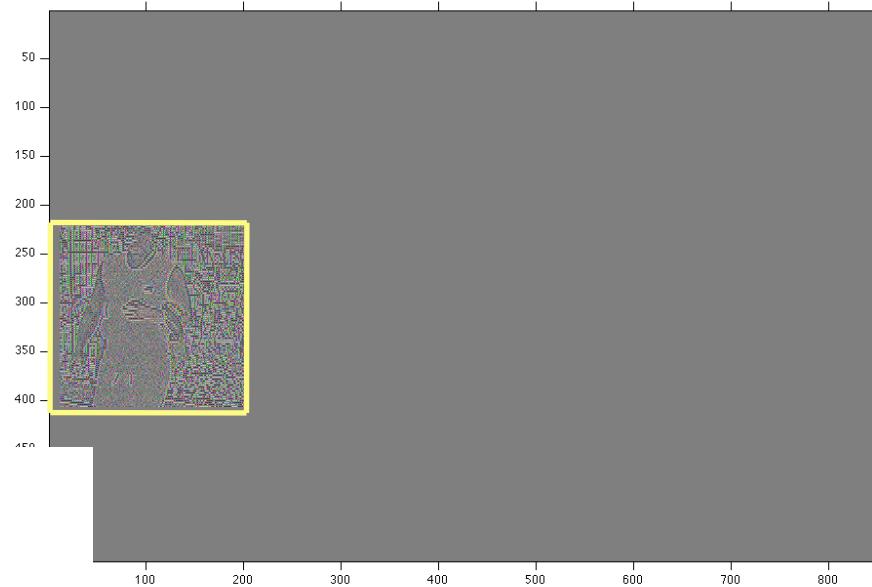
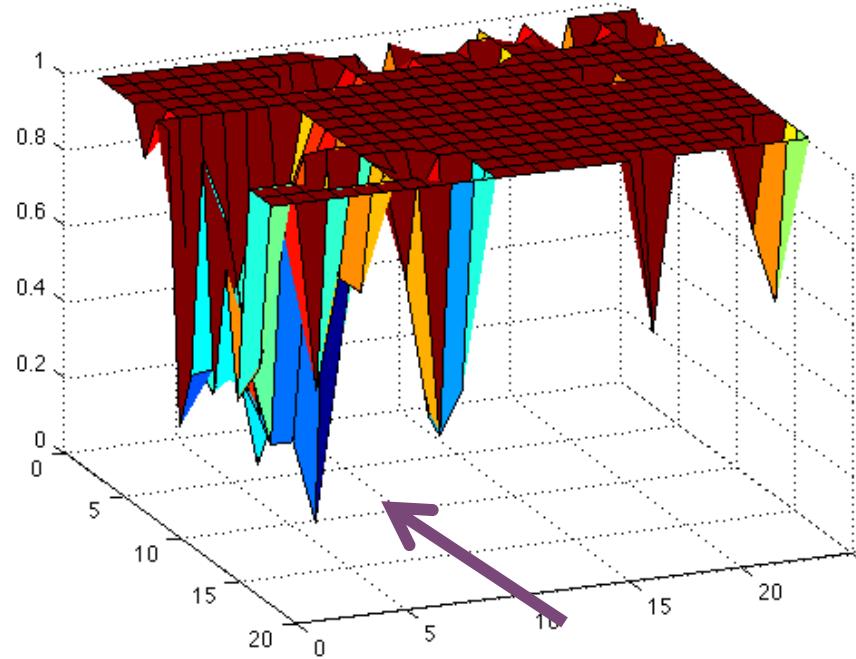
- Probability map: Fourier Transform
- Alphas
  - Describe interpolation

- Assume Bayer pattern
- Normal FFT block: high intensity in at least one of 3 points
- Pick the one with highest intensity
- Normalize this value with max value
- “FFT peak” : represents block
- 3 color channels



