

# Color Segmentation Based Depth Filtering

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# Overview

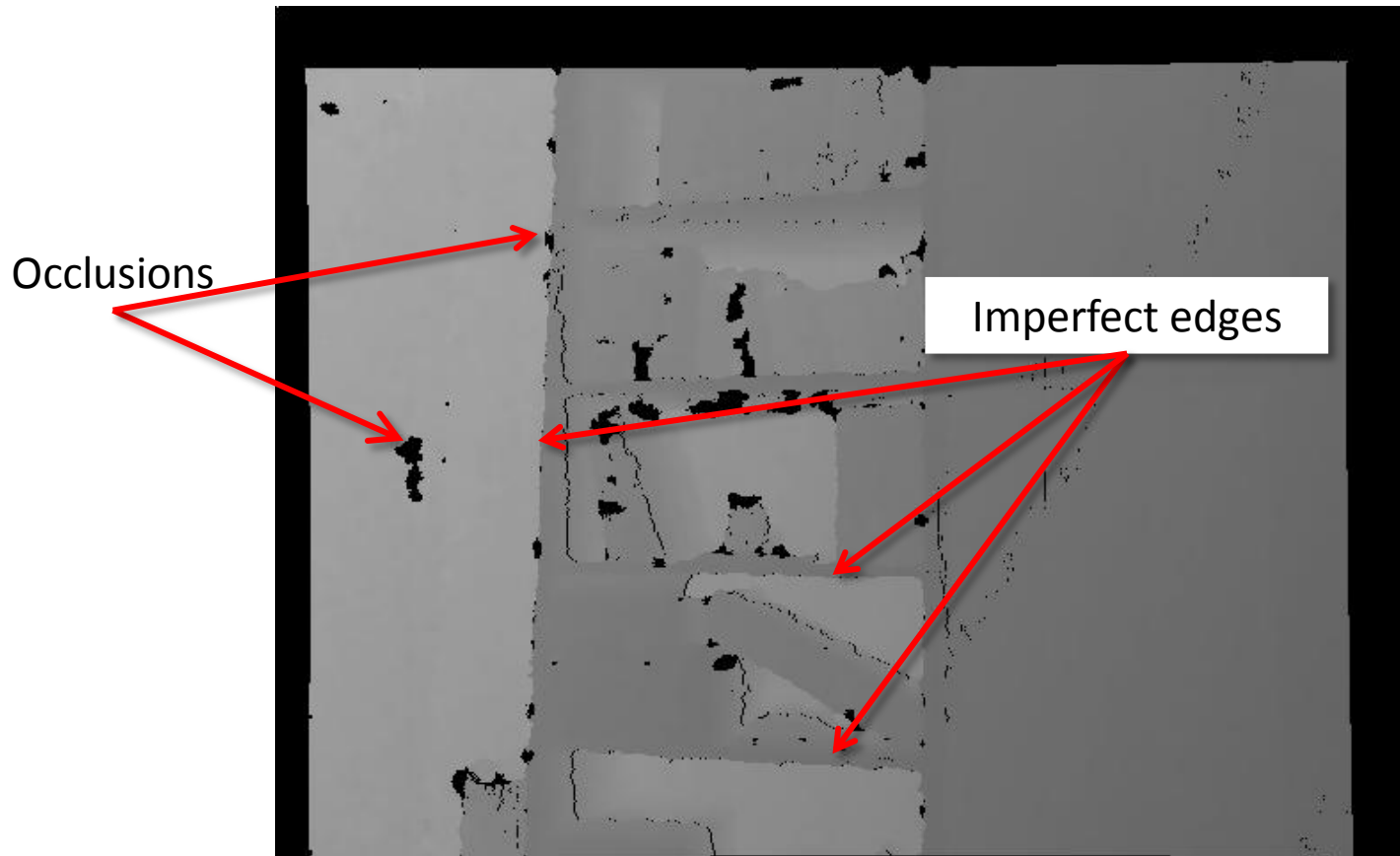
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- Depth Filtering
- Our Approach
- Results
  - Qualitative
  - Quantitative analysis method
- Conclusion

