# Autonomous Reconstruction of Unknown Indoor Scenes Guided by Time-varying Tensor Fields

Kai Xu, **Lintao Zheng**, Zihao Yan, Guohang Yan, Eugene Zhang, Matthias Niessner, Oliver Deussen, Daniel Cohen-Or, Hui Huang



**Shenzhen University** 



**Oregon State University** 



**Stanford University** 



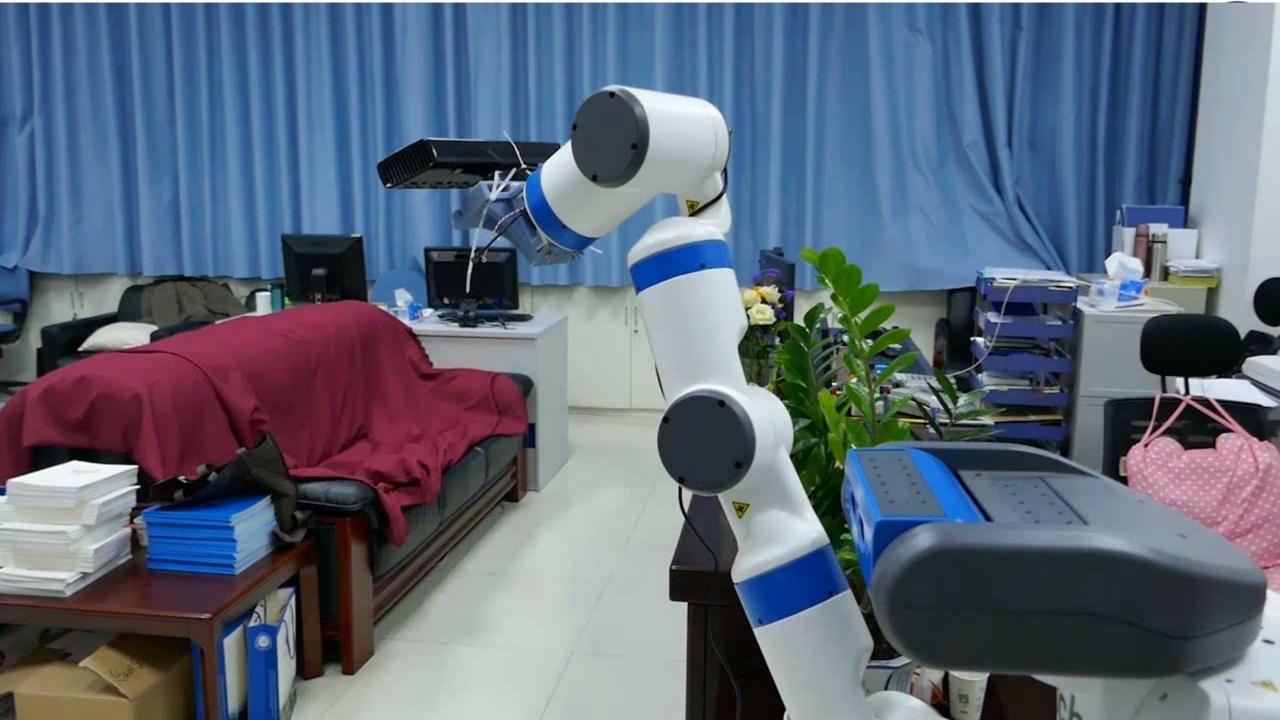
**National University of Defense Technology** 



University of Konstanz



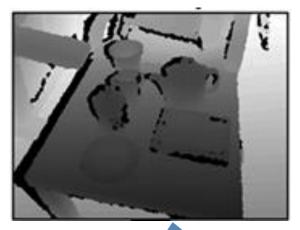
Tel-Aviv University



## **Background**

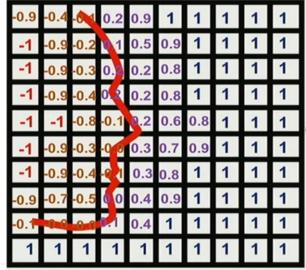


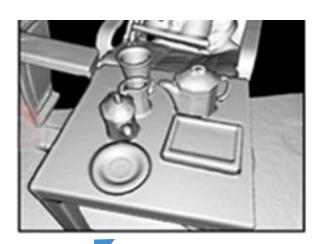
Commodity RGBD sensors & real-time reconstruction



