

Occlusion Handling for the Integration of Virtual Objects into Video

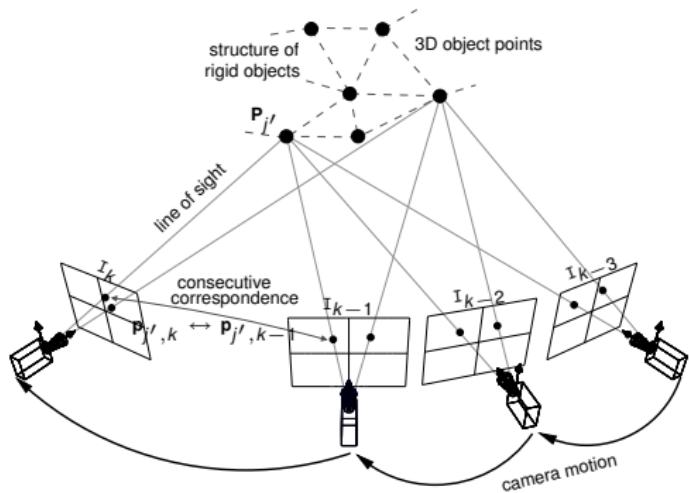
Kai Cordes, Björn Scheuermann,
Bodo Rosenhahn, Jörn Ostermann

Institut für Informationsverarbeitung
Leibniz Universität Hannover

VISAPP 2012 - International Conference on
Computer Vision Theory and Applications



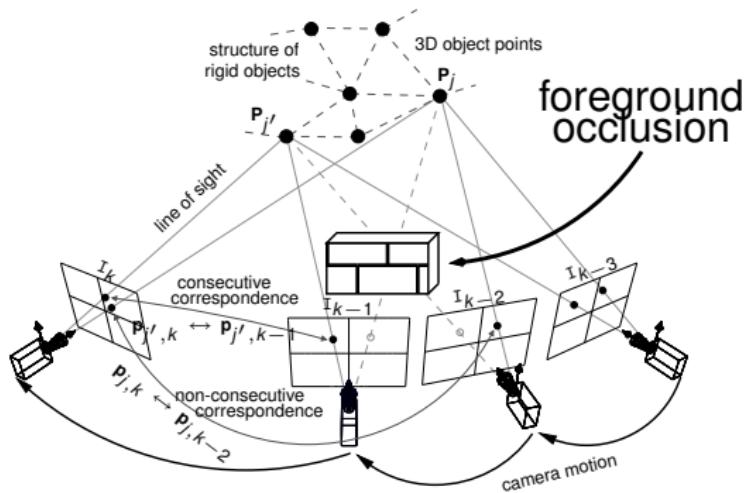
Structure and Motion Recovery (SAM)



- ▶ Feature detection
- ▶ Correspondence analysis
- ▶ Outlier elimination
- ▶ Bundle adjustment



Structure and Motion Recovery (SAM)



- ▶ Occlusion
- ⇒ Trajectories lost
- ⇒ Error-prone object points
- ⇒ Drift



Integration of Virtual Objects into Video

Camera Motion Estimation

- ▶ High-Accuracy needed: Reprojection Errors < 0.25 pel ^a
- ▶ Usage of non-consecutive correspondences ^b

^aHillman et al., ICIP 2010

^bCordes et al., ISVC 2011

Occlusion of Foreground Objects

- ▶ Objects are integrated by composition
- ▶ Problem: augmented objects should be occluded



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Interactive Segmentation:

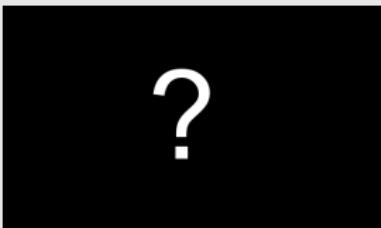
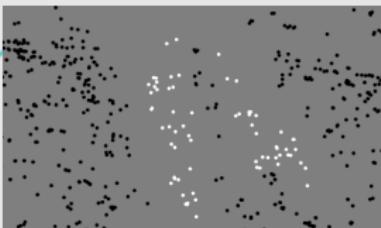


Boykov et al., ICCV 2001

Objectives

- Minimize Energy: $E(f) = \sum_{p \in \mathcal{I}} D_p(f_p) + \sum_{(p,q) \in \mathcal{N}} V_{p,q}(f_p, f_q)$
- ▶ User Input:
 - ▶ Mark foreground and background with strokes
 - Initialization of minimum cut / maximum flow

