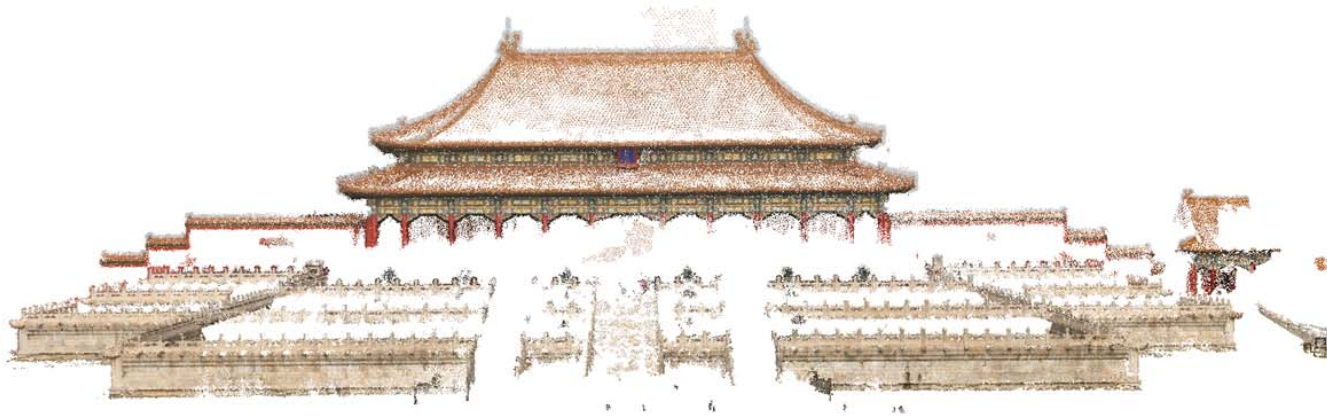


# **2016 Visual SLAM Report**

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# Simultaneous Localization and Mapping (SLAM)



# SLAM Components

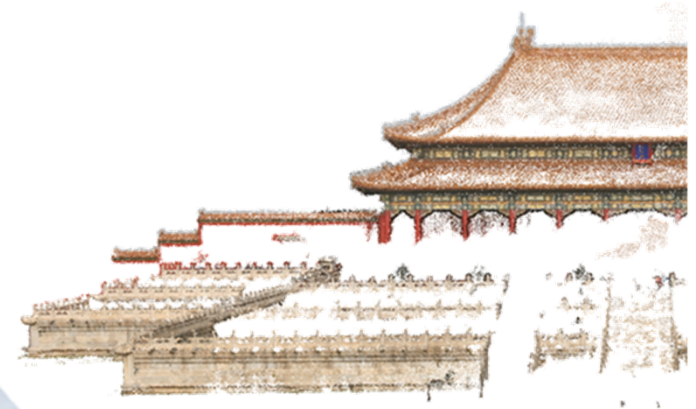
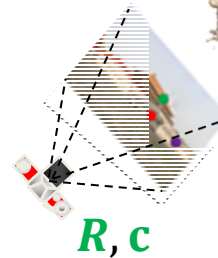
- Mapping (local)
    - Map reconstruction
  - Tracking
    - Map-to-image registration
  - Loops (global mapping)
    - Loop detection
    - Loop optimization
- } Odometry

# SLAM Categorization

- By tracking objective
  - Feature-based method
    - ORB-SLAM, PTAM, etc.
  - Direct method
    - LSD-SLAM, DSO
  - Semi-direct method
    - SVO

- By sensor type

- Camera (VO)
  - ORB-SLAM, LSD-SLAM, etc.
  - Event-camera SLAM (ECCV 2016 best paper)
- Camera + IMU (VIO)
  - OKVIS, ROVIO, VI-ORB, RKSLAM
- Depth camera (Fusion)
  - KinectFusion, BundleFusion, etc.



$$\rho(|\mathbf{Project}(\mathbf{Map}, \mathbf{R}, \mathbf{c}) - \mathbf{Observ}|)$$

# Major Progress in 2016

- By the tracking objective
  - ORB-SLAM2 (feature based method)
  - DSO (direct method)
  - SVO2 (semi-direct method)
- By the sensor
  - ORB-SLAM2, DSO, SVO2 (Camera)
  - VI-ORB, RKSLAM (Camera + IMU)
  - BundleFusion (Depth Camera)