Registration & fusion (Localization)

KinectFusion [Izadi et al. 2011]





Reconstruction (Mapping)

# **Background**



- Human scanning is a laborious task



#### **Motivation**



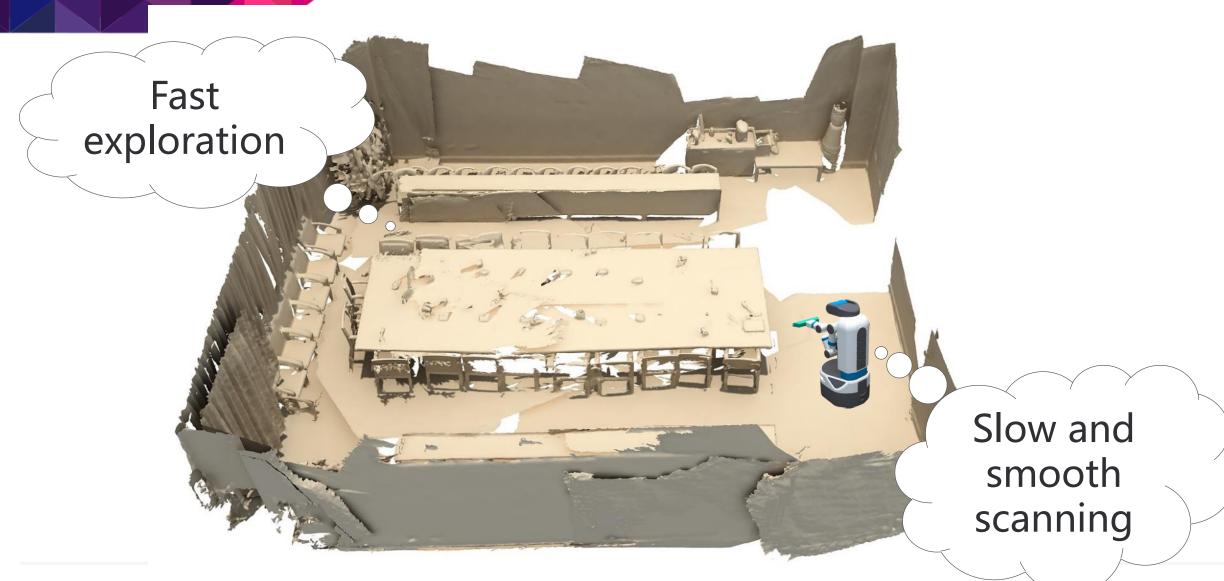
Never feel tired

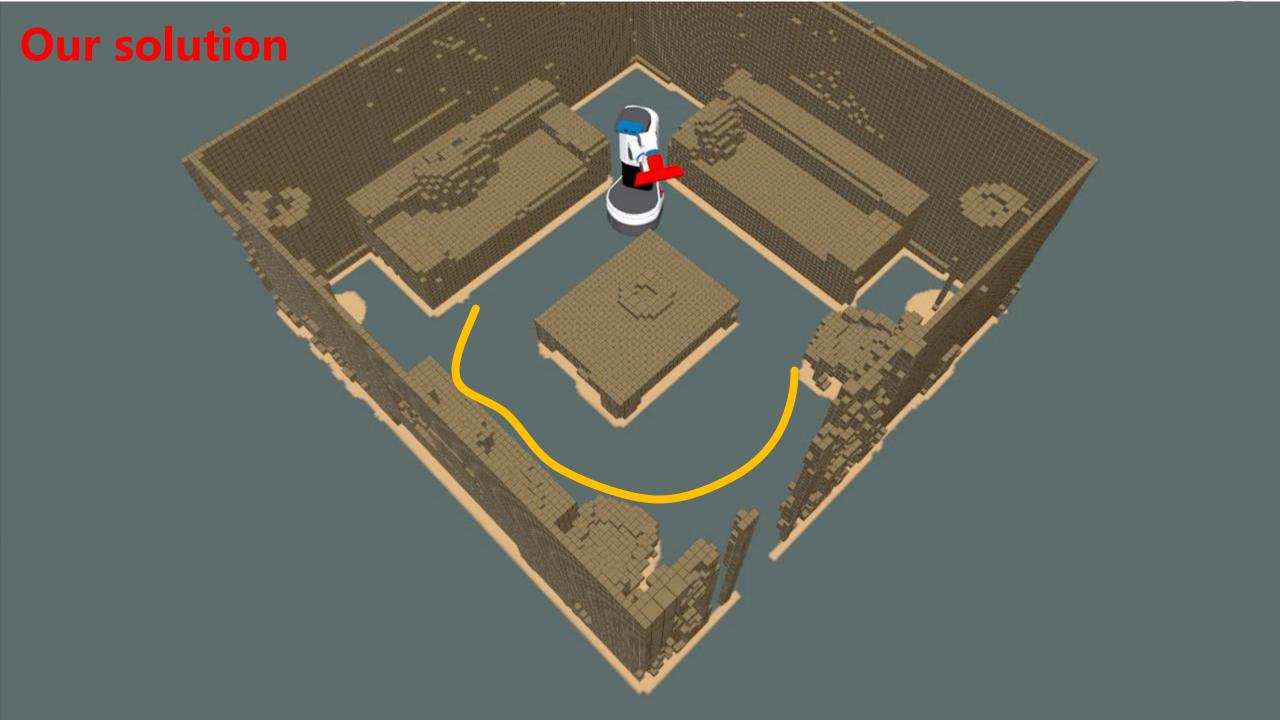
Automatic

Stable and accurate movement

### Difficulty of auto-scanning in unknown scenes



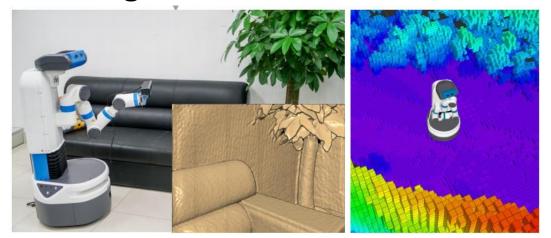




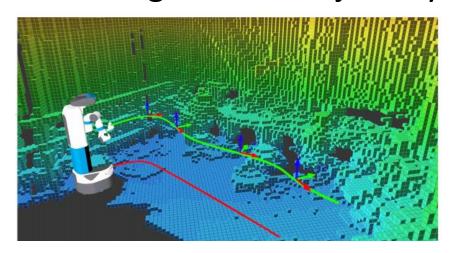
### **Pipeline**



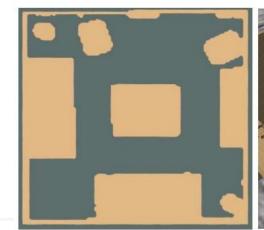
Scanning and online reconstruction

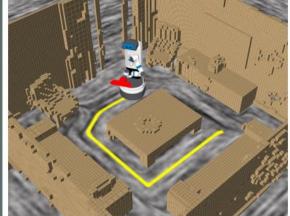


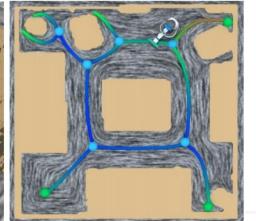
Estimating camera trajectory