# Depth Filtering

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- Depth generation methods:
  - Active
    - Laser Scanner
    - ToF
    - Structured Light
  - Passive
    - Depth from stereo
    - Depth from motion
    - Depth from X
- There are no perfect depth maps



## Example depth map

- Kinect (structured light camera)

# OUR APPROACH

# Our Approach

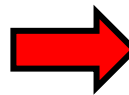
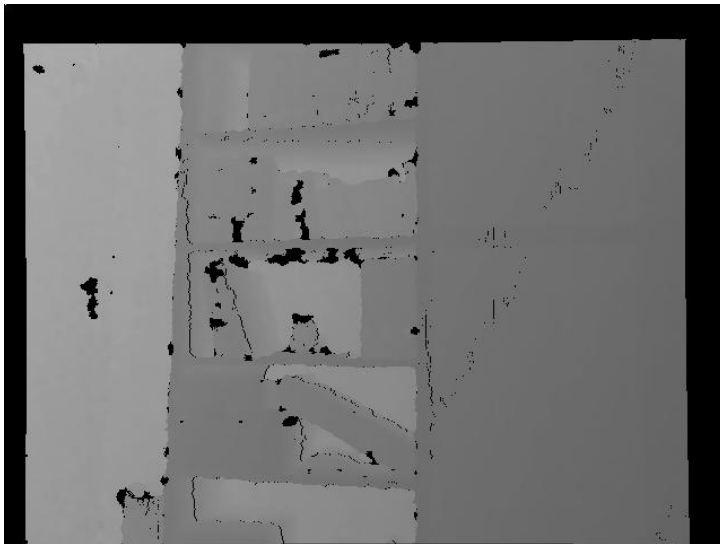
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- Focuses on edge restoration
- Takes edge information of associated color stream
- Workflow:
  1. Occlusion Filling
  2. Segmentation of color stream
  3. Computation of representative depth map
  4. Edge restoration
  5. Post processing

# Occlusion Filling

- Normalized convolution

$$D^{nc}(x) = \frac{\sum_{x' \in N_x^*} D(x)g(x, x')}{\sum_{x' \in N_x^*} g(x, x')}$$



# Color Segmentation

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- Edge information is taken from an oversegmentation (superpixel segmentation)
- We take Watershed segmentation because
  - Fast
  - Compact segments
  - Segments of approx. the same size (except thin “edge segments”)
- Color Segmentation:
  - Preprocessing of color stream (bilateral filter because of noise)
  - Apply Watershed
  - Cluster Splitting



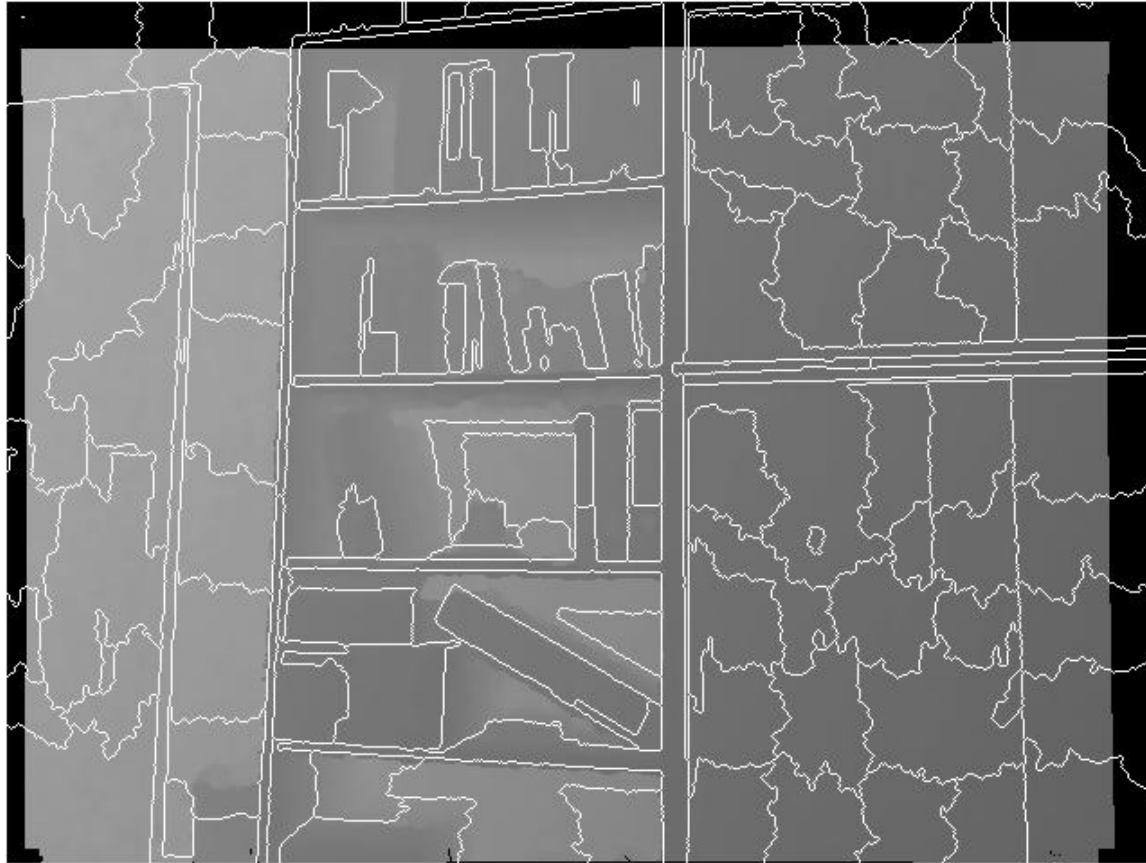
# Watershed Segmentation

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- Idea of Watershed:
  - Interpret Grayscale image as relief
  - Place water sources on it
  - Flood relief and draw borders where lakes meet
  - Apply Bilateral Filter prior to reduce noise



## Watershed Color Segmentation

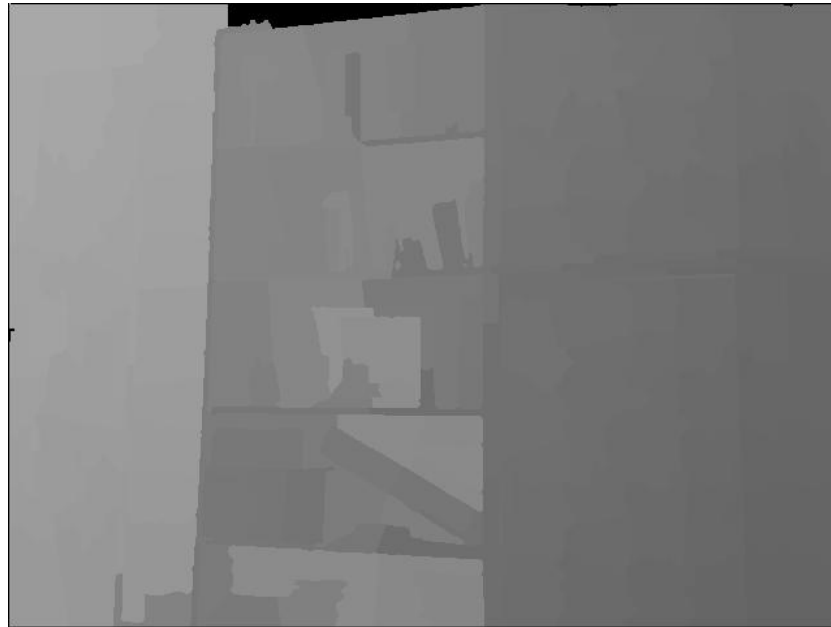


## Projected Color Segmentation in Depth

# Representative Depth Map

- Compute a representative depth value for each segment

$$D^r(x, y) = \{d_k : (x, y) \in S_k, d_k = \underset{(x', y') \in S_k}{\text{median}} d(x', y')\}$$



# Edge Restoration

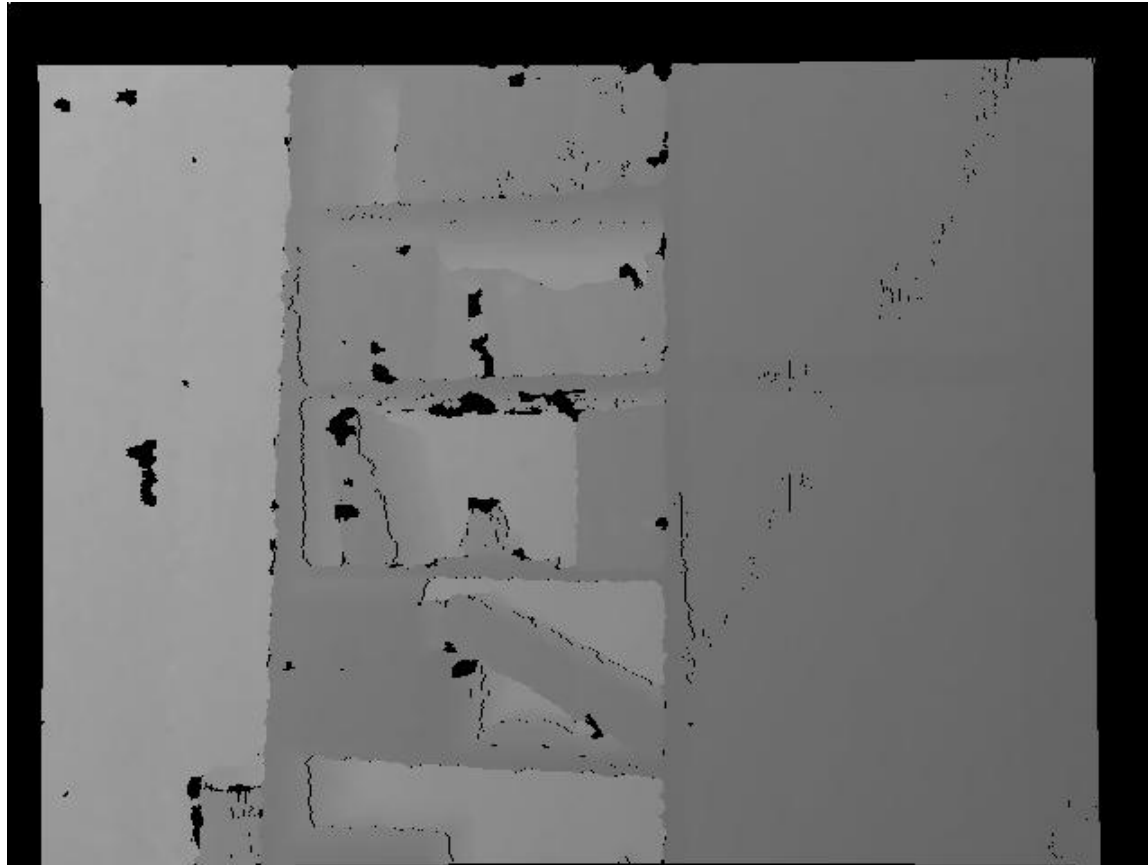
- Use representative depth map to enhance edges:

$$D^f(x, y) = \begin{cases} D^r(x, y) & \text{if } |D(x, y) - D^r(x, y)| > \theta \\ D(x, y) & \text{otherwise} \end{cases}$$