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FFT



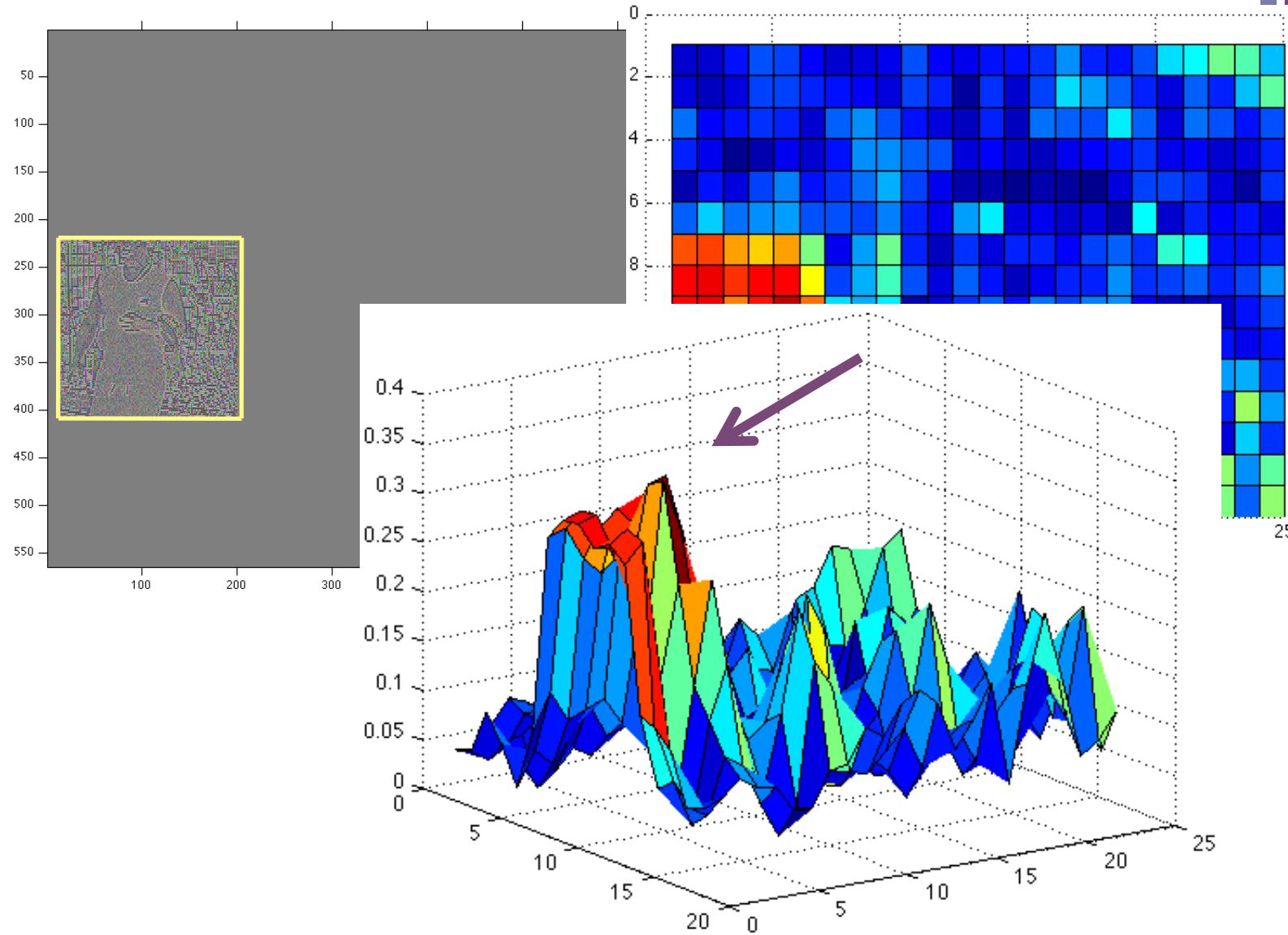


# alphas

- alpha vector for every block
- Find median alpha vector of image
- Calculate the norm between alpha vector of each block with median
- Tampered region= large norm values

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# alphas



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# Algorithm: Clustering

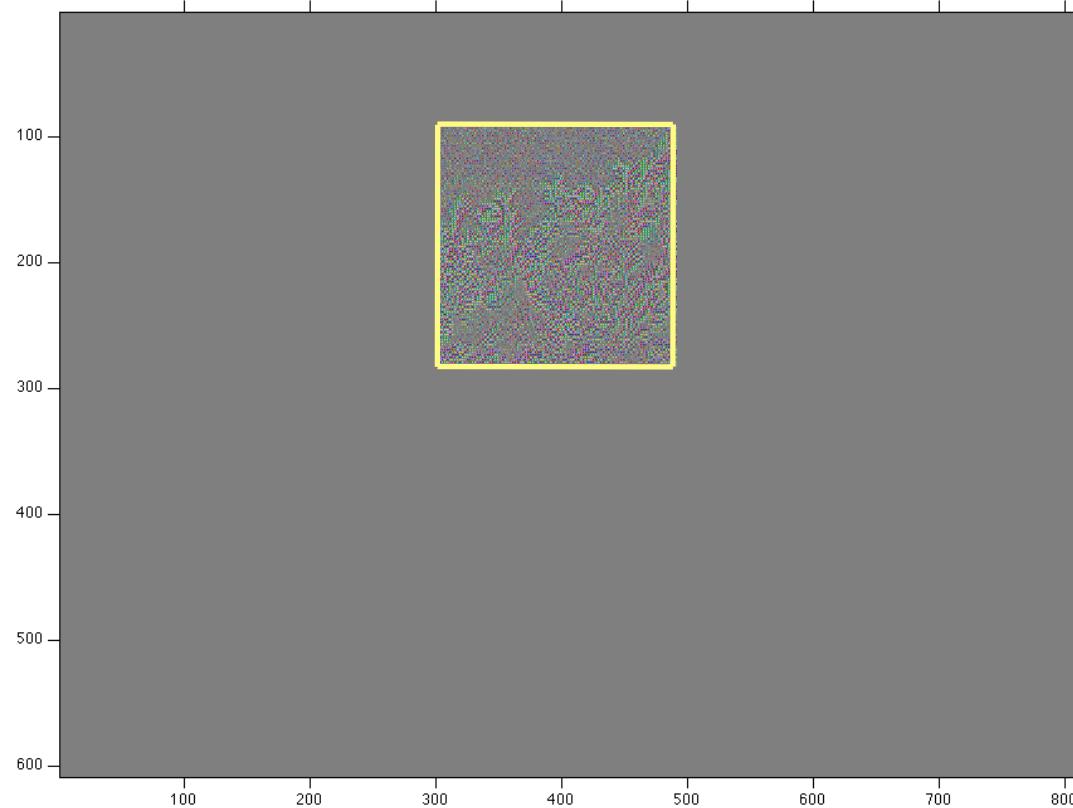
- 27-dim vector per block
- 2-cluster k-means
- Initialization
  - Use min FFT peak block as centroid for one cluster
  - Find other centroid far away
- Force one cluster to be smaller than the other
- Different weights between FFT peaks and alphas

