



Structural Change Detection by Direct 3D Model Comparison

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Introduction

- Detecting **structural changes in a 3D scene** over time is required in several application scenarios
 - Archaeological campaign documentation
 - Hydro-geological monitoring
 - Building construction management
 - Tracking urban architectural changes



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• Change detection methods

- Human visual inspection
- Comparison of **two 2D images** or sequences (Radke et al. 2005) requires fixed viewpoint and illumination compensation
- Compatibility check between **an image sequence and a 3D model** (Taneja et al. 2011) needs a subsequent classification step into structural (e.g. buildings) vs non structural (e.g. pedestrians, cars) changes



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• Our solution

- Fully automatic
- Based on the direct **comparison of 3D models** of a time-changing environment obtained by image sequences acquired at two different times
- Relatively insensitive with respect to illumination and viewpoint changes
- Detects only changes related to structural (fixed) elements in the scene (SfM)





 Input: image sequences of the same scene acquired at two different times

 Pre-processing: construction of sparse 3D models with off-the-shelf software (VisualSfM, COLMAP)

• Step 1: photometric rigid registration to align the 3D reconstructions

Step 2: structural change detection by comparison of the 3D models







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• Step 2: **structural change detection ("heat map")** by 3D model comparison



Photometric registration

- After reconstruction we obtain for each sequence:
 - A 3D model, expressed in an arbitrary reference system and scale
 - A set of 2D feature descriptors and inter-frame tracks
 - A list of 2D/3D correspondences
- Registration is obtained by **feature matching** between the images of the two sequences, and then...





Photometric registration

- ...2D matches are promoted to 3D, both in T_0 and T_1
- The 3D point map between T₀ and T₁ can finally be estimated from 3D correspondences





Photometric registration

• With initial 3D correspondences given, the transformation τ that maps T₁ onto T₀ is obtained by **minimizing the symmetric re-projection error** in a RANSAC framework:

$$argmin_{\tau} \sum_{\mathbf{x} \in T_0, \mathbf{X}' \in T_1} |\mathbf{x} - \tau(\mathbf{X}')|^2 + \sum_{\mathbf{x}' \in T_1, \mathbf{X} \in T_0} |\mathbf{x}' - \tau^{-1}(\mathbf{X})|^2$$

- To refine τ , the **Iterative Closest Point** (ICP) algorithm is used:
 - Sparse 3D models are **densified** with a region growing method (Furukawa & Ponce 2010)
 - T_1 is mapped with τ into the reference system of T_0
 - ICP is run to obtain a finer registration



- Structural changes are searched using a shifting 3D volume V that slides within a 3D box B including the dense 3D reconstructions
- V is shifted with a volume overlap of 3/4
- Corresponding volumes V on T₀ and T₁ are compared with three criteria:
 - Quantity
 - Orientation
 - Occupancy



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Quantity criterion

- Counts the number of 3D points included in V both in T₀ and T₁
- If a difference greater than α is found, the criterion is satisfied

• α is set to the average number of points within V, considering both T₀ and T₁







Orientation criterion

- For each 3D point a normal vector is computed and parametrized by its angles
- For each position of V, the distribution of the angles of the normal vectors is stored into a 2D histogram
- **Histograms are compared**: if their distance is greater than β , the criterion is satisfied







Occupancy criterion

- The volume V is split into 27 sub-volumes v_i
- Each v_i is labeled as *active* if it includes at least one 3D point, otherwise it is labeled as *inactive*
- If more than γ sub-volumes have different labels between TO and T1, the criterion is satisfied





Majority vote

- If two or more criteria are satisfied,
 a token is given to all the 3D points in V
- Once V has covered the whole B, the number of tokens received by a 3D point can be considered as a change score



3D heat-map of change







Experimental results

- Change detection was tested using two pairs of image sequences acquired in our lab:
 - Test 1 Sequence A, 22 images. Sequence B, 30 images
 - Test 2 Sequence C, 22 images. Sequence D, 18 images

• Images were obtained using a (non calibrated) smartphone, with a resolution of 640x480.







Experimental results

 Test 1 – Object removal: the statue and the jar have been removed in the second sequence





• Test 2: Object insertion / displacement: insert two cylindrical cans, lay down the statue, change position for the jar, the rocks, and the bricks









Experimental results: qualitative



• Test 2: Object insertion / displacement







Experimental results: quantitative

• Test 1: Object removal









• Test 2: Object insertion / displacement



Input

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Detection



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Experimental results: quantitative

• Test 1: Object removal









• Test 2: Object insertion / displacement



Input with Superpixel Segmentation

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Ground truth



Detection







Conclusions and Future Work

- We proposed a vision based method to detect structural changes by comparing 3D reconstructions
- The method is based on 3D model registration, followed by change detection based on three distinct criteria
- Preliminary tests on laboratory scenes show good results, with AUCs greater than 0.90

- Future work will address:
 - Replacing photometric registration (slow) with direct 3D registration
 - Automatic selection of parameters based on scene scale and image resolution