**Local path advection Global path routing** 









# **Key techniques**



- Tensor field update
  - 2D tensor field
  - Time-varying tensor fields update
- Field guided path planning
  - Local path generation by particle advection
  - Global path finding by field topology
  - Field topology control
- Path-constrained camera trajectory estimation

# **Key techniques**

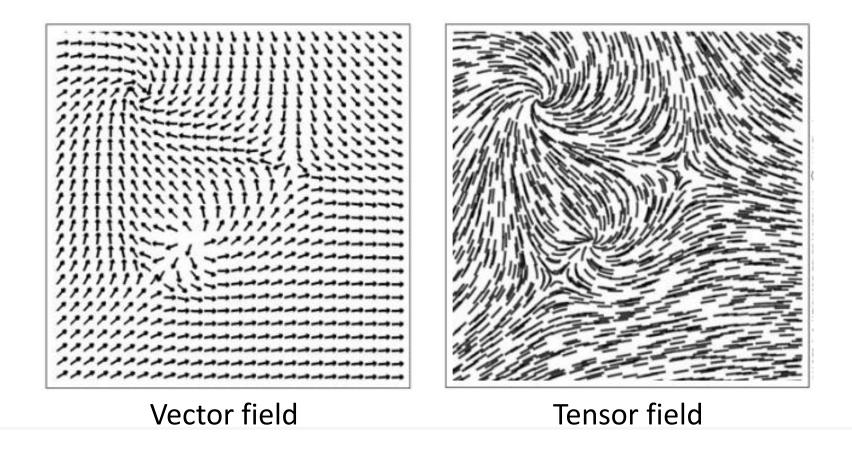


- Tensor field update
  - 2D tensor field
  - Time-varying tensor fields update
- Field guided path planning
  - Local path generation by particle advection
  - Global path finding by field topology
  - Field topology control
- Path-constrained camera trajectory estimation

