



# Depth Sensing Beyond LiDAR Range

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## Motivation

#### **Self-driving datasets**

Kitti	80 meters		
Waymo	80 meters		



Image sources: velodyne lidar

**Question**: can we achieve *dense* depth sensing beyond LiDAR range with *low-cost cameras*? (e.g., >300 meters)

60 mph = 96 km/h = 27 m/s 80 meters roughly means 3 seconds

Example application:

Autonomous trucks driving on highway

## Long-range depth sensing is hard



#### Long-range LiDAR: sparse and expensive

## Long-range depth sensing is hard

# **Basic idea**: use two **cameras with telephoto lens** to capture a stereo pair, then reconstruct a dense depth map.







Industrial cameras<sup>[1]</sup>

#### Nikon P1000

Canon SX70

[1] Industrial cameras are usually much cheaper than consumer ones.

## Long-range depth sensing is hard

Important camera setup constraint:

Baseline is restricted to ~2 meters because of typical vehicle size.

What does this mean?

Depth estimation is **very sensitive** to pose error, especially rotation error.

It's difficult for hardwares to achieve and maintain this precision.



## **Tentative solution - SfM**

#### Bas-relief ambiguity in SfM<sup>[1]</sup>

Big focal length  $\rightarrow$  Near-orthographic camera (Weak perspectivity)





Figure 2: Ground-truth (blue) and the reconstructed (red) scene points. The unit for x, y, z axes is meter.

Figure 3: Top-down view of ground-truth relative pose (solid) and the recovered one (dashed).  $\theta$  is exaggerated for illustration.

[1] Richard Szeliski and Sing Bing Kang. Shape Ambiguities in Structure From Motion. In *Proc. European Conf. on Computer Vision (ECCV)*, pages 709–721. Springer, 1996.

## Our approach: a new three-camera vision system





## Our approach: novel depth estimation pipeline



**Proposition 1.** When a small-FOV camera is rotated by a small amount, the homography warping the original image to the rotated view is approximately an affine transformation.



30 50 70 90 110 130 Estimated uncalibrated disparity

## Our approach: novel depth estimation pipeline

**Proposition 2.** For two pixels in the left image with the same depth, if they are  $m_l$  pixels apart, while their corresponding pixels in the back image are  $m_b$  pixels apart, then the depth of these two pixels in the left camera's coordinate frame is



**Intuition:** to estimate this unknown offset, one essentially needs to know the metric depth of at least one 3D point.

## Our approach: novel depth estimation pipeline



## Results on synthetic data<sup>[1]</sup>

	Failure	<1%	<2%	<3%
Ours	0	45.3%	80.1%	96.9%
Loop and Zhang [18]	0	1.14%	2.73%	5.99%
SfM+MVS [19, 20]	15	6.71%	12.7%	19.1%

Table 1: Quantitative results on 40 synthetic scenes for methods in Fig. 7. "Failure" means the number of scenes for which a method fails to output a depth map. The metric is the portion of pixels with relative depth error below certain threshold, i.e., 1%, 2%, 3%, averaged over the successful scenes.

[1] Synthetic scenes might not be in their real-world scale. In experiments, we fix the baseline/depth ratio to be  $\sim$ 1/150.



Figure 7: Comparison among different algorithms. For rectification-based methods, the ground-truth depth map has been warped to align with the rectified view. For SfM, we have used the full ground-truth intrinsic matrix.

### **Results on real-world data**



Pseudo-rectified left view





Estimated unknown offset



Estimated final depth

- Ground-truth depth is acquired by the laser rangefinder: only pointwise measurement.
- Ground-truth: 302m Estimated: 300.8m



## Advantages

- Low-cost camera-based solution;
- Not require *full* pre-calibration of camera intrinsics and extrinsics;
- Robust to small camera vibrations: important when mounted on moving vehicles.

## Limitations

- Due to lack of equipment and facilities, the system has not been built and tested on the road with real-autonomous cars.
- Our method relies on stereo matching as backbone, thus suffering from common issues as stereo matching, e.g., textureless areas.

## Thank you!

More technical details can be found in our paper:

