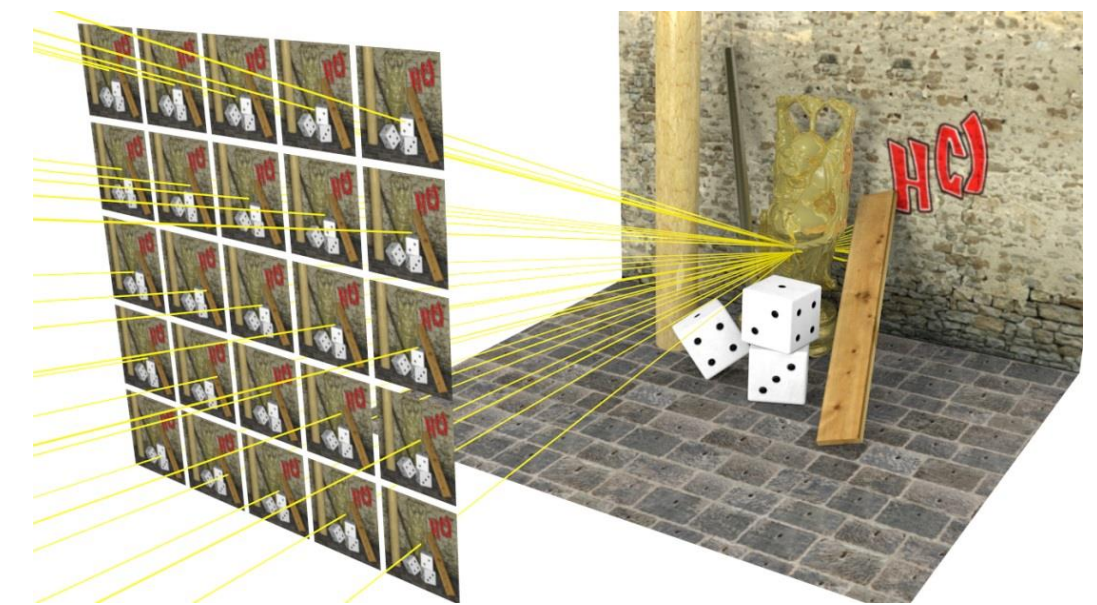


## Motivation

### Cameras for depth estimation

- Light-field camera
  - Low spatial resolution
  - Short depth range
- Stereo camera
  - Expensive



### Depth from accidental motion

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



User-friendly

### Contribution

- High quality depth map is estimated from popularized mobile phone
- Feature extraction — remove features on moving objects and extract very accurate features
- Sparse 3D reconstruction — set reliable initial camera poses
- Dense 3D reconstruction — geometry prior assists to remove an artifact on frontal parallel scene

## Feature extraction

- Feature extraction
  - Harris corner & KLT tracker
- Remove features on moving objects
  - Compute homography
  - Initial R, T compute by homography decomposition
  - Feature Inlier voting

$$H = K(R^T + \frac{Tn^T}{d})K^{-1}$$



## Sparse 3D reconstruction

- Geometric model for small motion
  - Rotation matrix - Small angle approximation
  - Key to estimate pose and depth without any prior depth information

### Bundle adjustment

- LM optimization
- Without depth information - Random depth initialization

$$F = \sum_{i=1}^{N_I} \sum_{j=1}^{N_F} ||p_{ij} - \pi(K(R_i P_j + T_i))||^2$$

## Dense 3D reconstruction

### Objective function

$$E(D) = E_c(D) + \lambda E_g(D)$$

### Color consistency

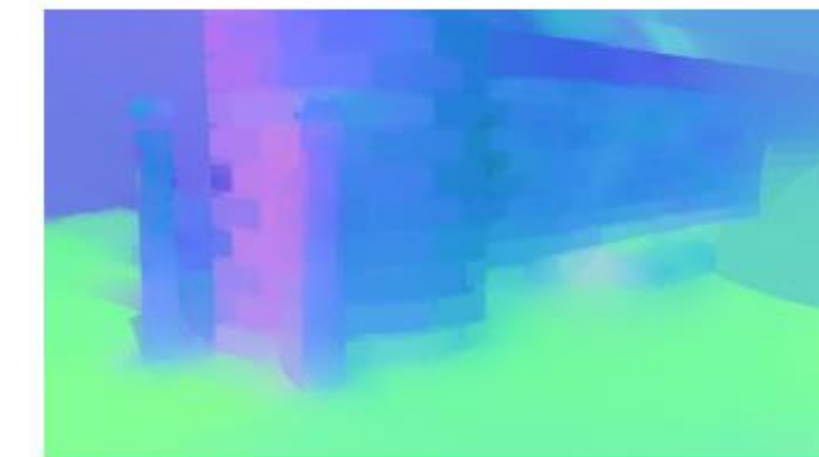
- Color term — Neighboring pixel with similar color have similar depth [2,3]

$$E_c(D) = \sum_{p \in I_{ref}} \left( D_p - \sum_{q \in N_8(p)} w_{pq}^c D_q \right)^2$$

$$w_{pq}^c = \frac{1}{N_c} \exp \left( -\frac{|I_p^{lab} - I_q^{lab}|}{\gamma_c} \right)$$