#### **2D Tensor Field**



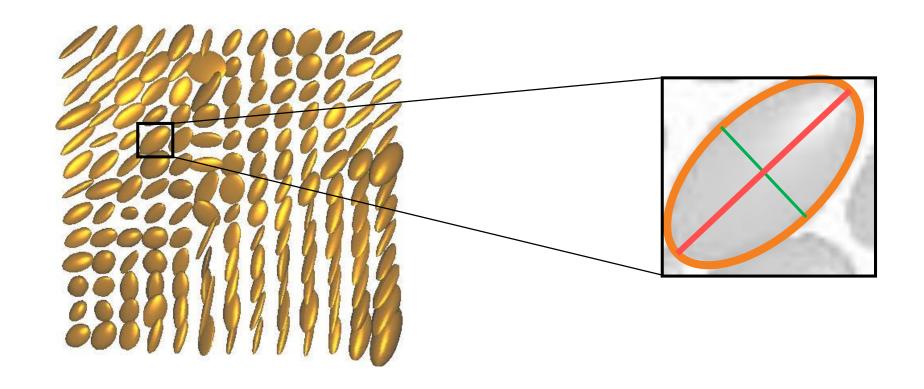
In a 2D domain, assign every point a direction, but **NOT** orientation



#### **2D Tensor Field**

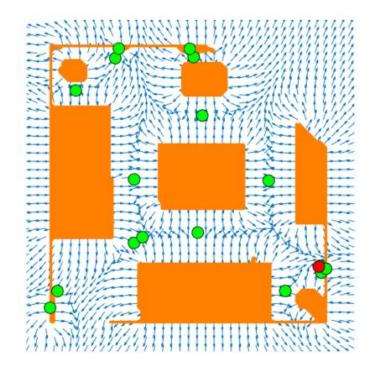


- Assign every point a tensor:  $T(p) = \begin{pmatrix} au_{11}(p) & au_{12}(p) \\ au_{21}(p) & au_{22}(p) \end{pmatrix}$ 



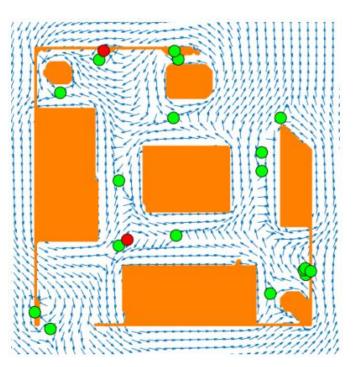


#### - Fewer singularities



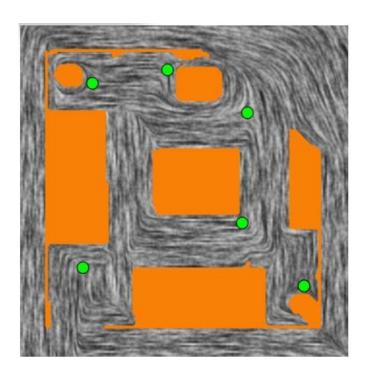
Potential field

[Khatib et al. 1986]



Gradient field

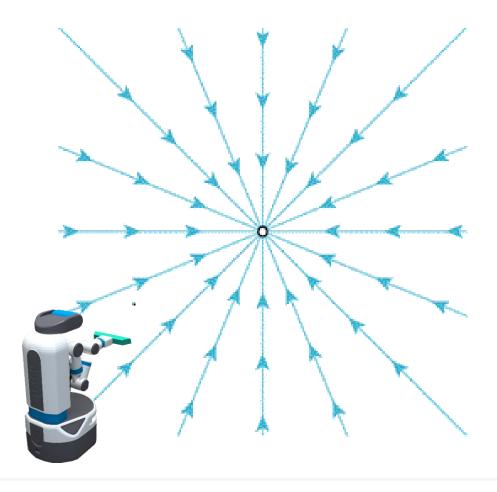
[Shade and Newman 2011]