- Outliers are corrected by depth values of the representative depth map
- Postprocessing: Bilateral Filter

$$I(p) = \frac{\sum_{q \in N} K_s(\|p - q\|) K_c(\|p - q\|) I(q)}{\sum_{q \in N} K_s(\|p - q\|) K_c(\|p - q\|)}$$

# RESULTS



**Original Depth Map**



## Normalized Convolution [9]





**Berdnikov et al. [6]**



**Wasza et al. [7]**



## Our method

# Qualitative Results

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Input depth map



Our method

# Quantitative Results - Method

- Test sequence: Clear foreground and background



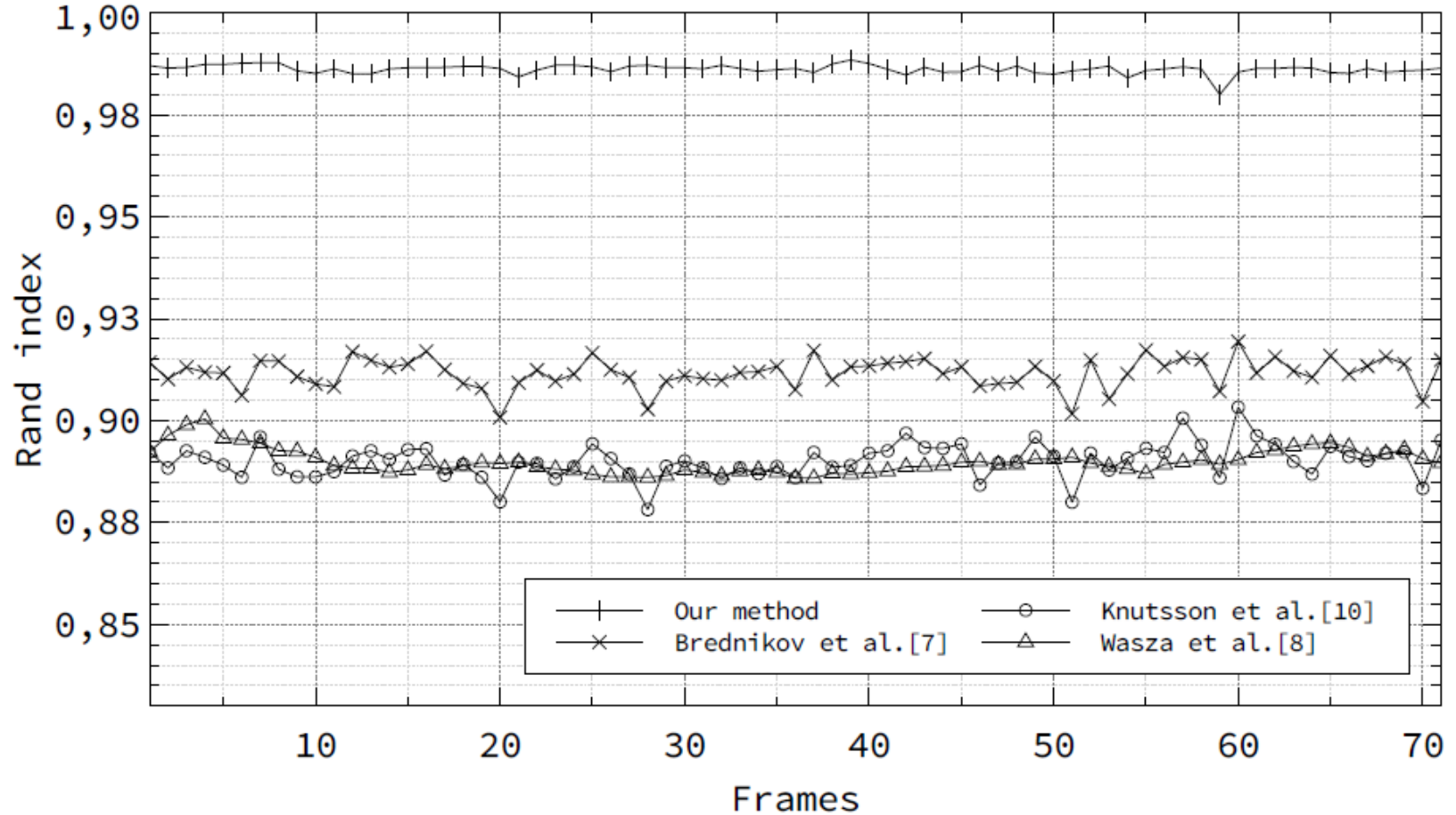
- Other geometry is possible

# Quantitative Results - Method

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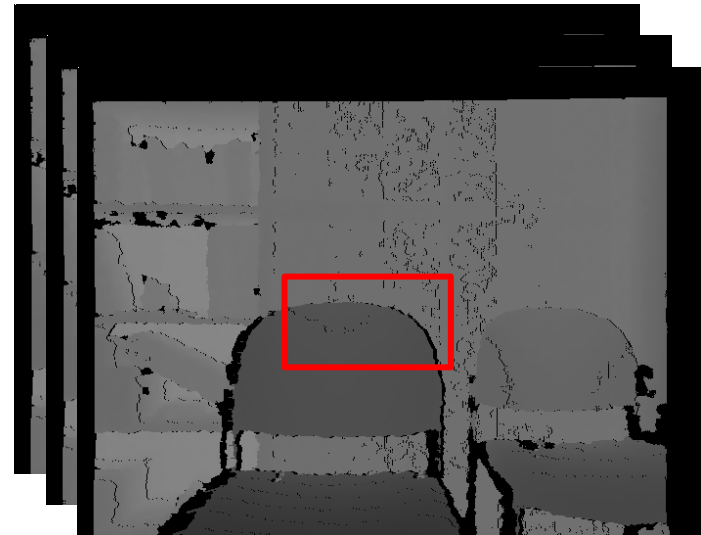
- Color frames define a clustering into foreground and background
- Depth frames define a clustering into foreground and background
- Perfect depth map -> Same clusterings
- Measure cluster similarity using Rand Index
  - Gives values between 0 and 1

# Quantitative Results



# Quantitative Results

- Test sequence 2:





# Quantitative Results

	Sequence 1	Sequence 2
Our method	0.9865	0.9778
Berdnikov [6]	0.9118	0.9129
Knutsson [9]	0.8952	0.9120
Wasza [7]	0.8899	0.9121

Mean Rand Index Values

# CONCLUSION

# Conclusion

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- We presented a new method for depth map enhancement
- Special focus on edge restoration