# ORB-SLAM2

What's new from ORB-SLAM:

Support stereo and depth cameras

- New tracking objective function for stereo

$$\rho(|x_{(\cdot)}^i - \pi_{(\cdot)}(RX^i + t)|)$$

$$\pi_s \left( \begin{bmatrix} X \\ Y \\ Z \end{bmatrix} \right) = \begin{bmatrix} f_x \frac{X}{Z} + c_x, f_y \frac{Y}{Z} + c_y, f_x \frac{X - b}{Z} + c_x \end{bmatrix}$$

$x_s = (u_L, v_L, u_R)$

- A new formulation of depth camera tracking
  - Synthesize  $u_R$  according to  $(u_L, v_L)$  and depth
  - Adopt the stereo slam formulation



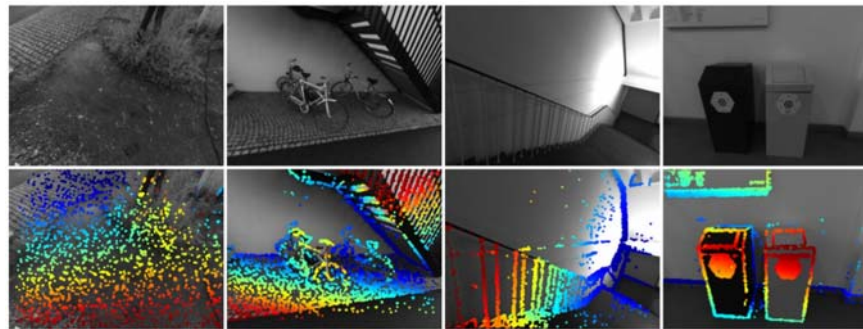
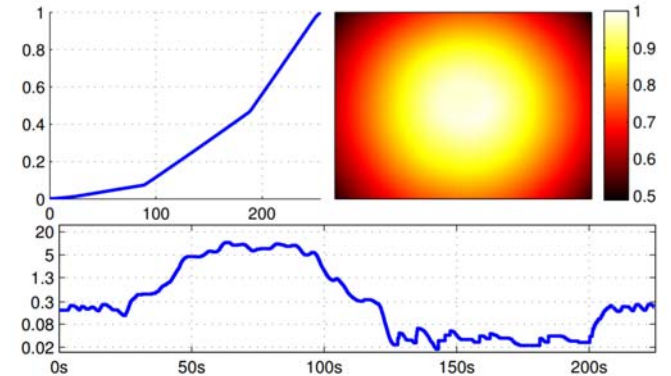
# DSO

## What's new from LSD-SLAM

- Auto photometric calibration on the fly

$$\rho \left( (I_j[\mathbf{p}'] - b_j) - \frac{t_j e^{a_j}}{t_i e^{a_i}} (I_i[\mathbf{p}] - b_i) \right)$$

- Photometric BA (solving both pts and cameras)
  - LSD solves pts and cameras iteratively
- Point sampling



# SVO2



## What's new from SVO

- Support multi-camera rig, wide FoV lens

$$\rho \left( I_k^c \left( \pi \left( T_{CB} T_{k,k-1} \mathbf{p} \right) \right) - I_{k-1}^c \left( \pi \left( T_{CB} \mathbf{p} \right) \right) \right)$$

- Include motion priors

$$T_{k,k-1} = \arg \min \sum_c \sum_{p \in R_{k-1}} \rho(\dots)^2 + |T_{k,k-1} - \tilde{T}_{k,k-1}|^2$$

- Use of edgelet features



360  
WWW.360.CN

SVO: Semi-Direct Visual Odometry for Monocular and Multi-Camera Systems,  
C. Forster, Z. Zhang, M. Gassner, M. Werlberger, D. Scaramuzza,  
IEEE Transactions on Robotics, 2016

SFU



# Better SLAM by Hardware or Software?

- DSO is more robust to motion blur
  - requires good lens, global shutter
- SVO is computationally efficient
  - works better on high fps camera
- VIO is more robust to blur & quick motion
  - requires good device synchronization
- Stereo SLAM is free from scale drift
  - requires two synchronized cameras
- Event camera is promising for brightness changes & quick motion
  - requires special hardware

# Thank you!