Tensor field

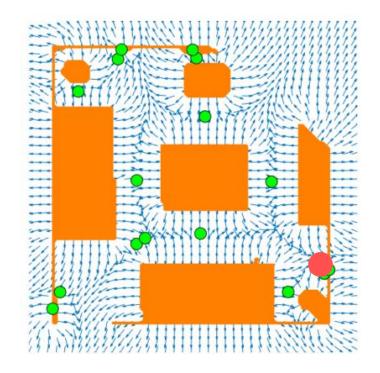


- Sink-free



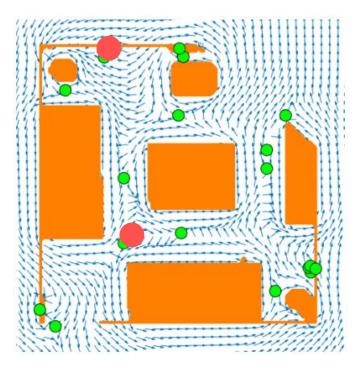


#### - Sink-free



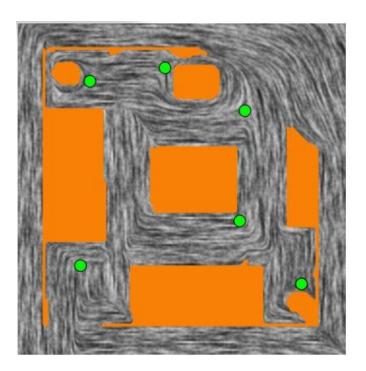
Potential field

[Khatib et al. 1986]



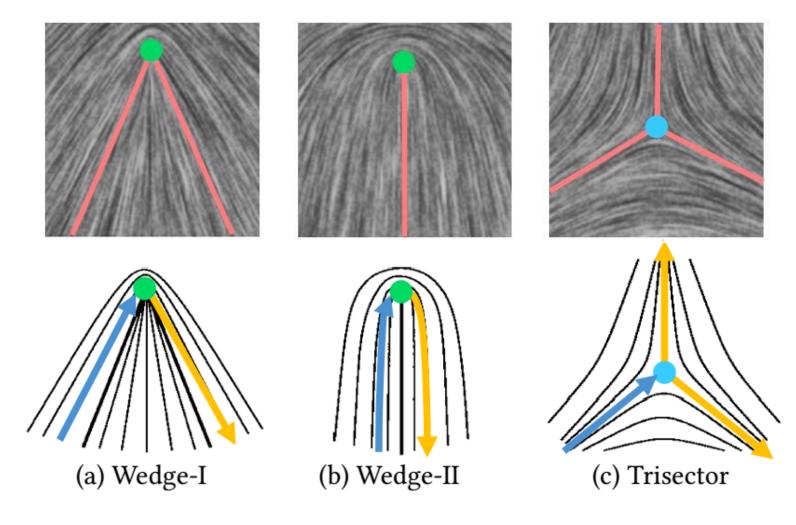
Gradient field

[Shade and Newman]



