

The Future of Real-Time SLAM

ElasticFusion: real-time dense SLAM without a pose graph

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18th December 2015 (ICCV Workshop)

State of the art in real-time dense SLAM

Plenty have limitations

- No (online) loop closure
- Only estimates trajectory
 - » Raw point cloud back projections
 - » Key frames
- Non-scalable
- Non-robust pose estimation

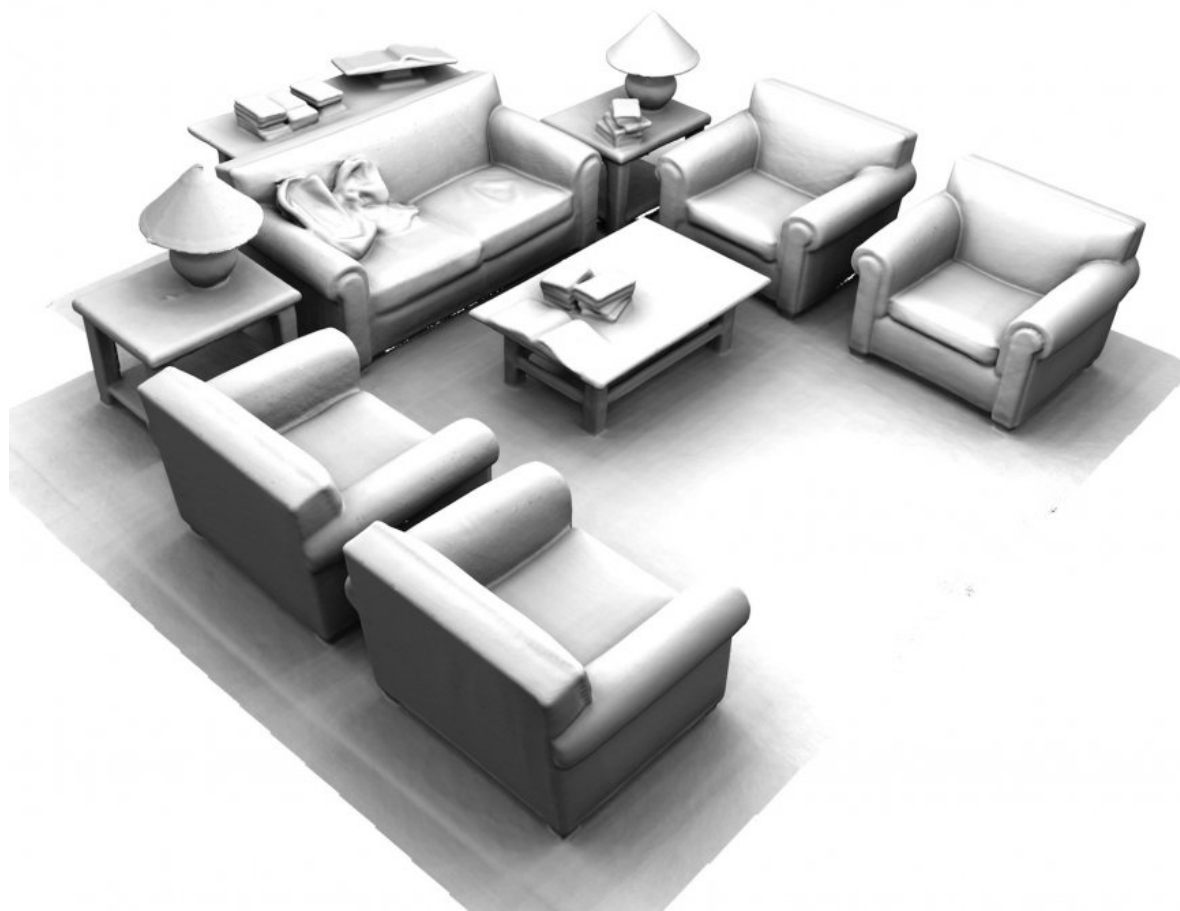
Some of these include;

- Henry et al., Endres et al., Meilland & Comport, Kerl et al., Keller et al., Chen et al., Nießner et al., Newcombe et al., Steinbruecker et al., Stueckler et al.

