

Motivation

- Depth from accidental motion

- Mobile phone camera (or) DSLR
 - Off-the-shelf Camera
 - Handshaking (Inevitable motion)
 - Narrow-baseline
 - 3sec (100 frames)



User-friendly

- Depth Applications

- Refocusing
- Background stylization



- Contribution

- High quality depth map is estimated from popularized cameras
- Feature extraction — remove features on moving objects using Essential matrices
- Accurately estimate depth of sparse 3D points with handling the rolling shutter artifacts
- Geometry guidance term assists to get high quality depth map

Sparse 3D reconstruction

- Feature extraction

- Harris corner & KLT tracker
- Outlier removal — Essential matrices from 5-pts algorithm
- Fundamental matrix from Essential matrices & pre-calibrated camera intrinsic parameters

$$\mathbf{l}_2 = \mathbf{F}\mathbf{x}_1, \quad \mathbf{l}_1 = \mathbf{F}^T\mathbf{x}_2, \quad \mathbf{l}_1^T\mathbf{x}_1 = 0, \quad \mathbf{l}_2^T\mathbf{x}_2 = 0$$

- Bundle adjustment

- LM optimization
- With camera pose from decomposition of Fundamental matrix
- Without depth information - Random depth initialization

$$C(\mathbf{r}, \mathbf{t}, \mathbf{X}) = \sum_{i=1}^{N_I} \sum_{j=1}^{N_J} \|\mathbf{x}_{ij} - \varphi(\mathbf{K}\mathbf{P}_{ij}\mathbf{X}_j)\|^2$$

- Geometric model for small motion

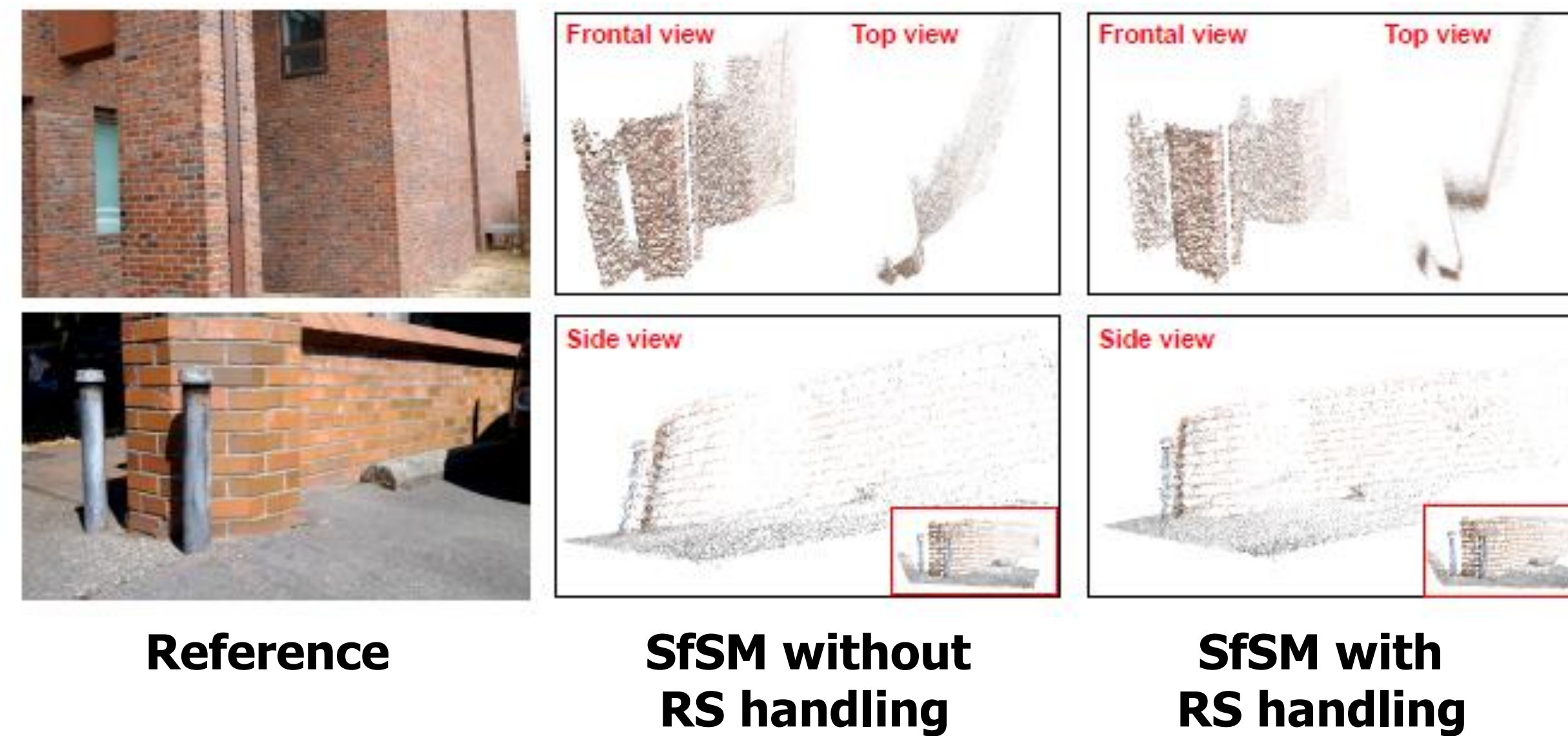
- Rotation matrix - Small angle approximation
- Key to estimate pose and depth without any prior depth information [1]