Tensor field

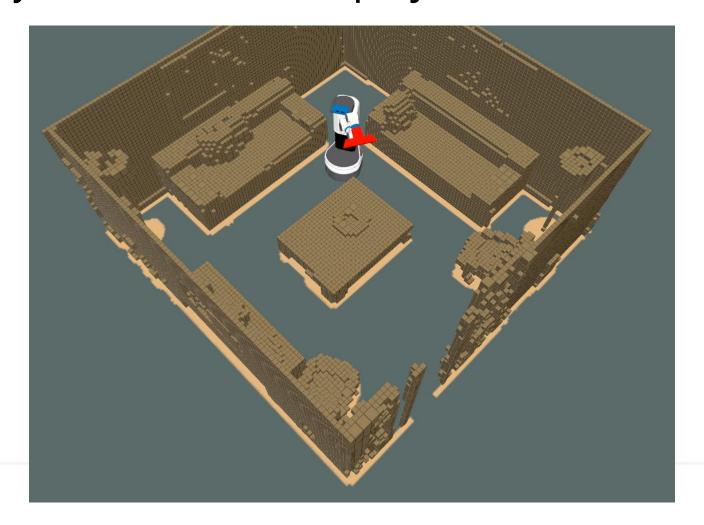




Tensor fields do have degenerate points

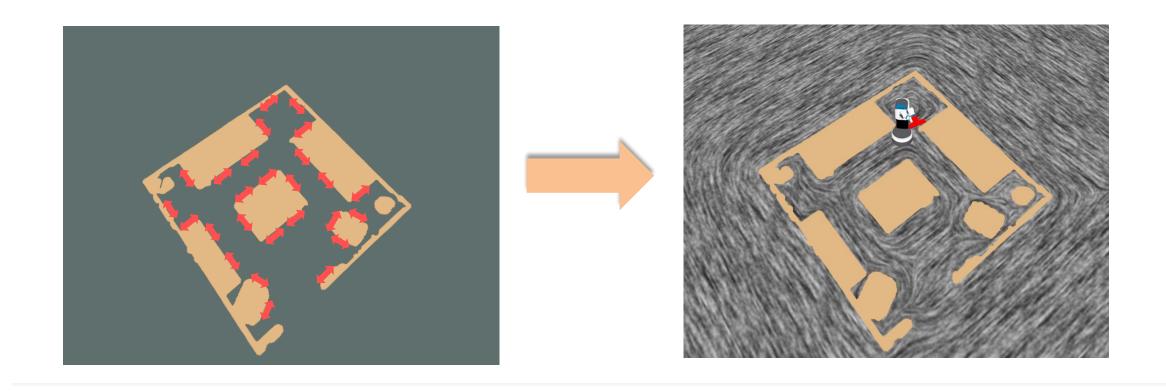


- The currently scanned scene is projected onto the floor plane

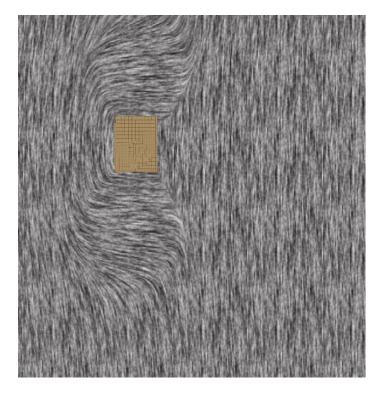




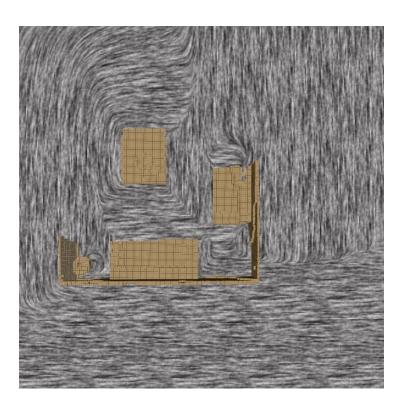
- Based on the tangential constraint of the 2D projection, a 2D tensor field is computed







A smooth transition from  $T^{t-1}$  to  $T^t$  ?



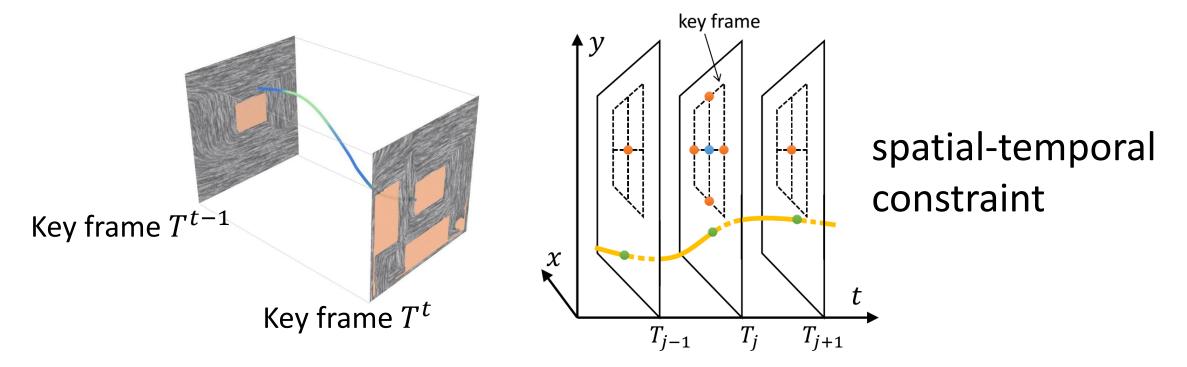
 $T^{t-1}$ 

Time-varying tensor fields

 $T^t$ 



Time-varying tensor fields update



Solve a spatial-temporal Laplacian system

# **Key techniques**



- Tensor field update
  - 2D tensor field
  - Time-varying tensor fields update
- Field guided path planning
  - Local path generation by particle advection
  - Global path finding by field topology
  - Field topology control
- Path-constrained camera trajectory estimation

