Comparison of Earthquake Damage Evaluation using Change Detection and Thematic Classification



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## **Damage Detection Analysis**



- Optical satellite images can provide critical information regarding earthquake damage
- Methods available to identify damage
  - Change detection (requires pre- and postearthquake images)
  - Thematic classification (requires only postearthquake image)
- Application of these methods to 2003 Bam, Iran earthquake
- Comparison of change detection and thematic classification results

## 2003 Bam, Iran Earthquake



### 2003 December 26 M<sub>w</sub> 6.6



 Pre- and postearthquake Quickbird images

30 Sept 2003 4 Jan 2004



## **Change Detection**



- Requires pre- and post-earthquake images
- Co-registered pre- and post- earthquake images
- Use image-to-image correlation

$$r = \frac{n\sum PV_a PV_b - \left(\sum PV_a\right)\left(\sum PV_b\right)}{\sqrt{n\sum PV_a^2 - \left(\sum PV_a\right)^2} \cdot \sqrt{n\sum PV_b^2 - \left(\sum PV_b\right)^2}}$$

 $PV_a$  = pixel value in pre-earthquake image  $PV_b$  = pixel value in post-earthquake image n = number of pixels in correlation window

### 15 by 15 pixel (9 m) window used





- Earthquake damage shows changes in texture
- Use texture measures based on the graylevel co-ocurrence matrix (CM)
  - Homogeneity, dissimilarity, contrast
  - Second moment, entropy
  - Mean, variance, correlation
- Considered texture over 31 by 31 pixel window, 15 pixel horizontal shift

## **Change in Texture**



### Heavily damaged area



## **Results of Change Detection**

### Using correlation coefficient and VAR31 feature



- Red damage
- Threshold > 0.5
- Vegetation and shadow mask

## **Thematic Classification**



- Requires only post-earthquake image
- A subset of data associated with "damaged" and "undamaged" areas is identified for training the algorithm.
- Apply Bayesian pair-wise feature selection algorithm in conjunction with a maximum likelihood classifier

## **Results of Thematic Classification**

Using maximum-likelihood classification and 14 spectral and textural features selected by feature selection



- Red damage
- Green vegetation
- Blue buildings
- White open areas
- Cyan roads





- Defined as percentage of damaged pixels within a 60 m by 60 m (100 pixels by 100 pixels) window
- Only consider pixels that are urban area
  DIDN'T WE CHANGE THIS??
- Threshold for earthquake damage
  DI > 40%



## **Damage Intensity-CD**







## **Damage Intensity-ML**







## **Damage Intensity**







ML	CD
38% of image DI3-DI5	24% of image DI3-DI5
DI3 – 5.16km <sup>2</sup>	DI3 – 2.98km <sup>2</sup>
DI4– 2.63km <sup>2</sup>	DI4 – 1.79km <sup>2</sup>
DI5– 0.44km <sup>2</sup>	DI5 – 0.48km <sup>2</sup>



## **Field Damage Survey**





# **Comparison with Field Survey**

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## Zone 1- Undamaged area



### **Pre-earthquake**



### Post-earthquake





## **Comparison of Results**



### **ML Classification**



### **Change Detection**





## Zone 2- Damaged area



### **Pre-earthquake**

### **Post-earthquake**







## **Comparison of Results**



### **ML Classification**

### **Change Detection**





## **Conclusions**



- Thematic classification identified more damage than change detection when considering the entire city
- Thematic classification is not always successful in distinguishing between different levels of severe damage



## **Conclusions**



- Change detection distinguished better different levels of severe damage
- Change detection identified some nonearthquake change that resulted in an overestimation of damage in isolated areas
- Future work
  - Developing multi-resolution techniques
  - Advanced textural features (e.g., wavelets)
  - Hierarchical classification