State of the art

Offline approaches

- Qian-Yi Zhou
Amazing results with RGB-D, strictly offline (>1hr processing time)



State of the art

Online approaches

- Kintinuous arguably the state of the art
 - » Online loop closure
 - » Estimates full 3D surface and trajectory
 - » Scalable (100's of metres)
 - » Full RGB and Depth pose estimation
- However, still not the “perfect” system
 - » A number of limitations

State of the art

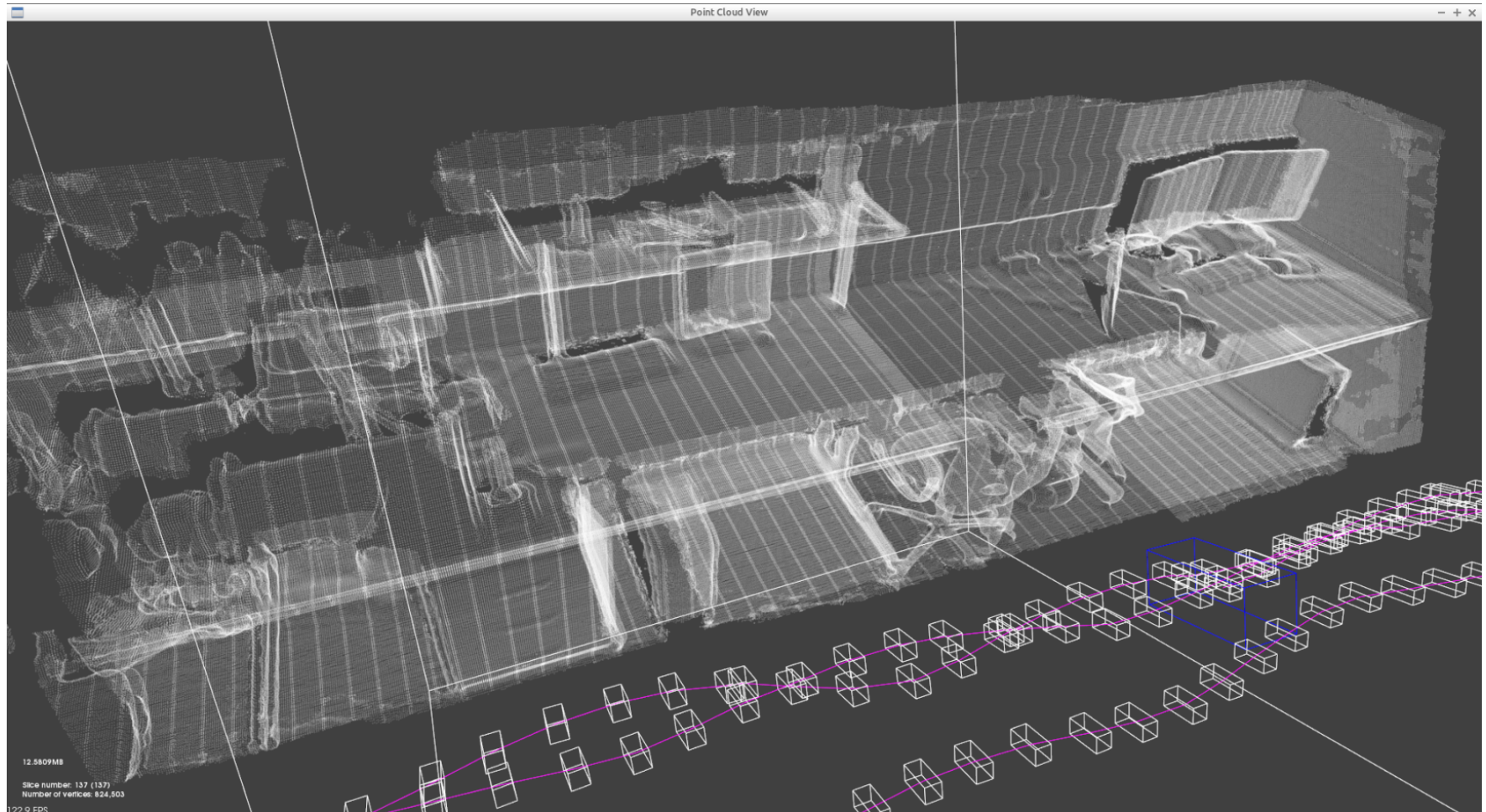
Kintinuous

- Surface aliasing



State of the art

Kintinuous



We want a SLAM system that...