# **Key techniques**

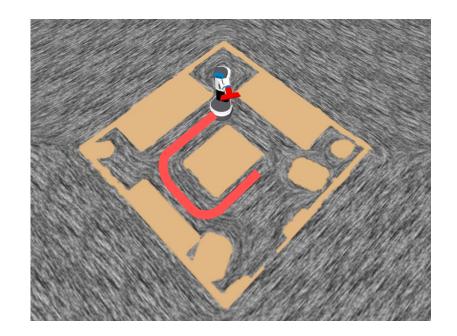


- Tensor field update
  - 2D tensor field
  - Time-varying tensor fields update
- Field guided path planning
  - Local path generation by particle advection
  - Global path finding by field topology
  - Field topology control
- Path-constrained camera trajectory estimation

#### **Local Path Generation**



Particle advection over tensor field



$$p(t) = p_0 + \int_0^t V(p(s); t_0 + t) ds$$

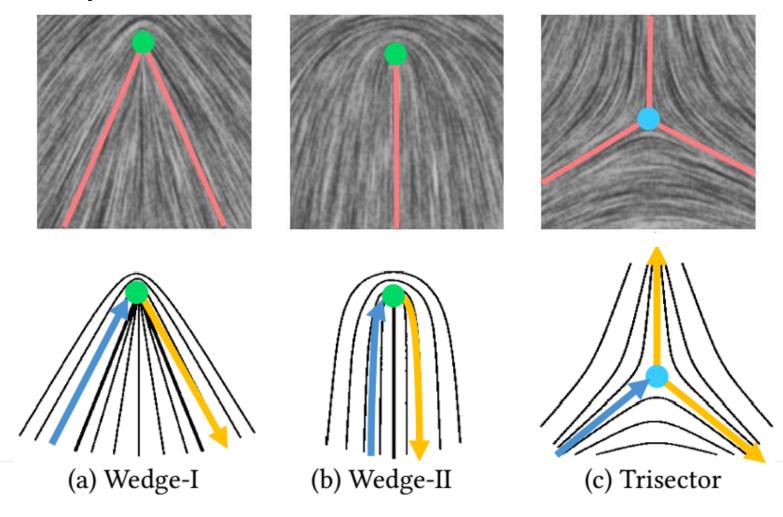
## **Key Points**



- Geometry-aware tensor field update
  - 2D tensor field
  - Time-varying tensor fields update
- Field guided path planning
  - Local path generation by particle advection
  - Global path finding by field topology
  - Field topology control
- Path-constrained camera trajectory estimation

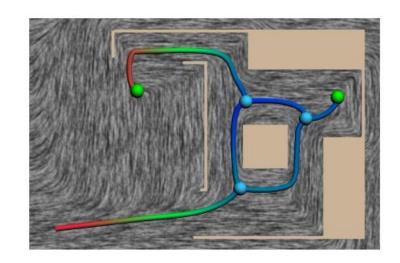


- Degenerate points

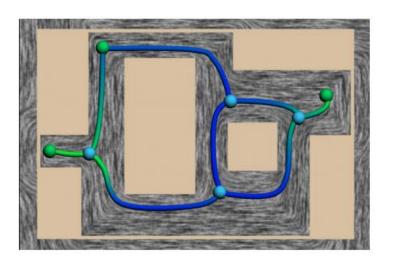




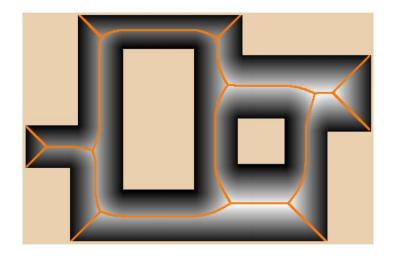
- Topological **graph** of tensor field
  - Node: Degenerate points
  - Edge: Separatrix lines connecting degen. points