Automatic Segmentation:



Proposed, VISAPP 2012

Objectives

- ▶ Automatic Foreground Segmentation
 - Generate images with foreground/background knowledge
 - Use segmentation for compositing
- ▶ Extract occlusion Information
 - Establish non-consecutive correspondences (KLT + SIFT)
 - Find occluded object points and their locations in the images

Feature Tracking in SAM Recovery

Automatic Foreground Segmentation

Experimental Results

Conclusion

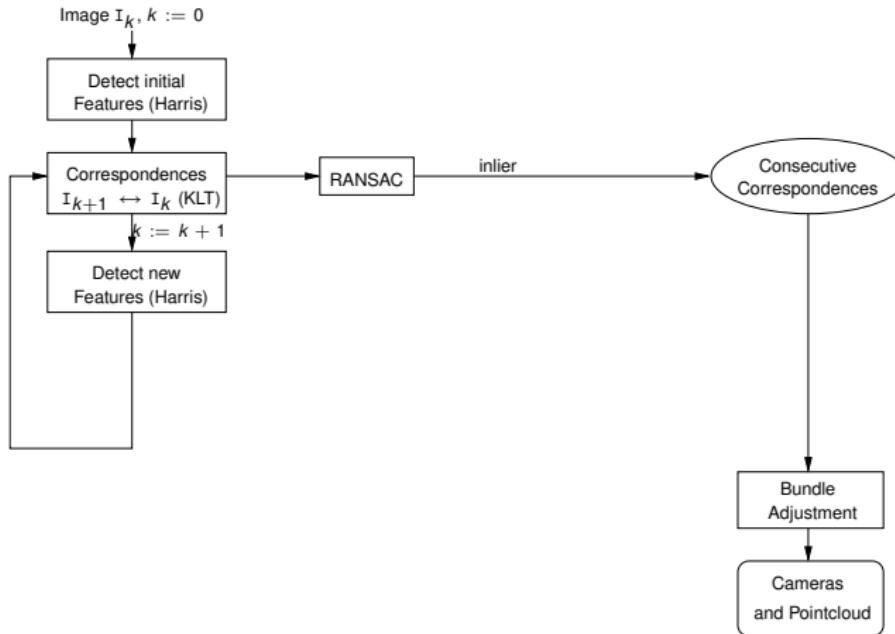
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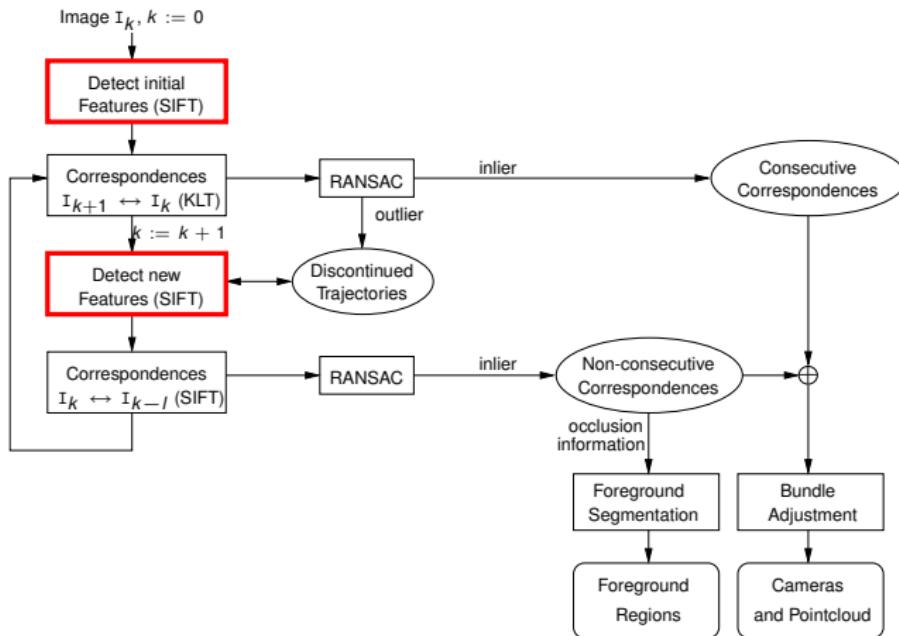
Workflow



SAM pipeline

- ▶ Standard SAM
- ▶ SIFT detection
- ▶ KLT tracking (consecutive)
- ▶ SIFT matching (non-consec.)
- ▶ Occlusion information
- ▶ Segmentation

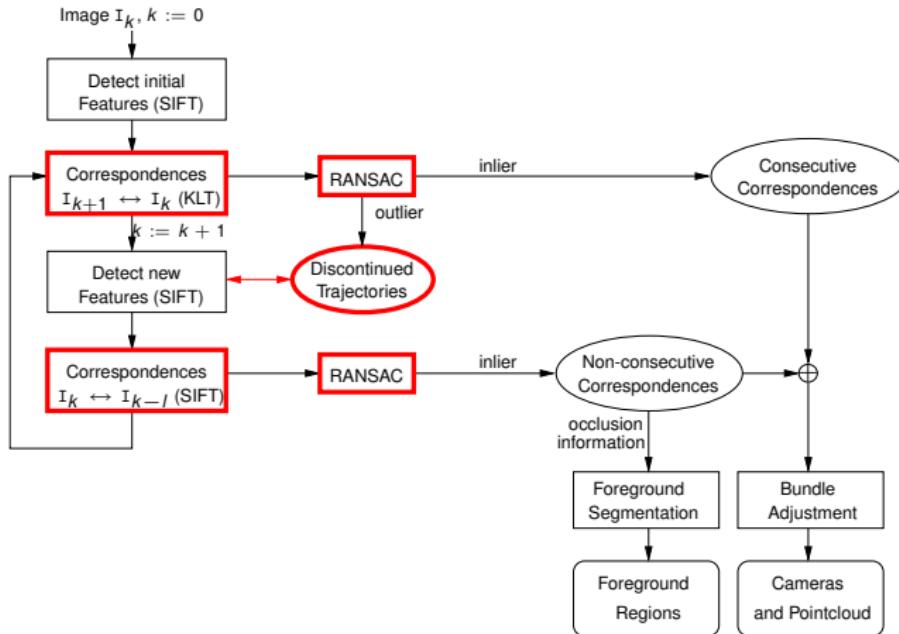
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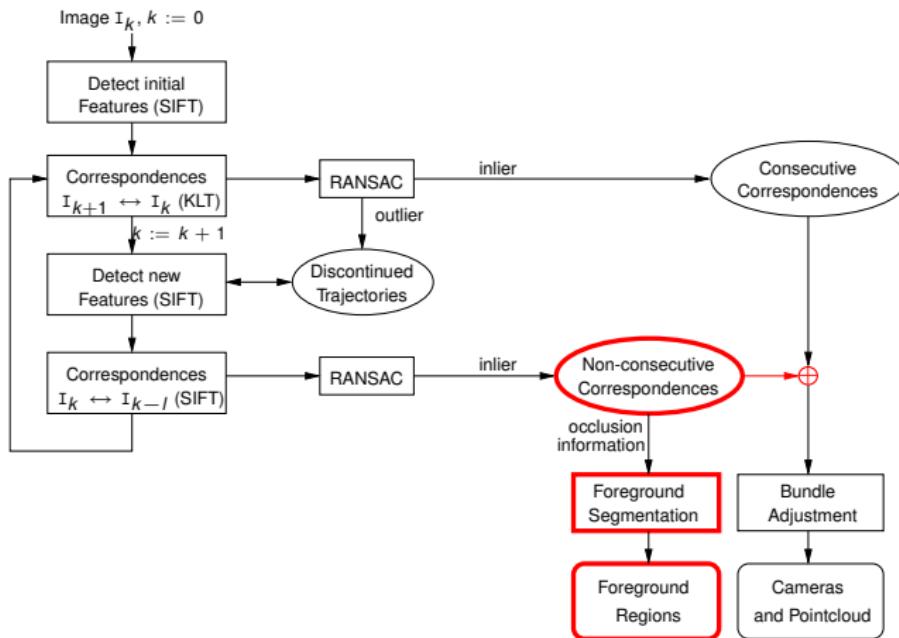
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Occlusion Information