Robustly estimates camera pose

- Geometry + photometry

Reliably estimates the surface

- Fused representation to remove noise

Scales well

- Room, house scale

Is completely closed loop (updating)

- Update revisited areas

Real-time

- Globally consistent map available at any point in time

Non-restrictive of motion

- Happy to deal with extremely loopy motion and many such loop closures

Introducing ElasticFusion

Robustly estimates camera pose

- Full RGB and Depth frame-to-model tracking

Reliably estimates the surface

- Point-based fusion is good quality and a nice representation

Scales well

- Room scale seems doable, at a minimum

Is completely closed loop (updating)

- No separation between front end and back end

Real-time

- Strictly

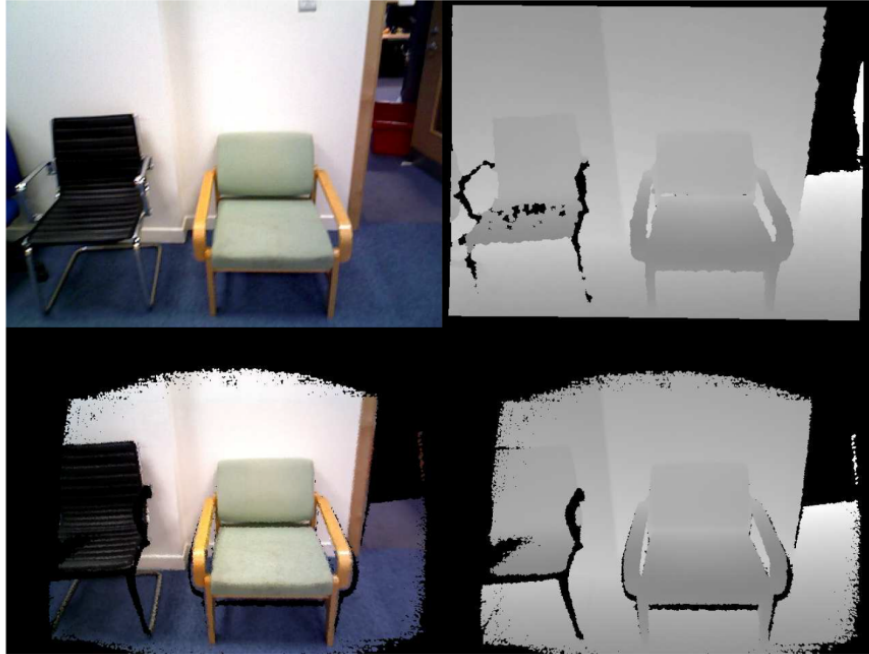
Non-restrictive of motion

- Since the front end and back end are one and the same, it is less restrictive given the full frame-to-model tracking

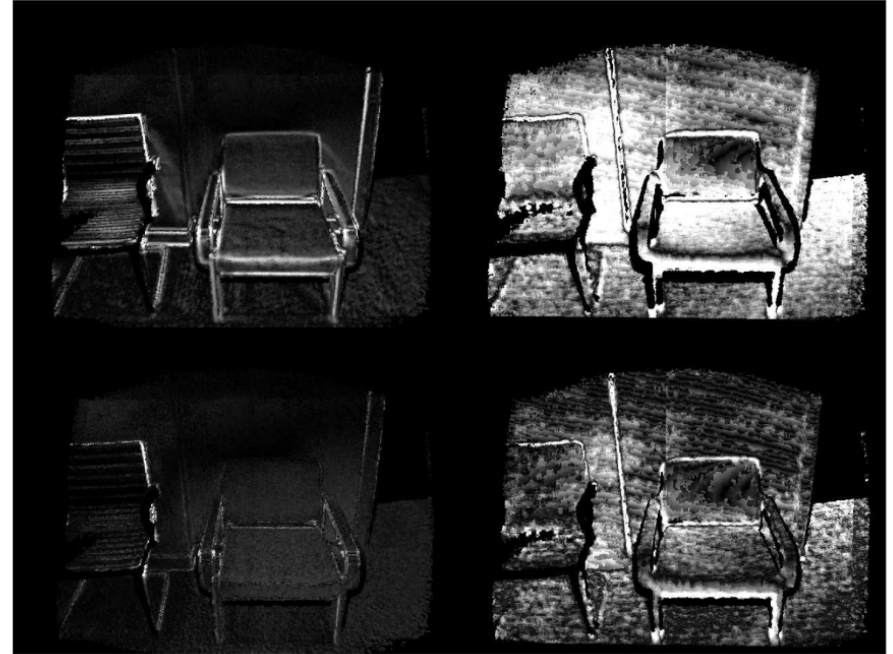
How it works

1. Reconstruct surfel-based map of environment
2. Split into active/inactive regions
3. Directly register multiple passes of the same surface together
4. Reflect this in the map with a non-rigid space deformation
5. Use fern encoded key frames for global loop closures

