

Paper presentation : UltraStereo: Efficient  
Learning-based Matching for Active Stereo  
Systems

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LOL

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# Time of flight

Varying active illumination to reconstruct a scene from multiple images

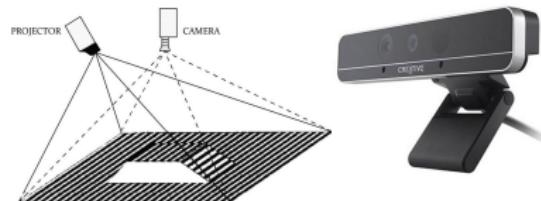


Shortcomings:

- Low mapping rate ( 30Hz)
- Motion artifacts
- Multipath-interference

# Structured Light

Spatial patterns projected on the scene

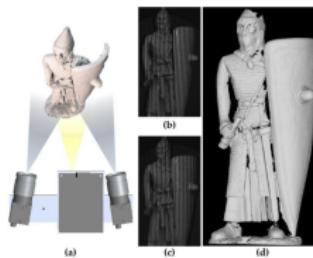


Shortcomings:

- Robustness issues
- Motion artifacts
- Limited depth-range
- Overlapping sensors cause interference

# Active Stereo

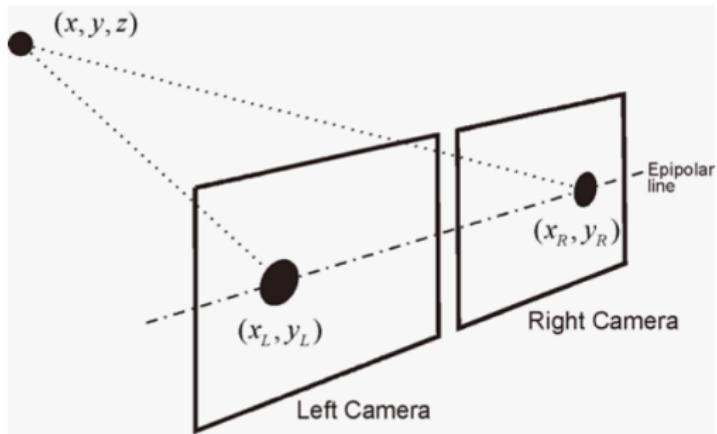
Using 2 calibrated cameras and 1 light source - solves a lot of the mentioned issues



- Mitigates multipath reflections
- Improves robustness
- Avoids interference between multiple systems
- BUT correspondence search has high computational cost!

# UltraStereo

- 2 IR cameras: monochrome Ximea cameras, 1280x1024 pixel at 210Hz
- DOE projector