$$\mathbf{P} = [\mathbf{R}(\mathbf{r}) | \mathbf{t}], \text{ where } \mathbf{R}(\mathbf{r}) = \begin{bmatrix} 1 & -r^z & r^y \\ r^z & 1 & -r^x \\ -r^y & r^x & 1 \end{bmatrix}$$

- Rolling shutter effect handling
 - Rotation, Translation vector - Linear Interpolation

$$\mathbf{r}_{ij} = \mathbf{r}_i + \frac{ak_{ij}}{h}(\mathbf{r}_{i+1} - \mathbf{r}_i) \quad \mathbf{t}_{ij} = \mathbf{t}_i + \frac{ak_{ij}}{h}(\mathbf{t}_{i+1} - \mathbf{t}_i)$$

- Structure from Small Motion result with/without Rolling Shutter handling



Dense 3D reconstruction

- Objective function

$$E(\mathbf{D}) = E_d(\mathbf{D}) + \lambda_c E_c(\mathbf{D}) + \lambda_g E_g(\mathbf{D})$$

- Data term & Color smoothness term

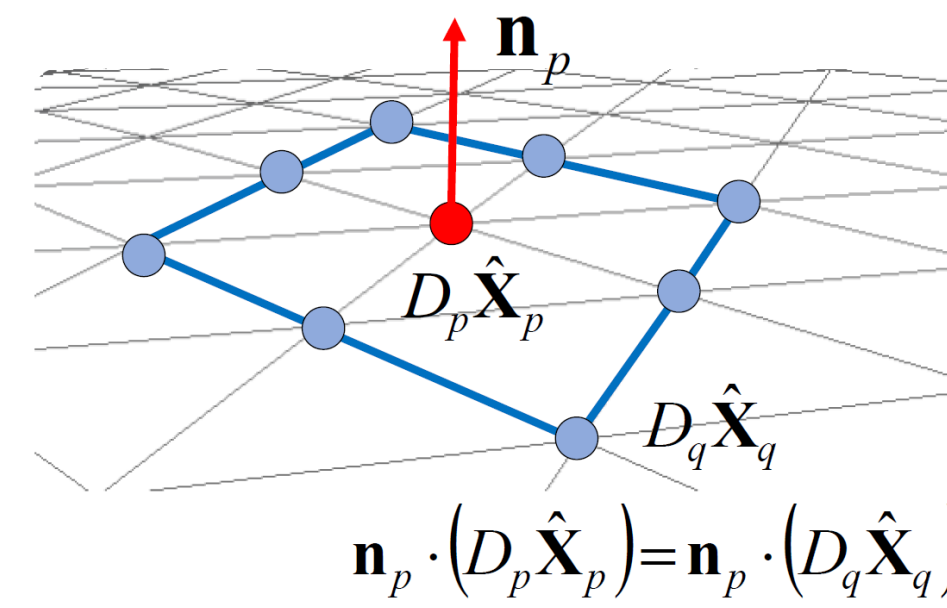
- Data term - Sparse 3D point cloud
- Color term — Neighboring pixel with similar color have similar depth

$$E_d(\mathbf{D}) = \sum_j \left(D_j - Z_j \right)^2 \quad E_c(\mathbf{D}) = \sum_p \sum_{q \in W_p} \left(D_p - \frac{w_{pq}^c}{\sum_q w_{pq}^c} D_q \right)^2$$

- Geometry guidance term

- Low spatial resolution
- Normal should be pre-calculated

$$E_g(\mathbf{D}) = \sum_p \sum_{q \in W_p} w_p^g \left(D_p - \frac{\mathbf{n}_p \cdot \hat{\mathbf{x}}_q}{\mathbf{n}_p \cdot \hat{\mathbf{x}}_p} D_q \right)^2$$



- Normal estimation

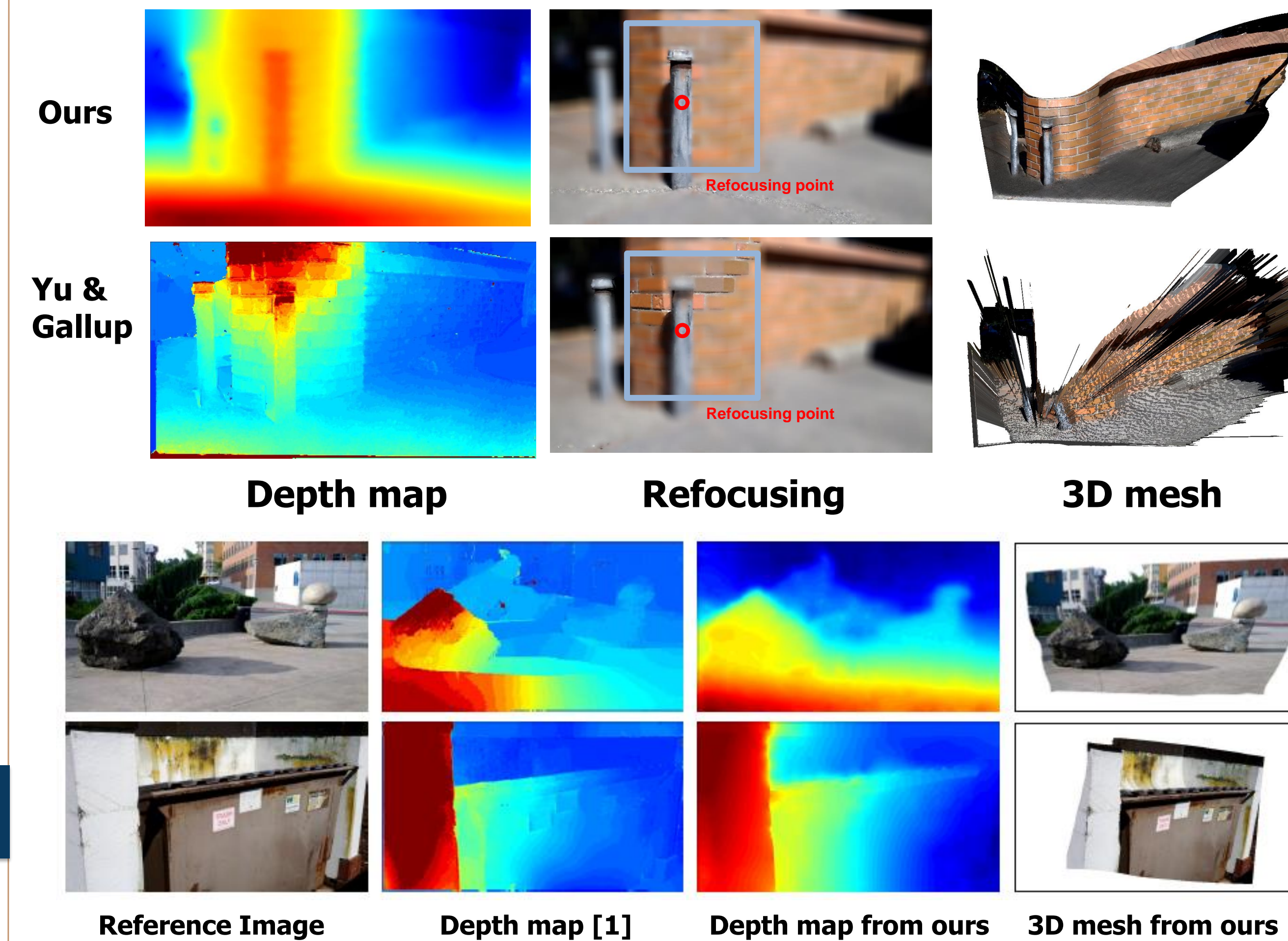
- Using local plane fitting
- Color-based propagation