How it works: Tracking



Data terms

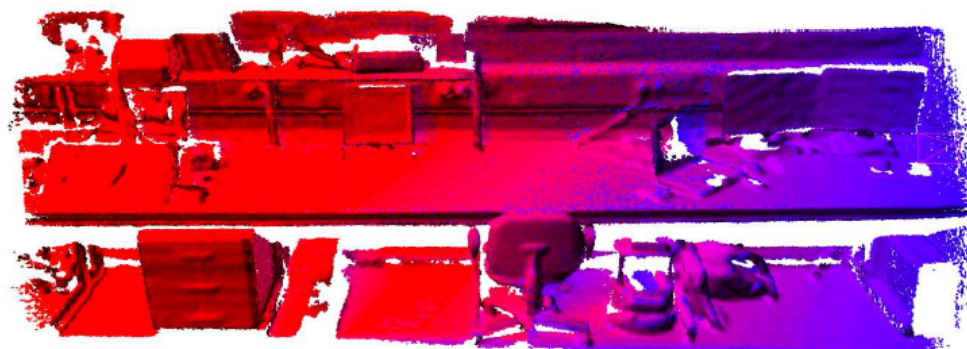


Residuals

$$E_{rgb} = \sum_{\mathbf{u} \in \Omega} \left(I(\mathbf{u}, \mathcal{C}_t^l) - I\left(\pi(\mathbf{K} \exp(\hat{\xi}) \mathbf{T} \mathbf{p}(\mathbf{u}, \mathcal{D}_t^l)), \hat{\mathcal{C}}_{t-1}^a \right) \right)^2$$

$$E_{icp} = \sum_k \left(\left(\mathbf{v}^k - \exp(\hat{\xi}) \mathbf{T} \mathbf{v}_t^k \right) \cdot \mathbf{n}^k \right)^2$$

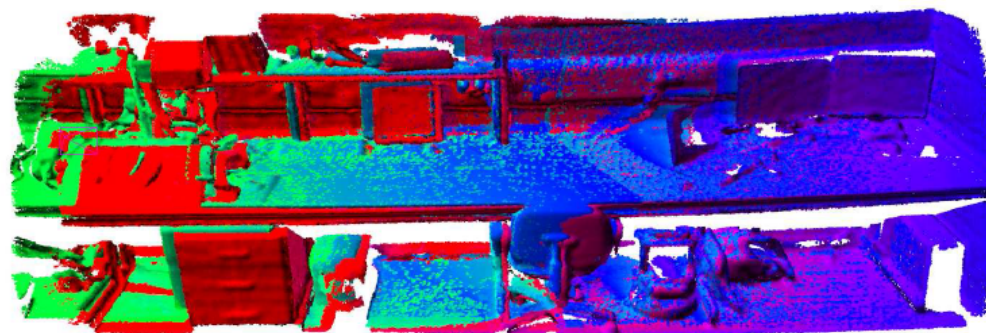
How it works: Building a Deformation Graph