Non-consecutive correspondences

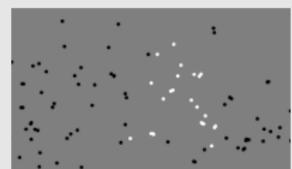
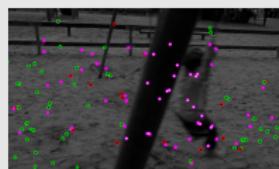
- $\mathbf{p}_{j,k} \leftrightarrow \mathbf{p}_{j,k-l-1}$ induces trajectory:

$$\mathbf{t}_j^* = (\mathbf{p}_{j,k}^{\text{visible}}, \mathbf{p}_{j,k-1}^{\text{occluded}}, \dots, \mathbf{p}_{j,k-l}^{\text{occluded}}, \mathbf{p}_{j,k-l-1}^{\text{visible}}, \mathbf{p}_{j,k-l-2}^{\text{visible}}, \dots)$$

→ $\mathbf{p}_{j,k-1}^{\text{occluded}} = \mathbf{A}_{k-1}\mathbf{P}_j^*, \dots, \mathbf{A}_{k-l}\mathbf{P}_j^*$ (Reprojection)

→ Object point \mathbf{P}_j^* is

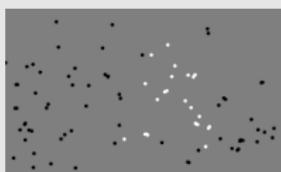
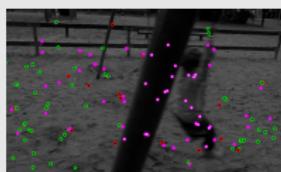
- visible in the current image I_k
- occluded in the images I_{k-1}, \dots, I_{k-l} .



Occlusion Information

Similarity Constraint to Verify Occlusion

- ▶ Non-Consecutive Correspondences induced by:
 1. **Occlusion**
 2. Scene content temporarily leaves field of view
 3. Repeated Texture
 4. Noise, e.g. motion blur
- ▶ Use neighborhood color histogram of t_j^* - features



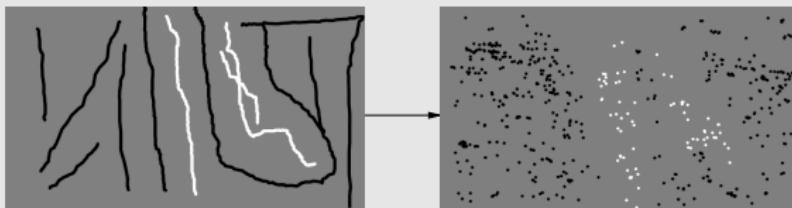
Automatic Segmentation

Use Graph-Cut based approach

- ▶ User interaction: manually drawn strokes
- ▶ Overhead: less than a second for 3 MPixels

Scheuermann et al., EMMCVPR 2011

Strokes derived automatically



Compositing

Combine Input and Augmented Sequence



+



Compositing

Combine Input and Augmented Sequence



+



Outline

Feature Tracking in SAM Recovery

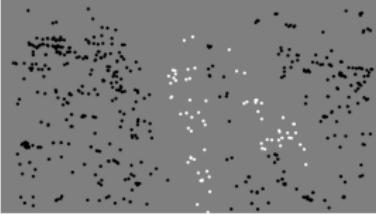
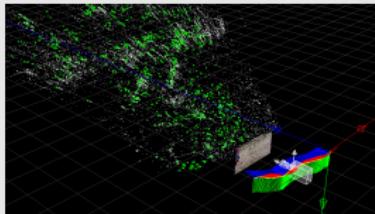
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Experimental Results

Playground Sequence



Video at:

<http://www.youtube.com/watch?v=RuSsIMfM1CE>

Extended Structure and Motion Recovery

- ▶ Combination of KLT and SIFT
- ▶ Extract Occlusion points
- ⇒ Automatic foreground segmentation

Realistic results

- ⇒ Highly accurate reconstruction
- ⇒ Integration and occlusion of virtual objects convincing

Future Work:

- ▶ Use temporal constraint for segmentation
- ▶ Use complementary feature types to maximize distribution

<http://www.tnt.uni-hannover.de/staff/cordes/>

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Limitations

- ▶ Errors resulting from a misleading segmentation
- ▶ Left: no occlusion information in the fence, but visually similar
- ▶ Right: point is correct, but boundary smeared (motion blur)
- Use temporal constraints for segmentation