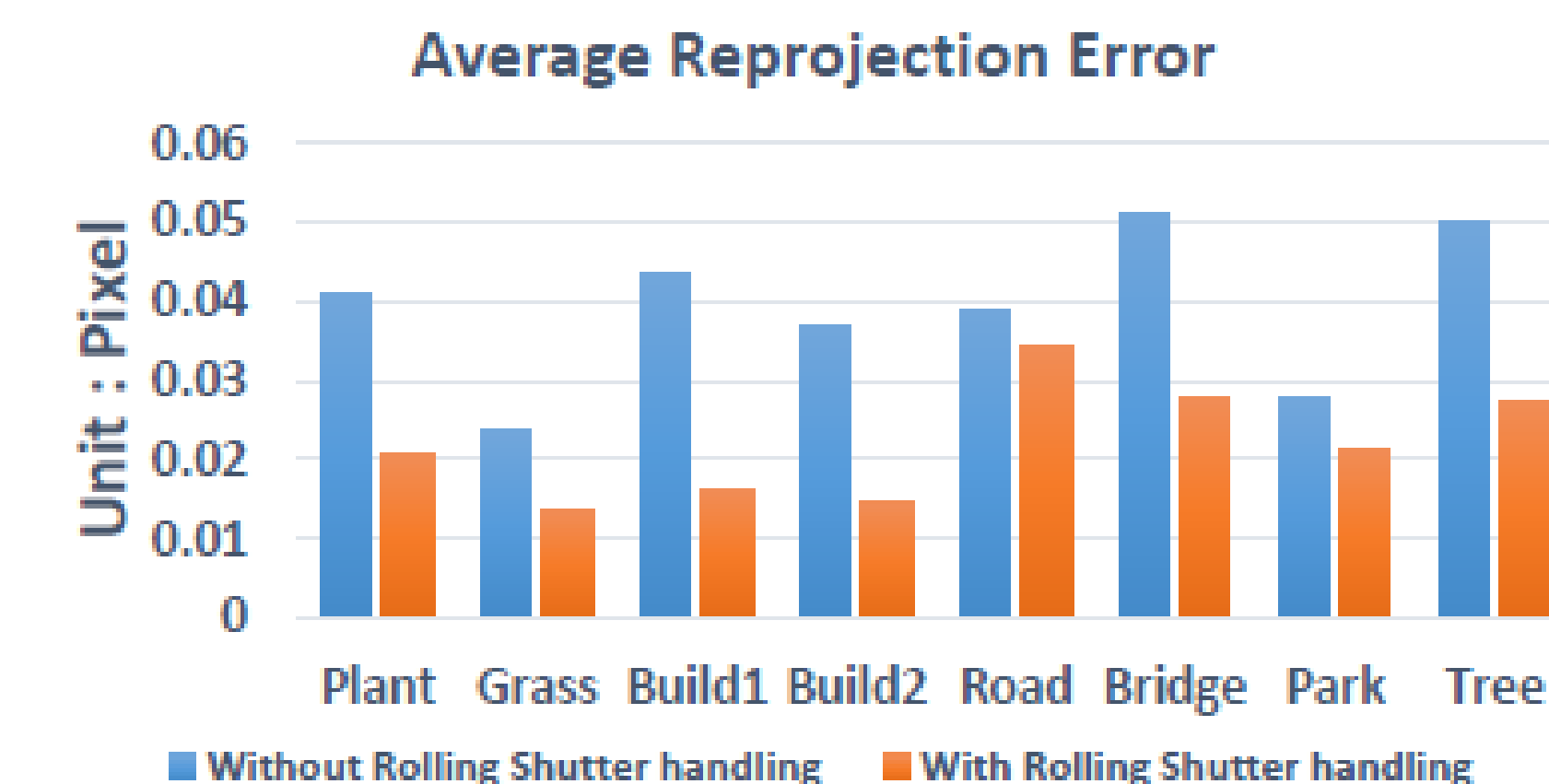
Experiments

- Comparison with state-of-the-art [1]

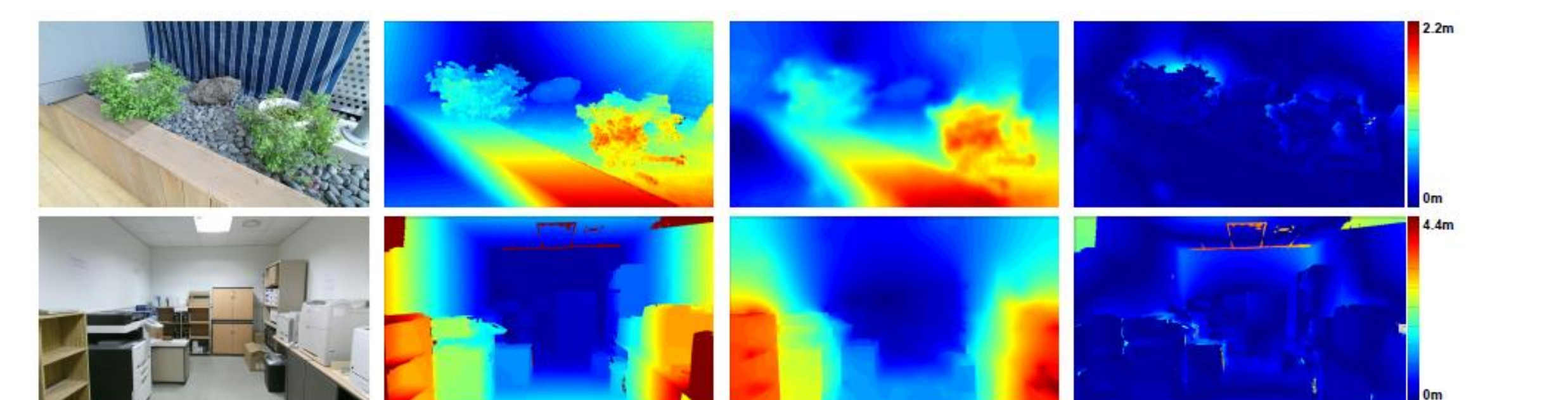
(Captured by Goole Galaxy Nexus)



- Structure from Small Motion result with/without Rolling Shutter handling



- Result comparison with the depth from Kinect (Captured by Kinect RGB sensor)