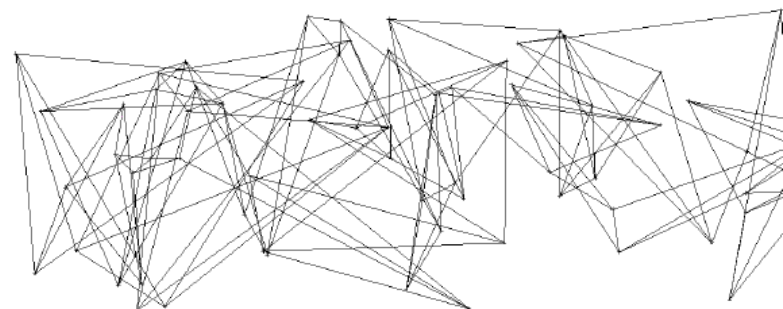
Mapping left to right



Time scale

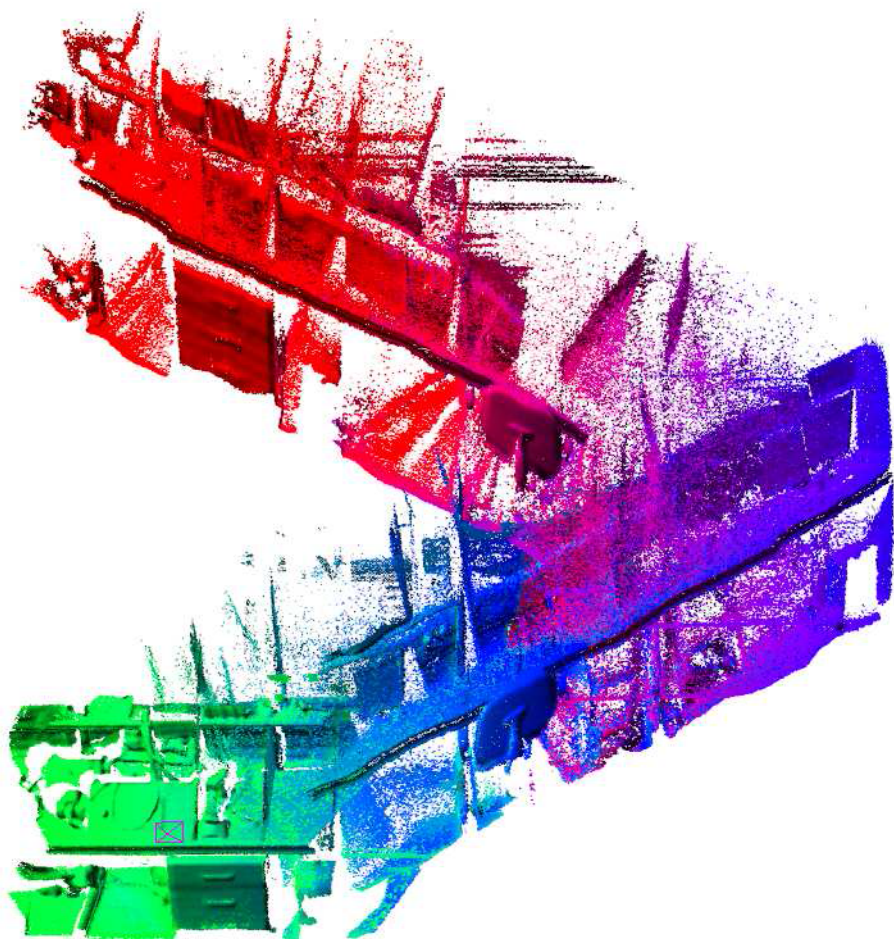


Mapping right to left

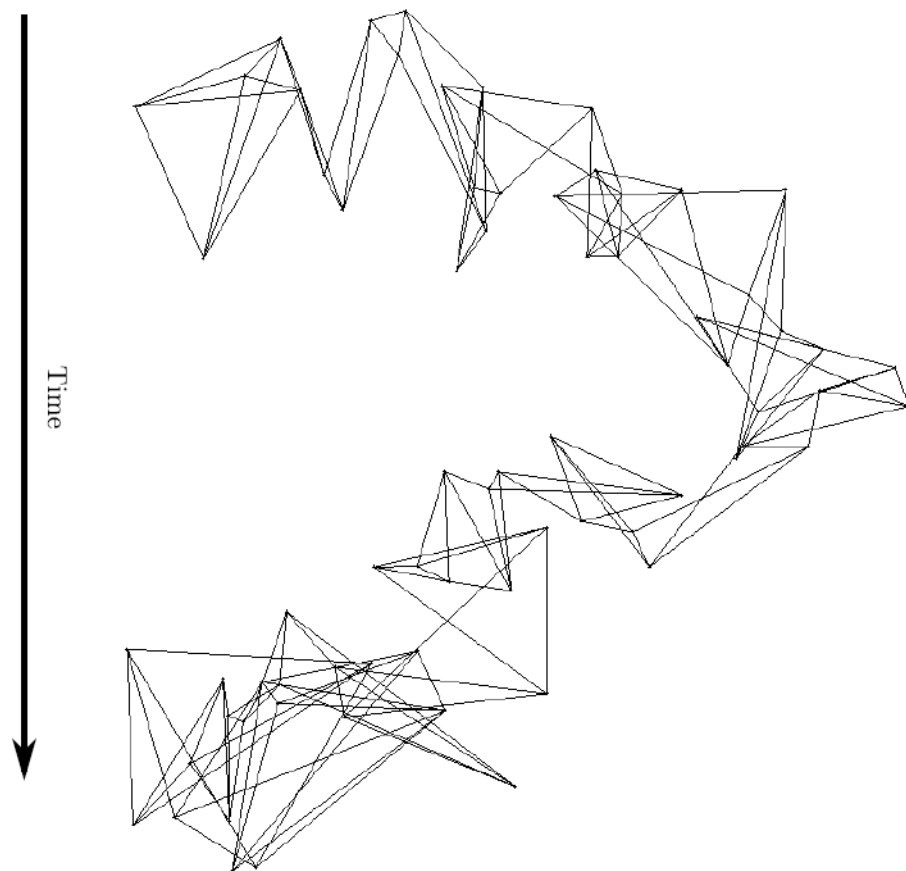


Deformation graph

How it works: Time Stretched Visualisation



Time stretched map



Time stretched graph

How it works: Loop Closure

$$E_{rot} = \sum_l \left\| \mathcal{G}_{\mathbf{R}}^l{}^\top \mathcal{G}_{\mathbf{R}}^l - \mathbf{I} \right\|_F^2 \quad \text{As-rigid-as-possible}$$

$$E_{reg} = \sum_l \sum_{n \in \mathcal{N}(\mathcal{G}^l)} \left\| \mathcal{G}_{\mathbf{R}}^l (\mathcal{G}_{\mathbf{g}}^n - \mathcal{G}_{\mathbf{g}}^l) + \mathcal{G}_{\mathbf{g}}^l + \mathcal{G}_{\mathbf{t}}^l - (\mathcal{G}_{\mathbf{g}}^n + \mathcal{G}_{\mathbf{t}}^n) \right\|_2^2 \quad \text{Smoothness regulariser}$$