# + Results: Clustering



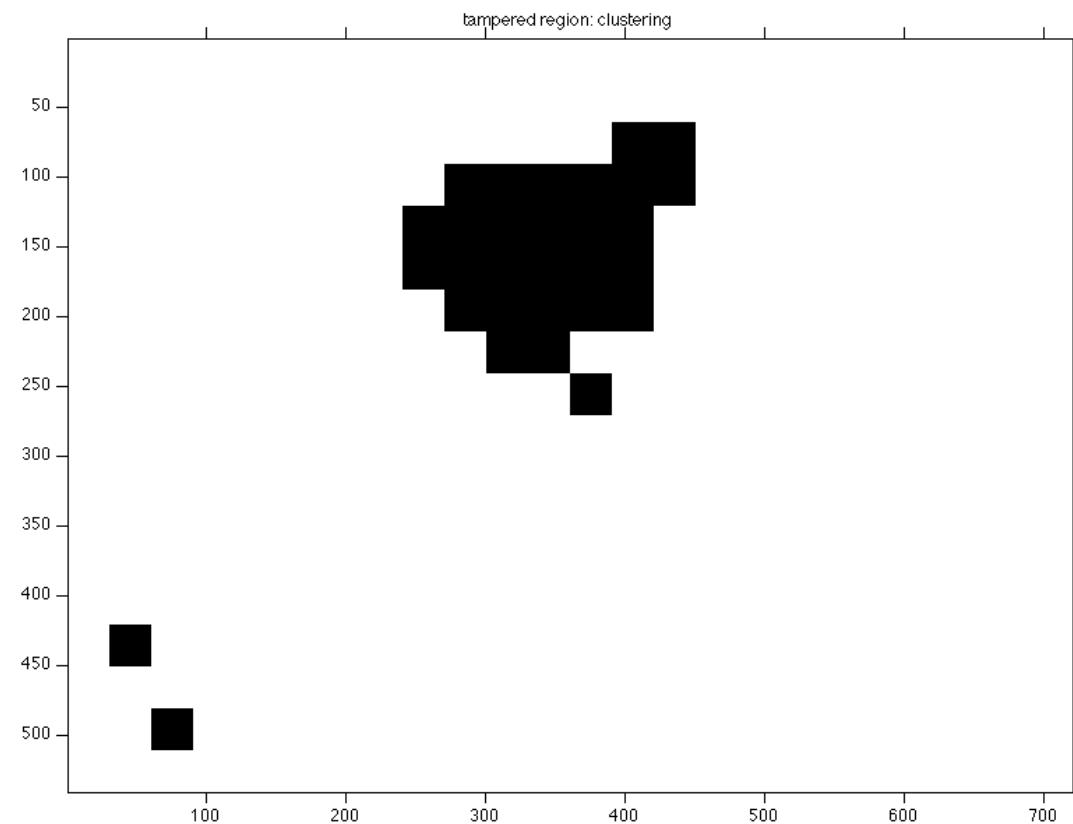


# Results: Clustering



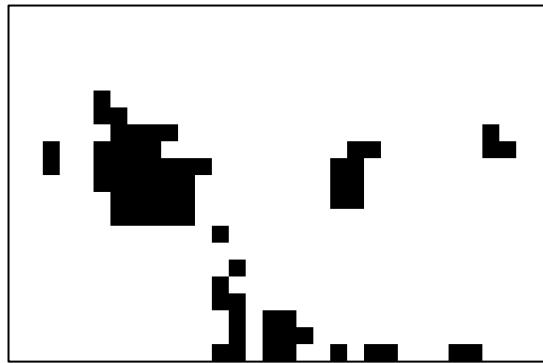


# Results: Clustering

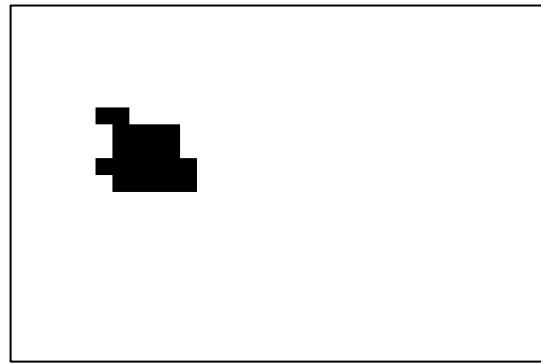




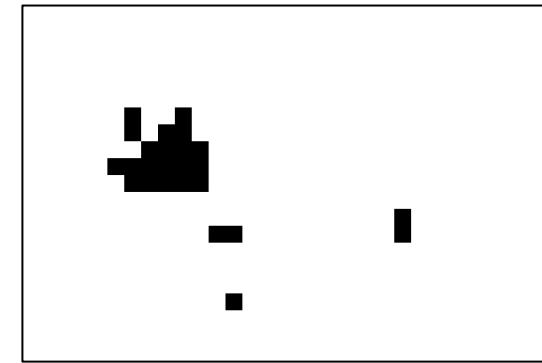
# Weight sequence



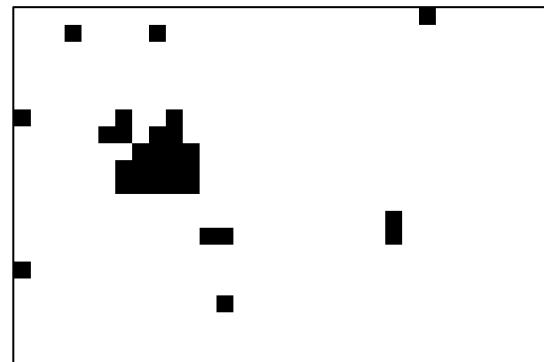
1



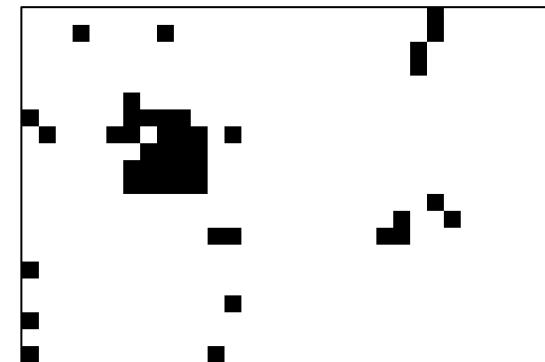
2



3



4



5

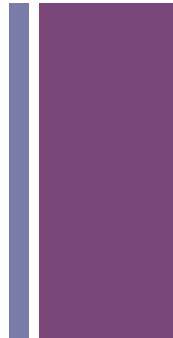


# Assumptions

- Only one tampered zone
- Tampered zone is not too small
  - Block size = 64 x 64
  - Lower bound for detection
- Tampered zone is not too big
  - < 60 % of image size