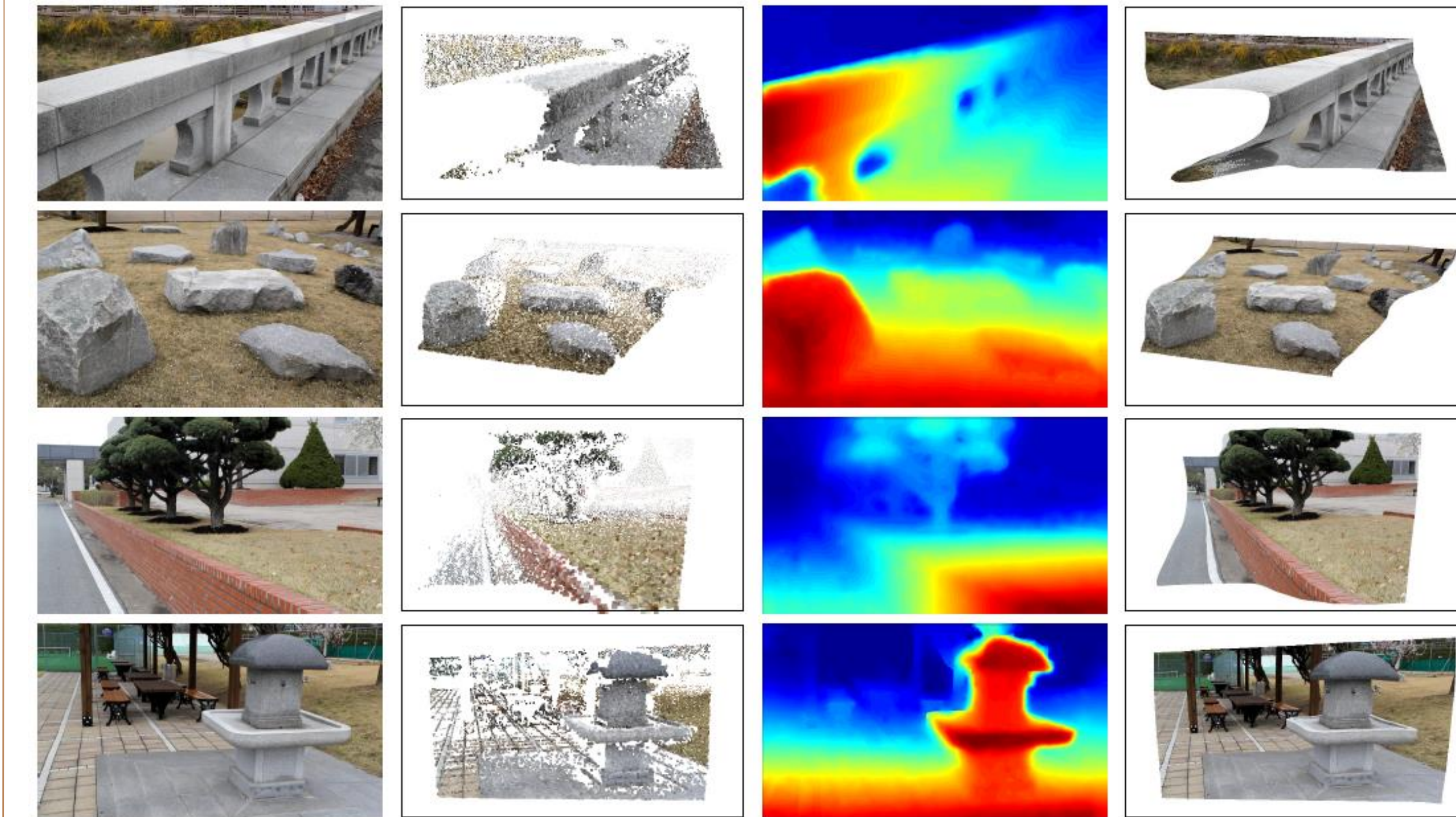


Dataset	Max. depth	R10	R20
Pot	2.2m	94.14%	99.07%
Room	4.4m	85.50%	96.31%

R10 : The percentage of pixels that have distance error less than 10% of the maximum depth value in the scene

- Our final results

(Captured by Canon EOS60D)



- Applications

- Refocusing



- Stylization



- Face Reconstruction



References

- [1] Yu, Fisher, and David Gallup. "3D Reconstruction from Accidental Motion." *Computer Vision and Pattern Recognition (CVPR), 2014 IEEE Conference on*. 2014.
- [2] N. Joshi and C. L. Zitnick. Micro-baseline stereo. Microsoft Research Technical Report MSR-TR-2014-73, 2014.
- [3] J. Hedborg, P.-E. Forssen, M. Felsberg, and E. Ringaby. 'Rolling shutter bundle adjustment. In Proc. of Comp. Vis. and Pattern Rec. (CVPR), 2012.



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