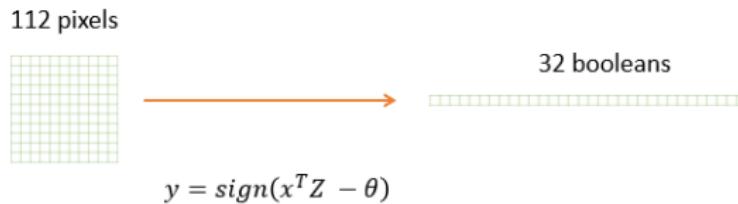


$$\text{disparity} = \frac{bf}{x_L - x_R} \quad (1)$$

$f$  = focal length

$b$  = baseline

# Dimensionality Reduction

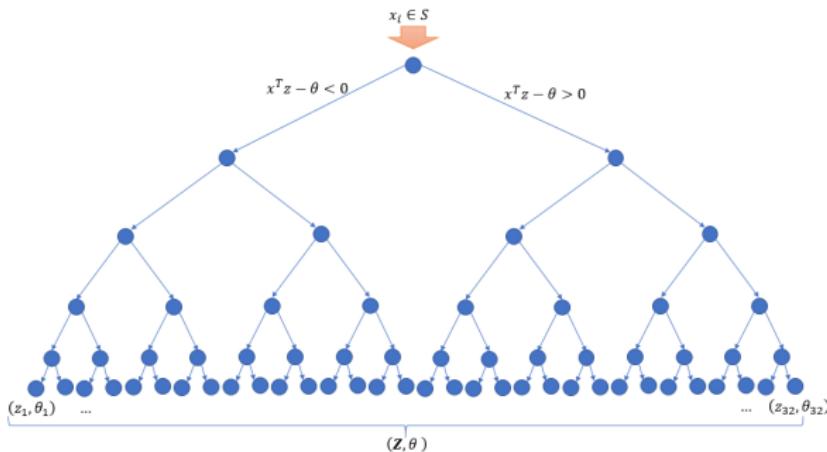


$x \in R^W$  with  $W = 11 \times 11$

$Z \in R^b$  with  $b = 32$

$\theta \in R^b$   $y \in (0, 1)^b$

# How to chose the parameters

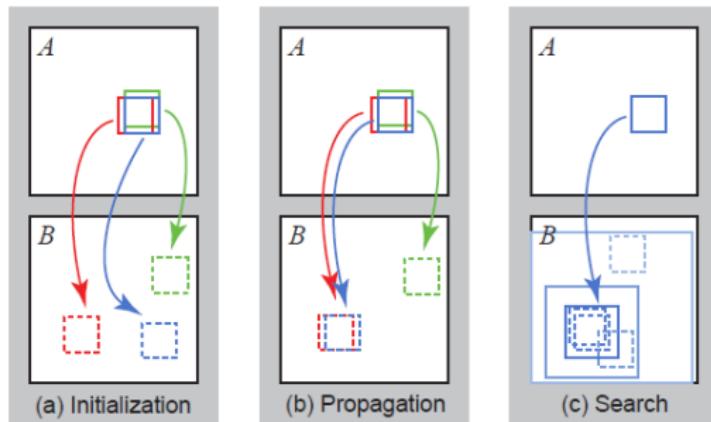


- $k \ll W \rightarrow$  sparse hyperplanes
- Information gain:  $I(\delta) = H(S) - \sum_{d \in L, R} \frac{S_d(\delta)}{S} H(S_d(\delta))$

$\delta$ : set of learned parameters  $(z, \theta)$

# Matching Framework

- Based on PatchMatch framework
- Iterative and randomized process
- 3 steps: initialization, propagation, post-processing



**Figure 2:** Phases of the randomized nearest neighbor algorithm:  
(a) patches initially have random assignments; (b) the blue patch checks above/green and left/red neighbors to see if they will improve the blue mapping, propagating good matches; (c) the patch searches randomly for improvements in concentric neighborhoods.

# Computational Analysis

- Linear with respect to the image size
- $O(1)$  complexity of the binary representation

# Evaluation

- 2500 training images and 900 test images (500 articulated hand images and 400 interior images for 5 different environments)

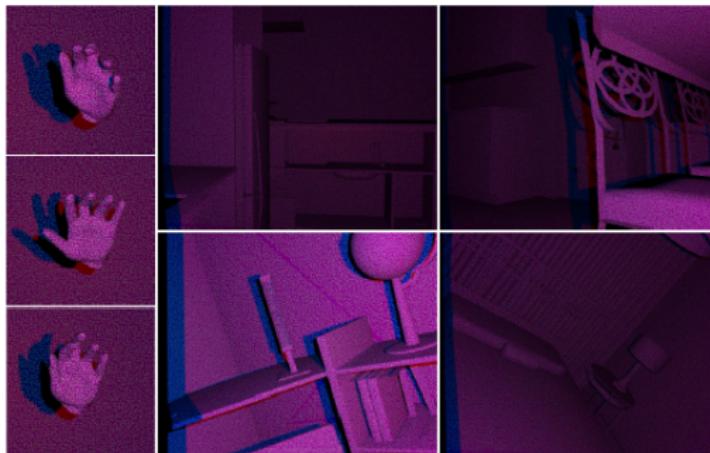


Figure 2. **Synthetic Data.** Representative examples of our synthetic dataset which comprises several indoor environments and a hand sequence.