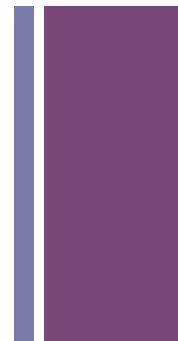
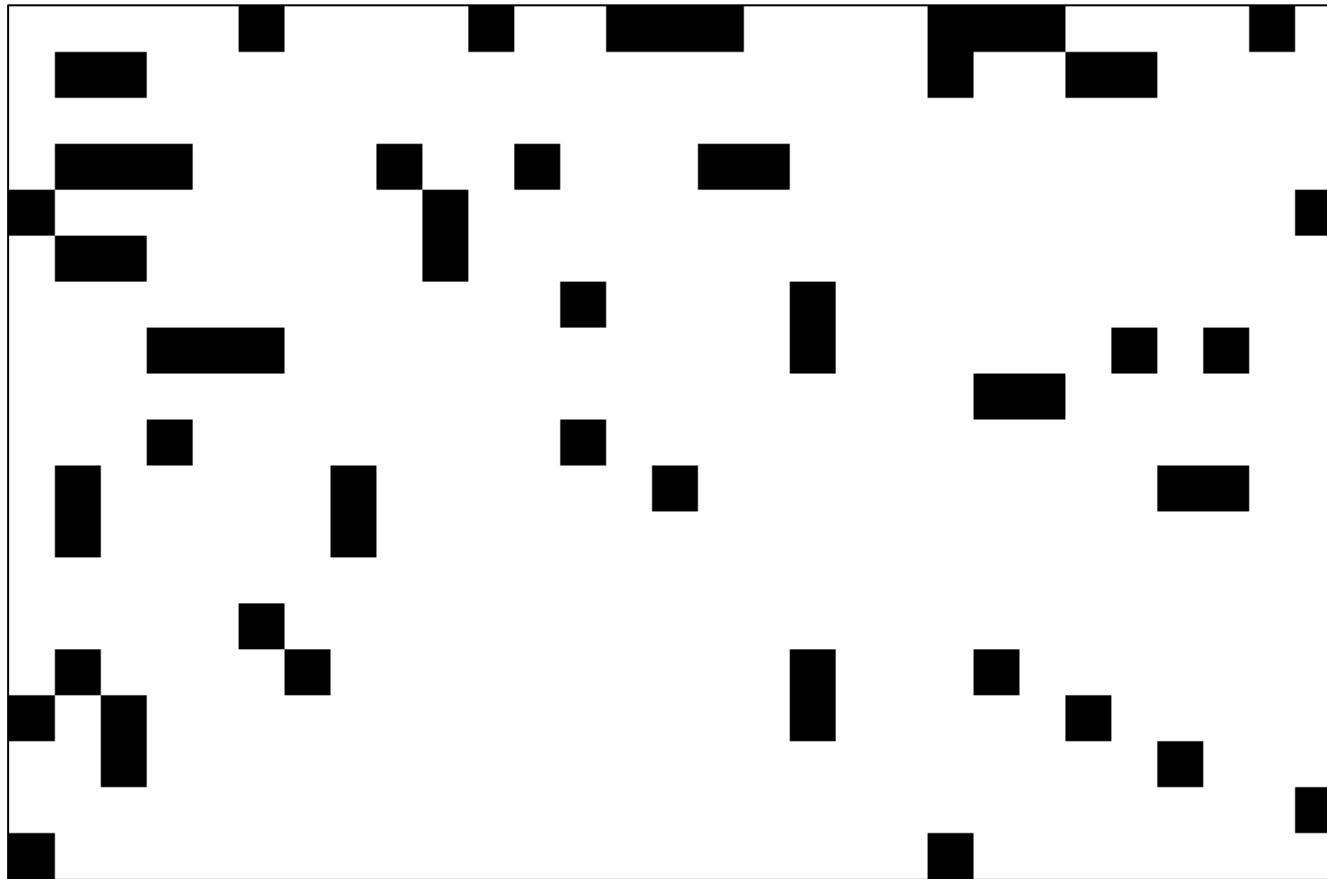
# Algorithm: Classification



- Goal : separate signal from noise
- Properties of the signal
  - One connected component of moderate size
- Properties of the noise
  - Small multiple components or one very big (case of large uniform zones)
- Introduction of parameters to classify our map properly

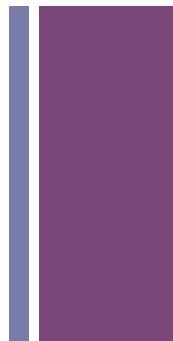
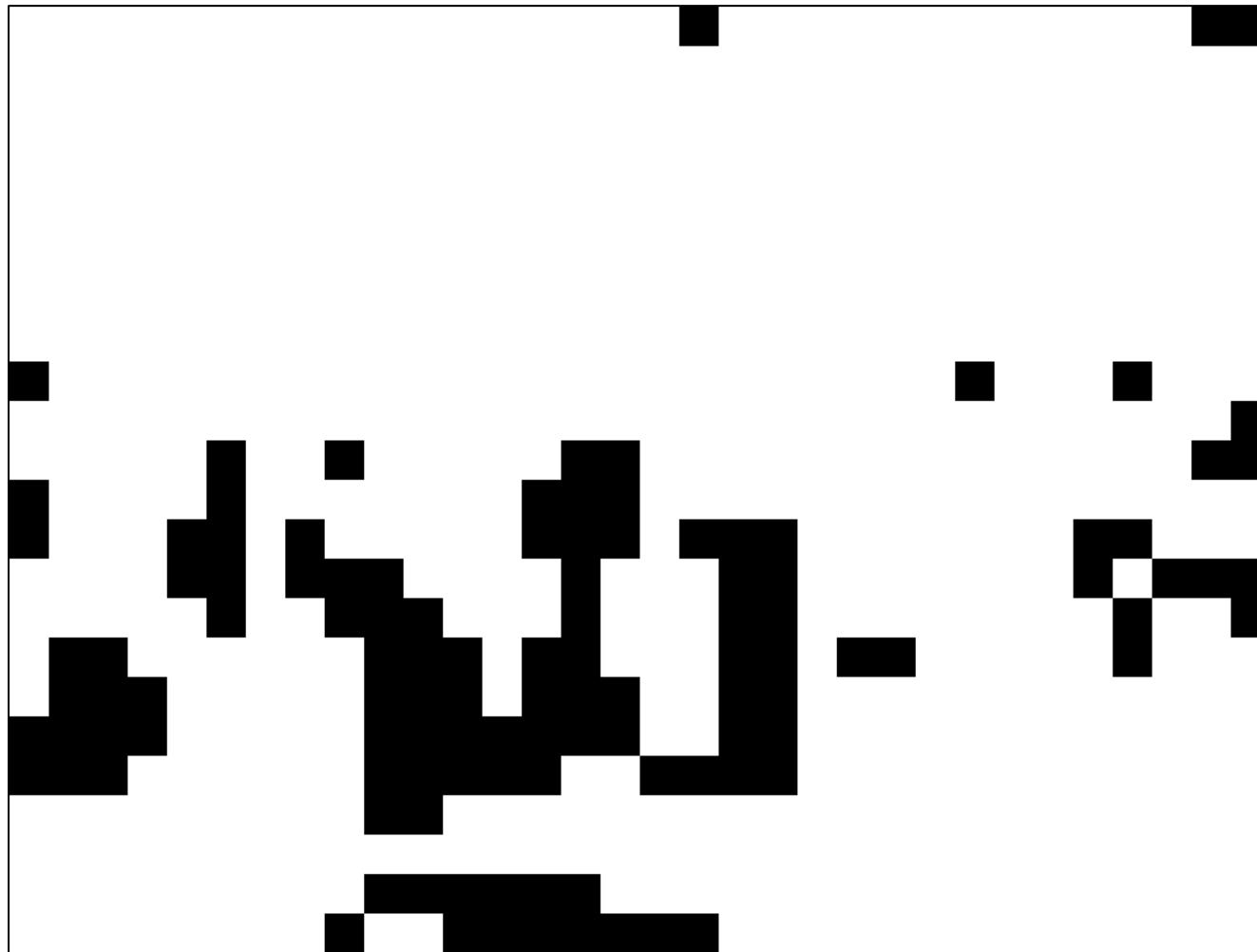


# Results: Real Image 1



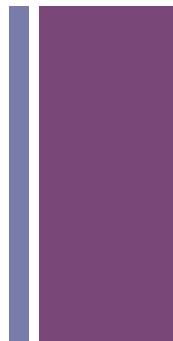


## Results: Real Image 2



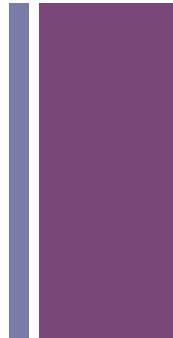


## Results: Real Image 3





# Algorithm: Classification



- Parameters
  - Smallest admissible noise
  - Comparison signal to total noise
  - Maximum size of our signal
- Apply for each of the 5 block maps
- Keep the only connected component that verifies our constraints
- If none of them does, the image is classified it as real

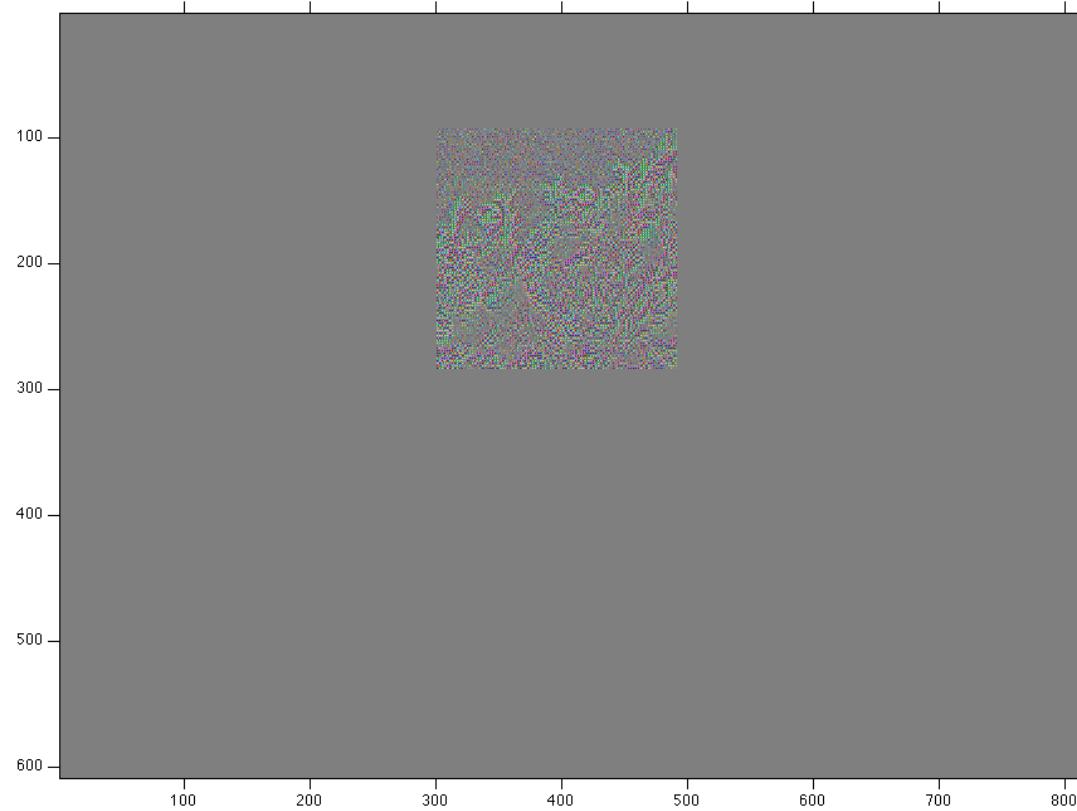


## Results: Image 1

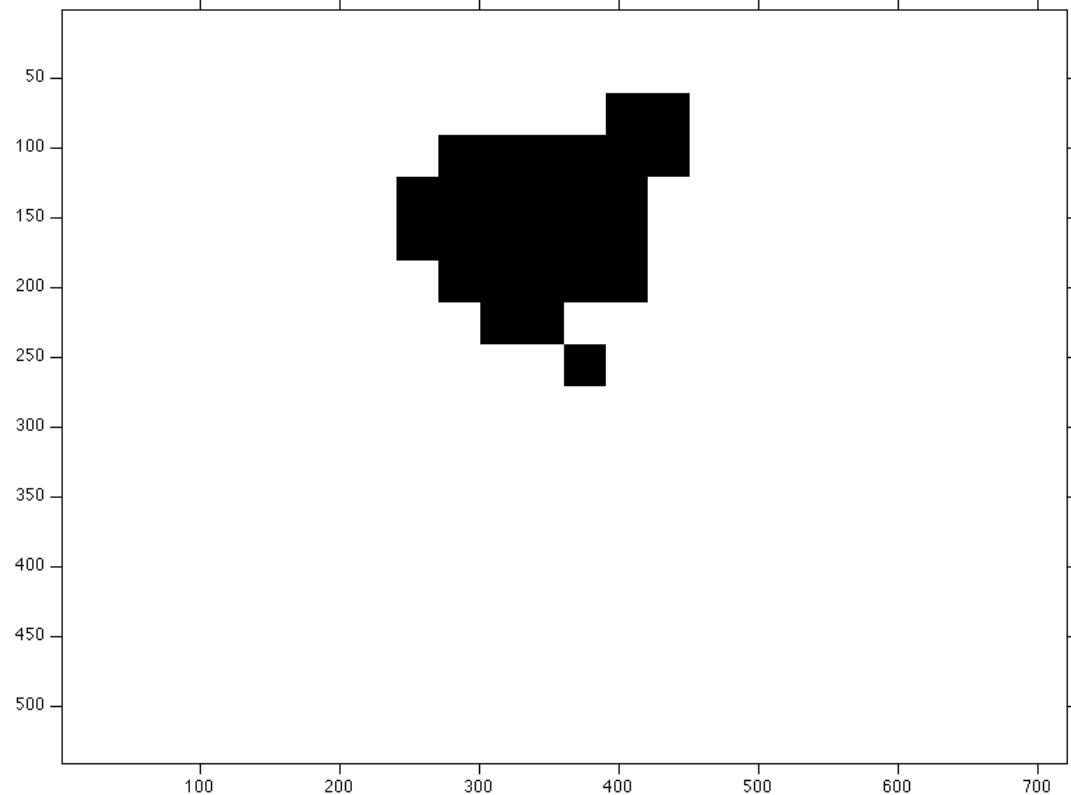




# Results: Image 1



# Results: Image 1



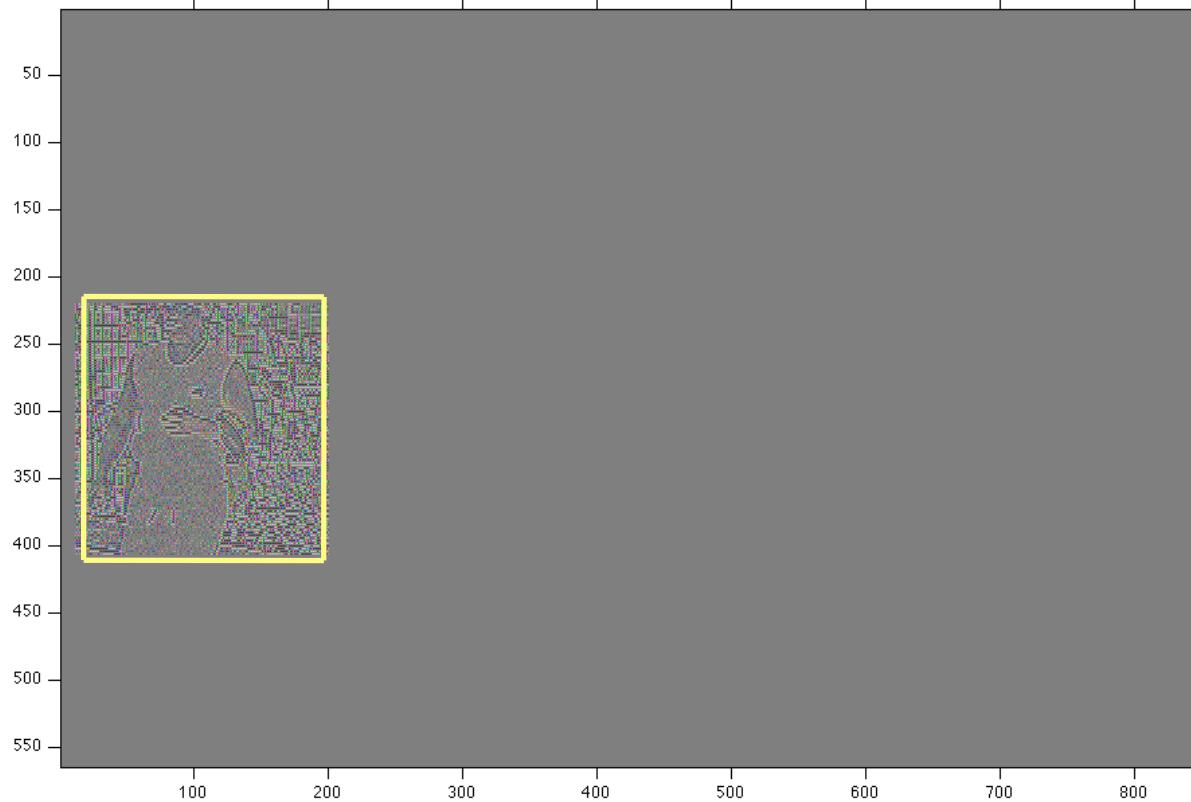


# Results: Image 2



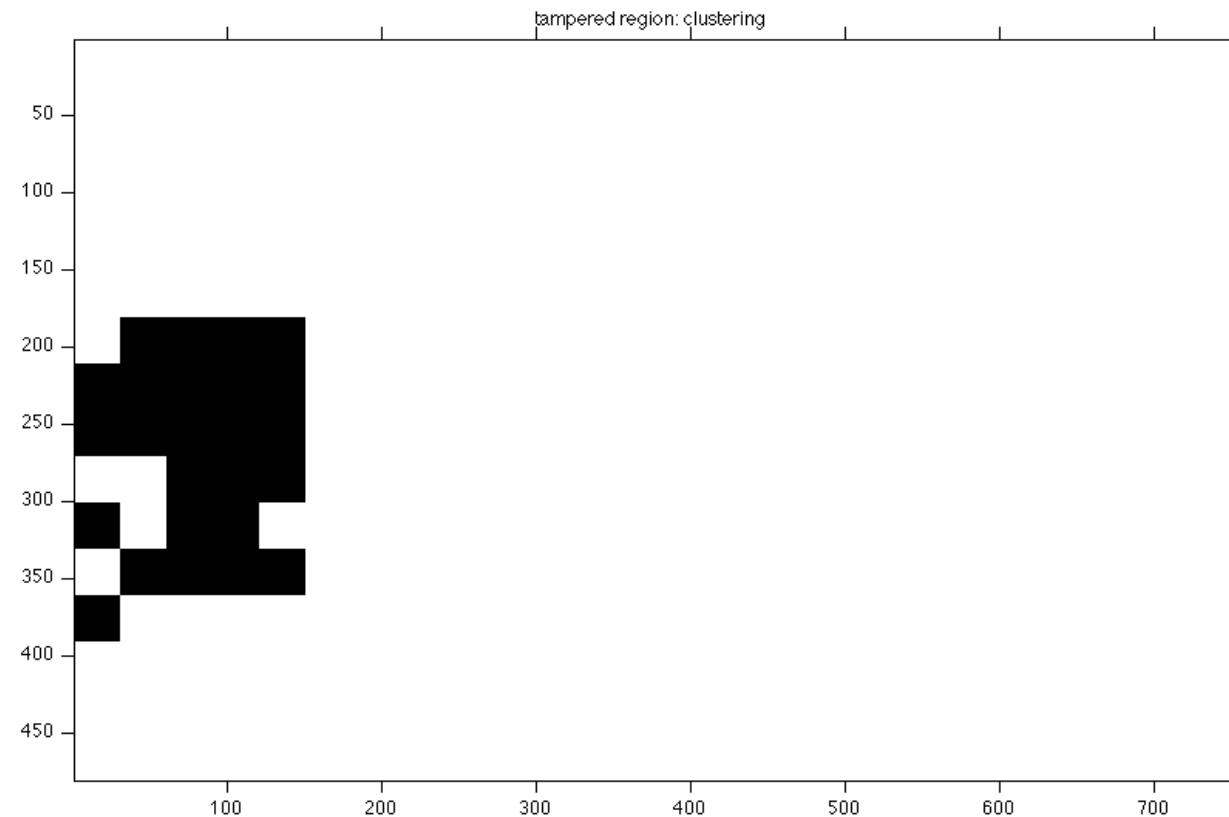


# Results: Image 2





# Results: Image 2



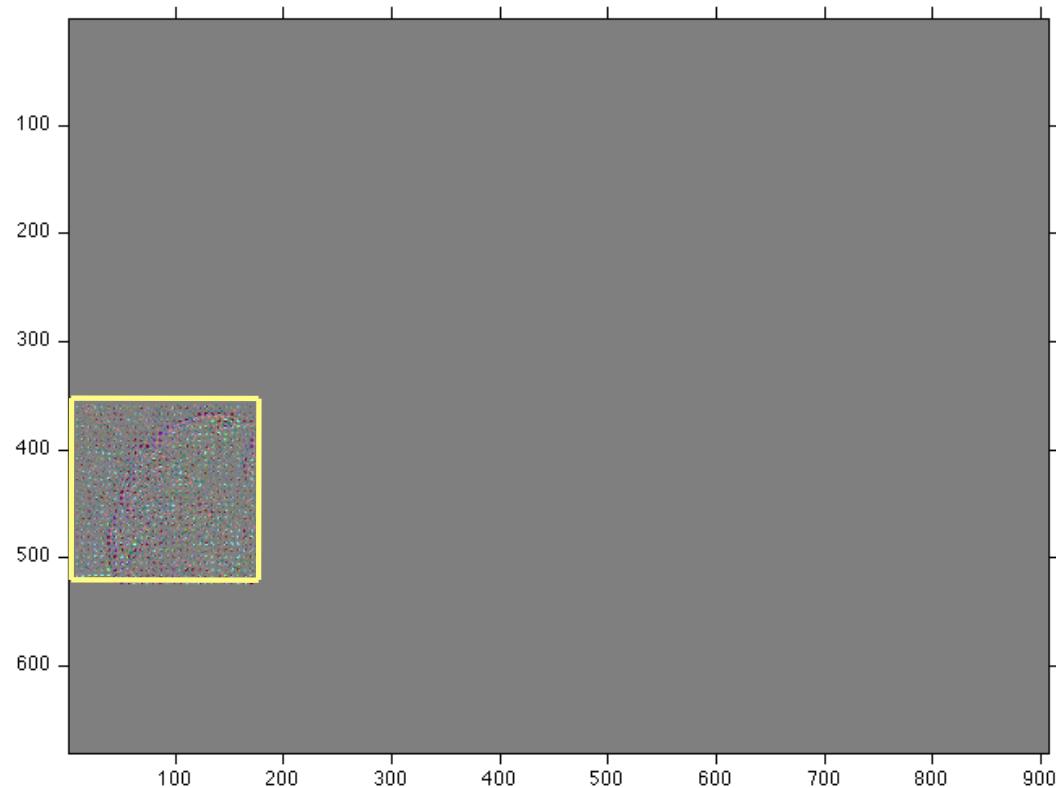


## Results: Image 3



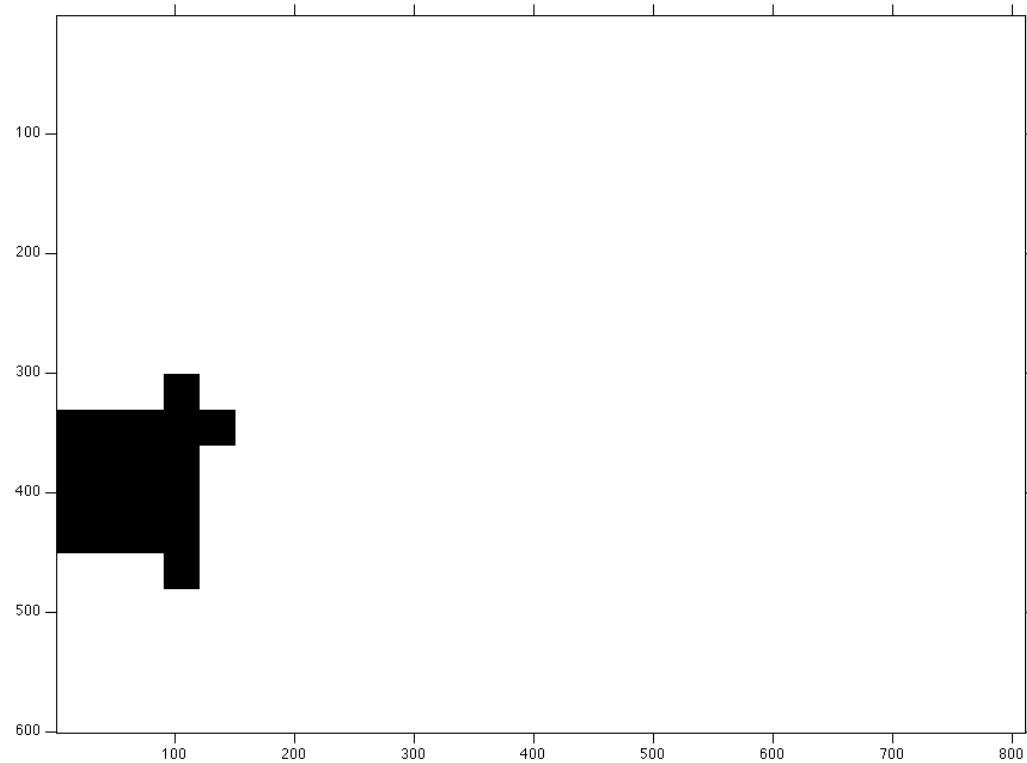


# Results: Image 3



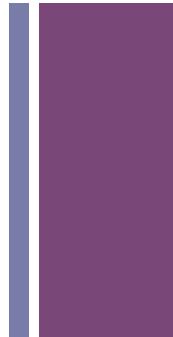


# Results: Image 3





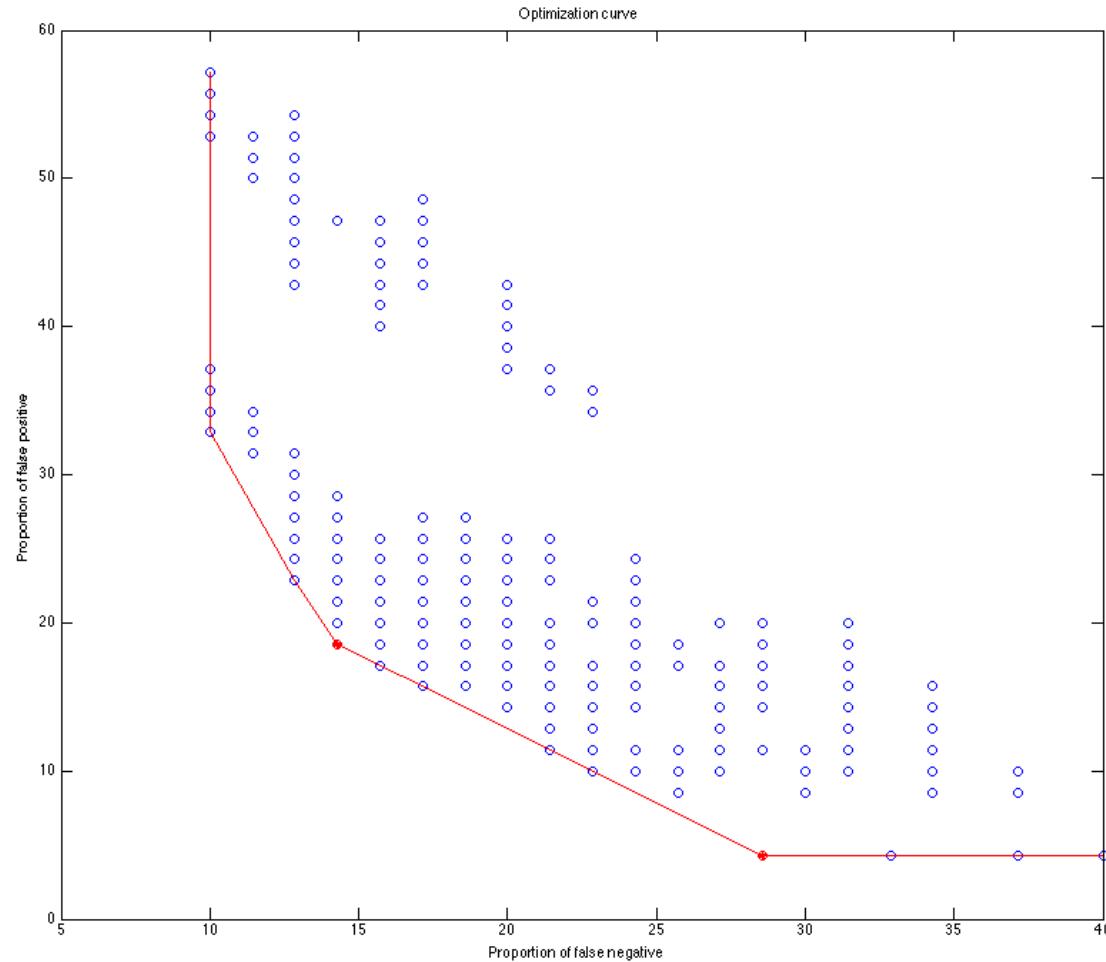
# Results: Classification



- 140 given images (70 tampered, 70 untampered)
- Based on test images, we choose our parameters to achieve the best performances
- Degree of freedom on our sets of parameters depending on our priorities : Pareto curve

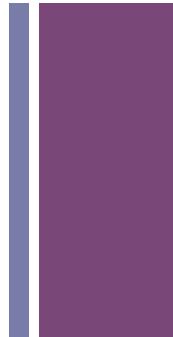


# Results: Classification





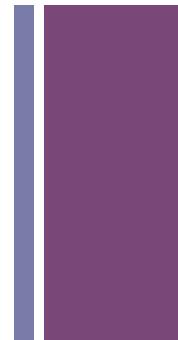
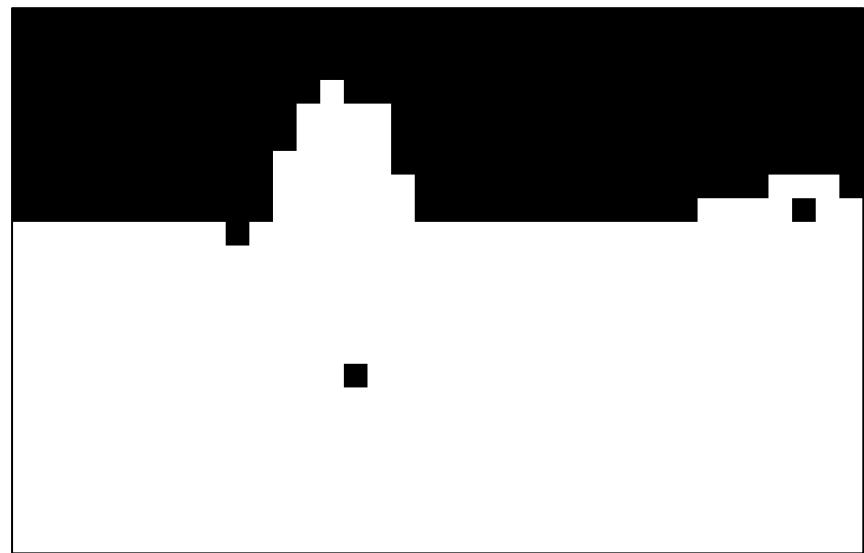
# Analysis - Reasons of failure



- Size of tampered region
- Sliding blocks of 64x64
- Large uniform zone in real images
  - No peaks, constant



# Analysis - Reasons of failure





# Ideas for Improvement

- Increase or decrease sliding block size
- Decrease sliding interval
  - More time
- Detect uniform regions
  - Process differently



# Conclusion

- Main goal: detect local tampering in image
- For images that follow assumptions, classification & localization works well
- Classify between cameras
  - Use alpha values