### Normal estimation

- Using local plane fitting
- Color-based propagation



Input

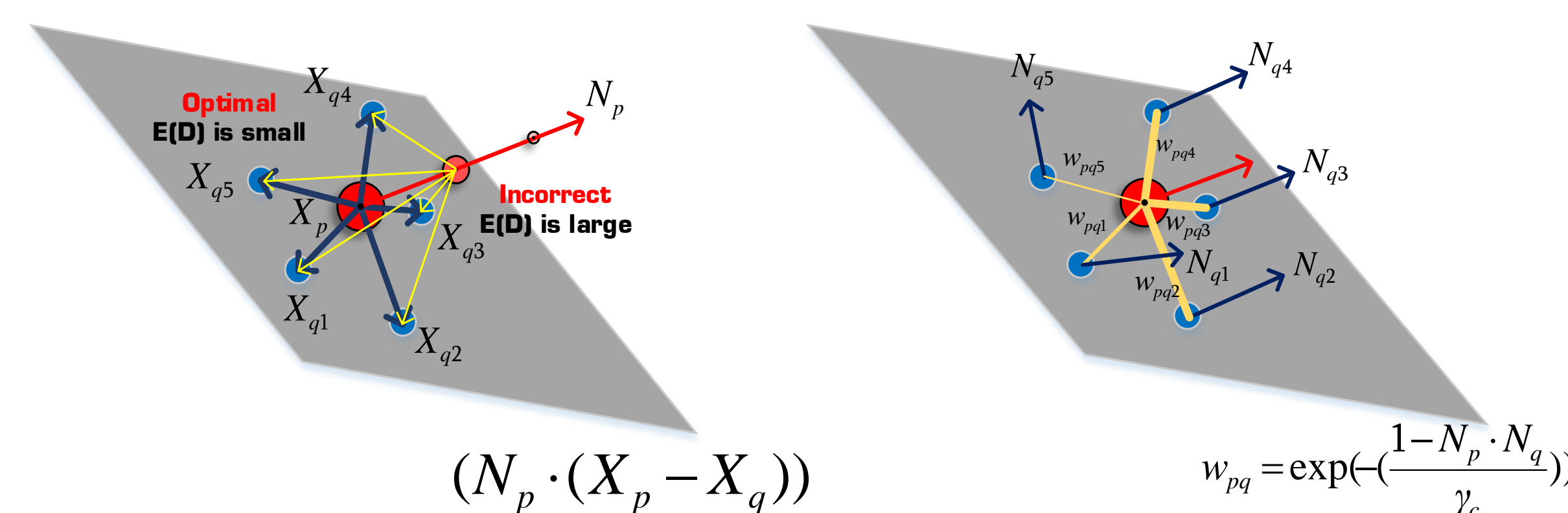
Sparse points

Normal map

### Geometry consistency

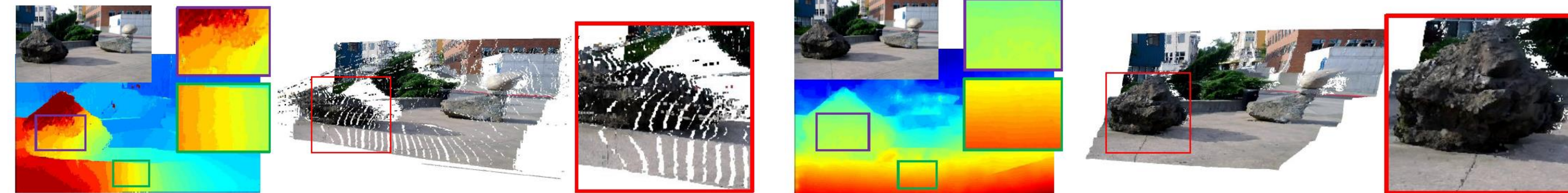
- Low spatial resolution
- Normal should be pre-calculated

$$E_g(D) = \sum_{p \in I_{ref}} \sum_{q \in G_w(p)} w_{pq}^g |n_p \cdot (D_p X_p - D_q X_q)|$$



## Experiments

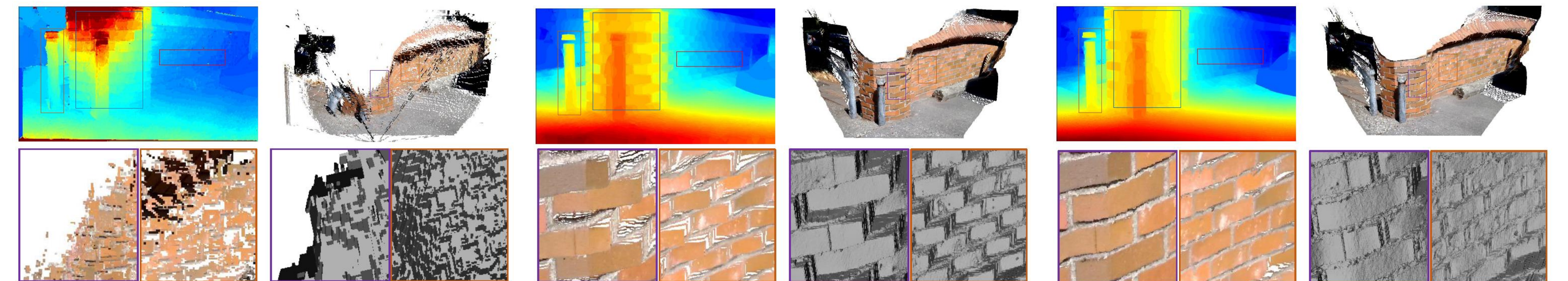
### Comparison with state-of-the-art (Captured by Google Galaxy Nexus)



Yu and Gallup [1]

Proposed

### Comparison with/without Geometry consistency (Captured by Google Galaxy Nexus)



Yu and Gallup [1]

Conventional Propagation [3]

Proposed Propagation

### Result of our datasets (Captured by Canon EOS60D)



### Reference

- Yu, Fisher, and David Gallup. "3D Reconstruction from Accidental Motion." Computer Vision and Pattern Recognition (CVPR), 2014 IEEE Conference on. IEEE, 2014.
- Wang, Liang, and Ruigang Yang. "Global stereo matching leveraged by sparse ground control points." Computer Vision and Pattern Recognition (CVPR), 2011 IEEE Conference on. IEEE, 2011.
- Park, Jaesik, et al. "High quality depth map upsampling for 3D-TOF cameras." IEEE International Conference on Computer Vision (ICCV), 2011.

### Acknowledge

This research is supported by the Study on Imaging Systems for the next generation cameras funded by the Samsung Electronics Co., Ltd (DMC R&D center) (IO130806-00717-02).

### Homepage



<https://sites.google.com/site/shimrcv/>