# Bias

- Average absolute depth error present in the whole set

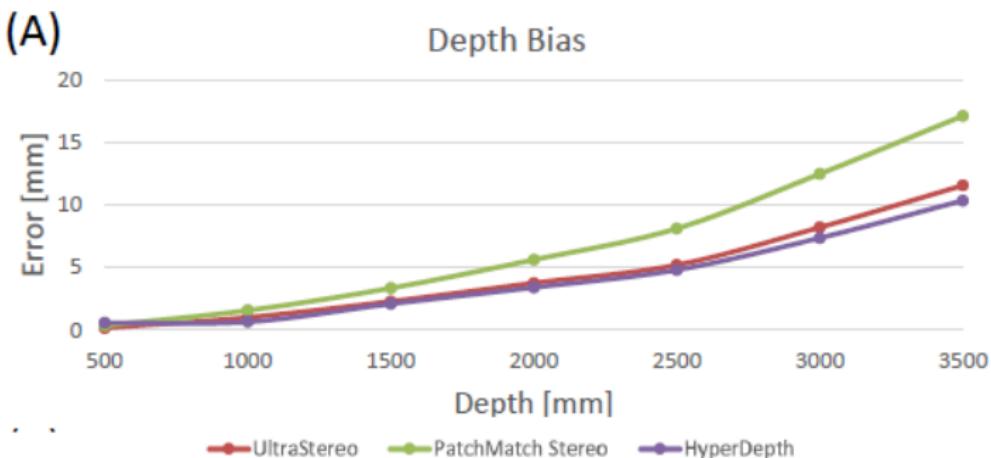


Figure: synthetic data

# Bias

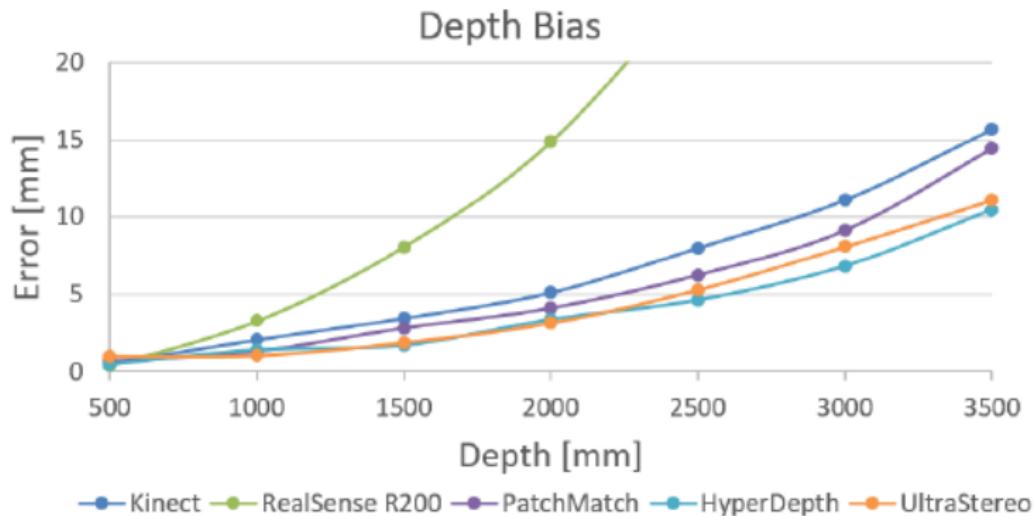


Figure: real data

# Invalidation

Issues which the pixels from the final depth image won't contain any estimations :

- Occlusions
- Saturation of infrared sensors
- Low signal noise ratio

How limits this errors ?

- An algorithms does an invalidation pass durint the post-processing step

# Invalidation

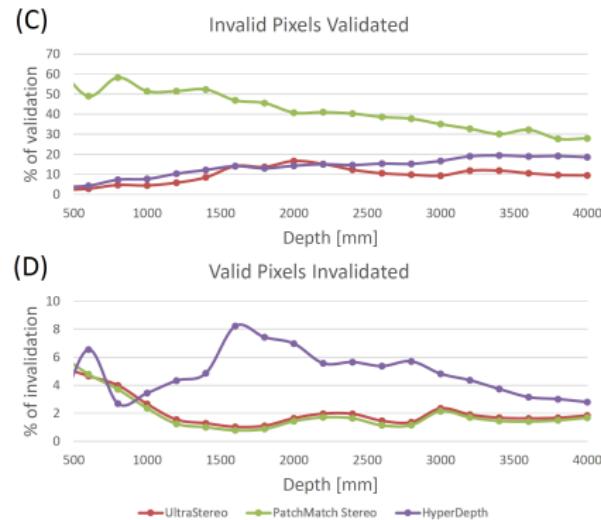


Figure: Quantitative results on syntatic data

# Example of depth-map produced with UltraStereo

Look at the thin structures like plants

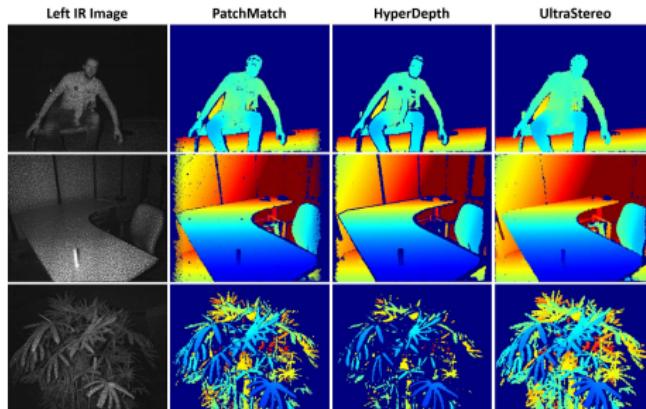


Figure: Qualitative Evaluation

# Edge fattening

Other issue is the edge fattening. To measure it they :