$$E_{con} = \sum_p \left\| \phi(\mathcal{Q}_{\mathbf{s}}^p) - \mathcal{Q}_{\mathbf{d}}^p \right\|_2^2 \quad \text{New-to-old loop closure constraints}$$

$$E_{pin} = \sum_p \left\| \phi(\mathcal{Q}_{\mathbf{d}}^p) - \mathcal{Q}_{\mathbf{d}}^p \right\|_2^2 \quad \text{Old-to-old anchoring constraints}$$

$$E_{rel} = \sum_p \left\| \phi(\mathcal{R}_{\mathbf{s}}^p) - \phi(\mathcal{R}_{\mathbf{d}}^p) \right\|_2^2 \quad \text{Relative constraints (previous loop closures)}$$

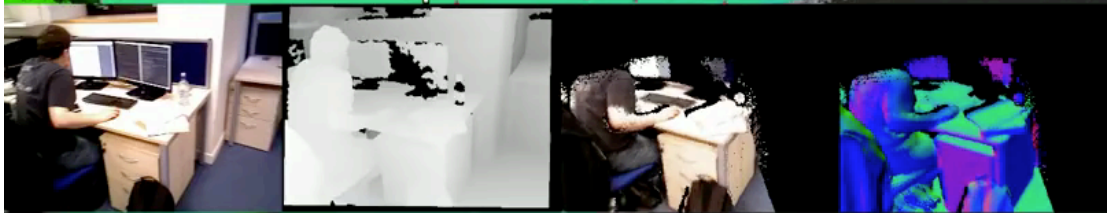
ElasticFusion

Overview (Real-time)

Revisited inactive areas are detected and trigger
local dense surface loop closures