- We introduced a new method to quantify our results
- Our method shows promising results and outperforms others in terms of Rand Index values
- Future Work:
  - Add a temporal component
  - Make color segmentation temporal stable

# References

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- [1] Fehn, C., de la Barre, R., Pastoor, R.S.: Interactive 3-DTV-Concepts and Key Technologies. Proceedings of the IEEE 94 (2006) 524
- [6] Berdnikov, Y., Vatolin, D.: Real-time Depth Map Occlusion Filling and Scene Background Restoration for Projected-Pattern-based Depth Camera. In: 21th International Conference on Computer Graphics and Vision (GraphiCon2011). (2011)
- [7] Wasza, J., Bauer, S., Hornegger, J.: Real-time Preprocessing for Dense 3-D Range Imaging on the GPU: Defect Interpolation, Bilateral Temporal Averaging and Guided Filtering. In: IEEE International Conference on Computer Vision Workshops (ICCV Workshops). (2011)
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- [12] Beucher, S., Lantuejoul, C.: Use of Watersheds in Contour Detection. In: International Workshop on Image Processing: Real-time Edge and Motion Detection/Estimation, Rennes, France. (1979)
- [13] Rand, W.M.: Objective Criteria for the Evaluation of Clustering Methods. Journal of the American Statistical Association 66(336) (1971) pp. 846

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Thank you.

Questions?