- Used a hand to test their algorithms
- Put a hand at 1 m from the sensors
- Defined key hand pose for each frame

# Edge fattening

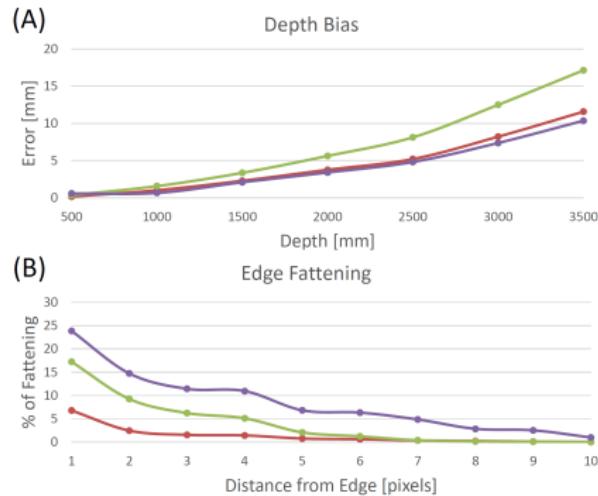
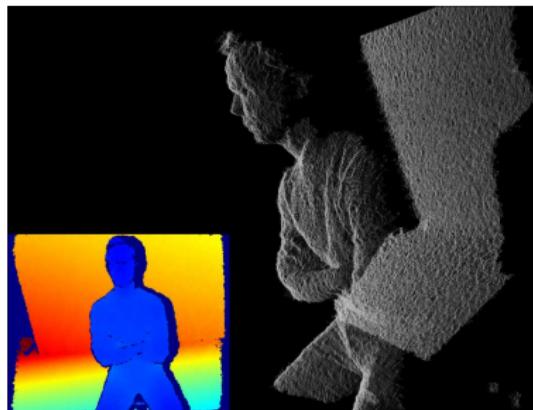


Figure: Quantitative results on syntactic data

# Edge fattening



**Figure:** Example of pointcloud produced with our algorithm. Notice the absence of quantization and flying pixels

# Binary representation

- Compare UltraStereo with Census and Locality Sensitive Hashing (LSH)
- Collect 1000 images with the Kinect

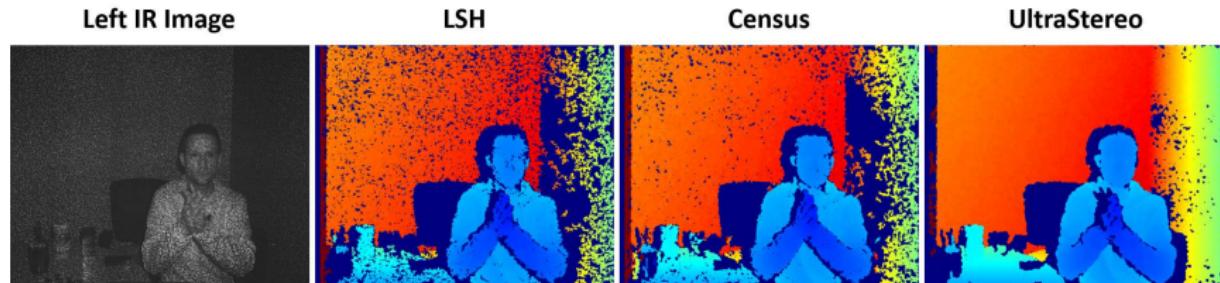
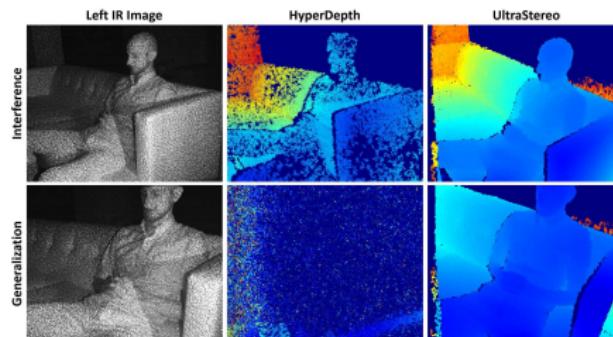


Figure: Census use 121 bits, LSH and UltraStereo use only 32 bits

# Interference and Generalization

- Interference caused by multiple sensors
- Camera calibrations

# Interference and Generalization



**Figure:** Examples of depth-maps produced with UltraStereo and state of the art competitors.

# Conclusion

- Breakthrough in the field of active stereo depth estimation
- Does not depend of the windows size or the size of the disparity space
- Use machine learning algorithms
- Run on GPU
- Does not suffer from camera calibrations nor interference problems

# Questions

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# Thank you for your attention !