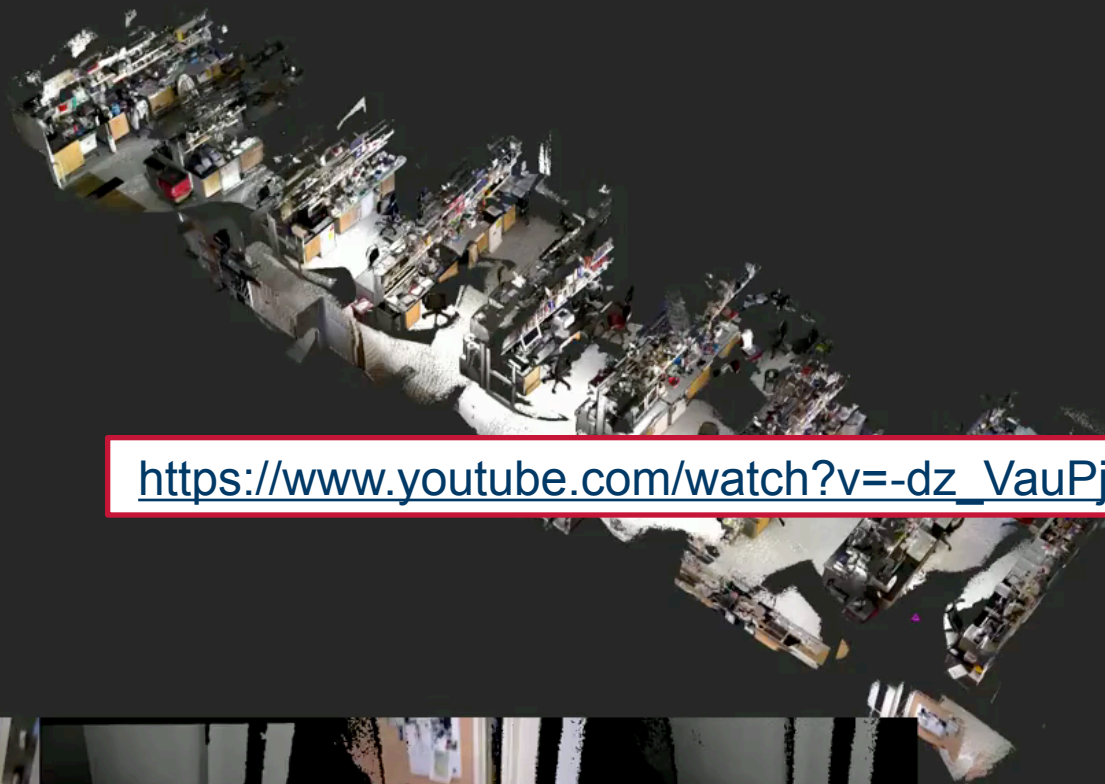
<https://www.youtube.com/watch?v=XySrhZpODYs>

Live Colour Live Depth Active Colour Active Surface

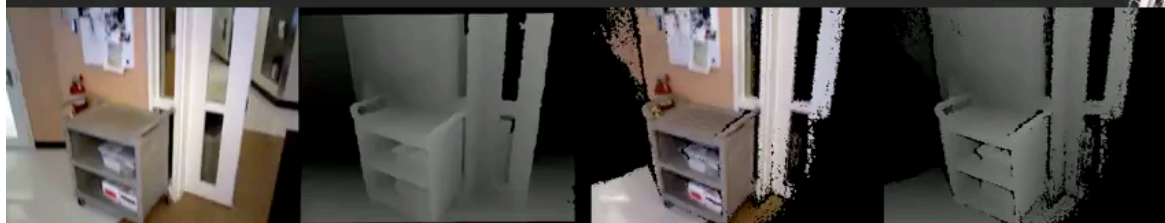


ElasticFusion – Extras

MIT_76_417b dataset (Real-time)



https://www.youtube.com/watch?v=-dz_VauPjEU



Quantitative Results

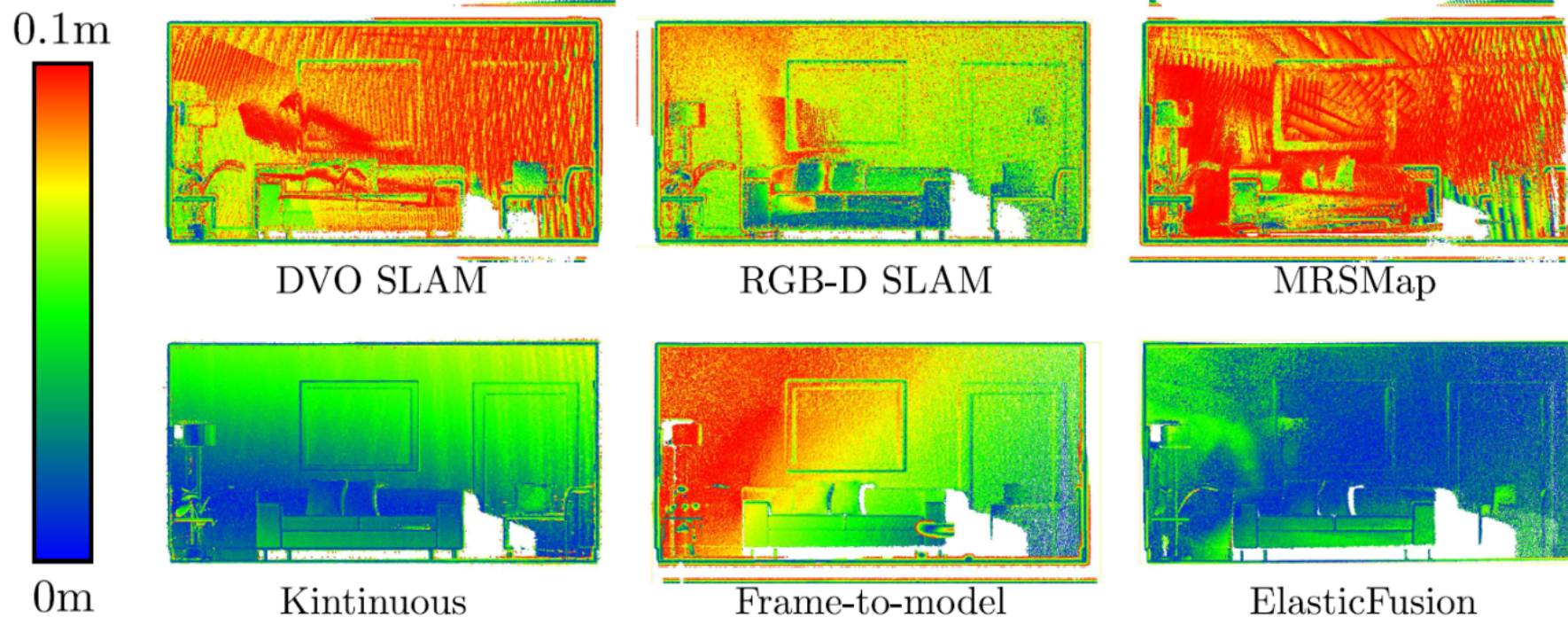
System	fr1/desk	fr2/xyz	fr3/office	fr3/nst
DVO SLAM	0.021m	0.018m	0.035m	0.018m
RGB-D SLAM	0.023m	0.008m	0.032m	0.017m
MRSMap	0.043m	0.020m	0.042m	2.018m
Kintinuous	0.037m	0.029m	0.030m	0.031m
Frame-to-model	0.022m	0.014m	0.025m	0.027m
ElasticFusion	0.020m	0.011m	0.017m	0.016m

TUM RGB-D Trajectory Error

System	kt0	kt1	kt2	kt3
DVO SLAM	0.032m	0.061m	0.119m	0.053m
RGB-D SLAM	0.044m	0.032m	0.031m	0.167m
MRSMap	0.061m	0.140m	0.098m	0.248m
Kintinuous	0.011m	0.008m	0.009m	0.150m
Frame-to-model	0.098m	0.007m	0.011m	0.107m
ElasticFusion	0.007m	0.007m	0.008m	0.028m

ICL-NUIM Surface Error

Quantitative Results



Main Advances

Real-time deformation

- Great for overcoming the drift problem in a dense map

Fully closed loop

- No frontend/backend division opens up many possibilities

Open source

- <https://github.com/mp3guy/ElasticFusion>

Light Source Estimation

Reflectance-driven

- Detect speculars

Why?

- It's cool
- Convincing AR effects
- Can be used to improve tracking
- Aids in path planning (i.e. avoid bright areas)

Light Source Estimation

Reconstruct diffuse appearance

Bright raw observations are reflected rays

- Vote in voxel space (hough-like scheme)

Intersections of high votes and geometry are potentially light sources

- Geometry helps determine extent of light source, directionality and removes need for comprehensive reflected ray coverage

Light Source Estimation