For a partial scene



For the full scene

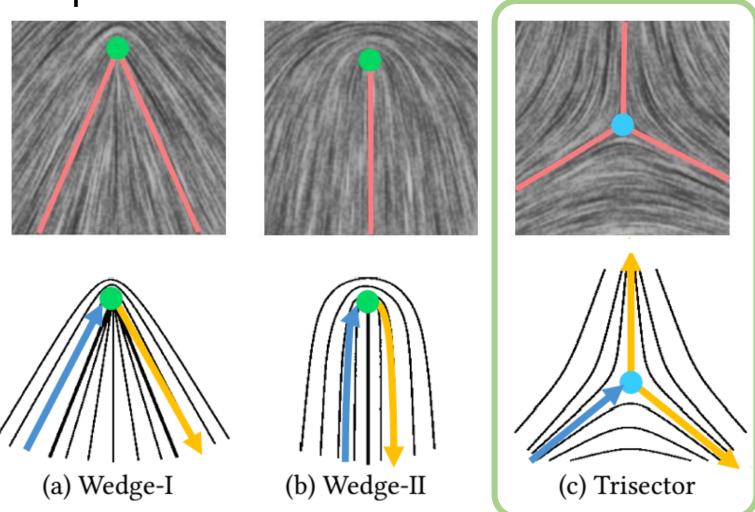


Medial axis

Robot path finding  $\rightarrow$  Finding paths over the field topo. graph!

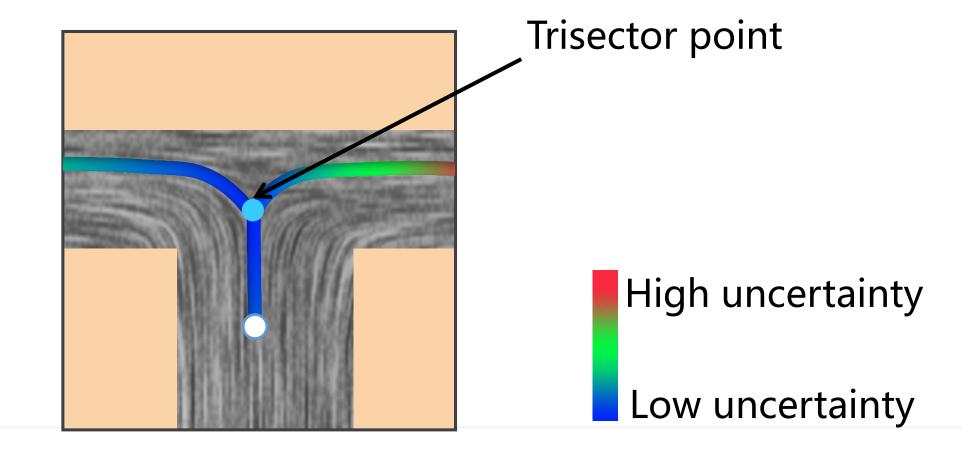


- Degenerate points



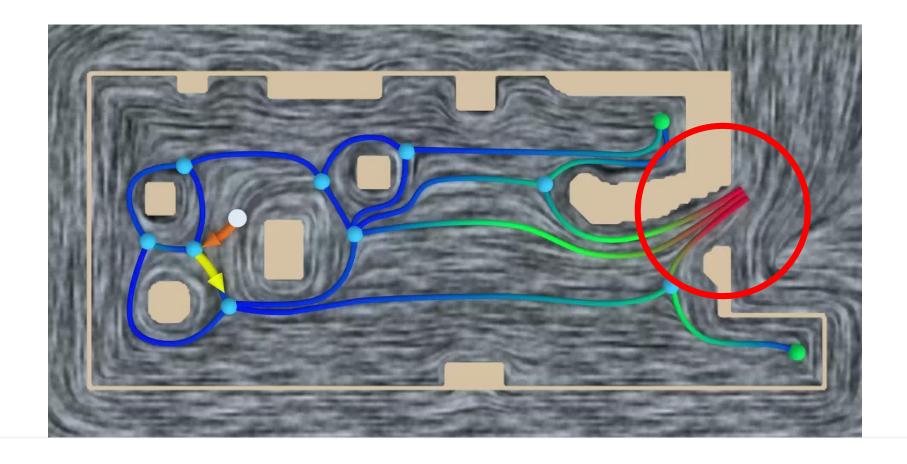


- How to select brunch at a trisector?



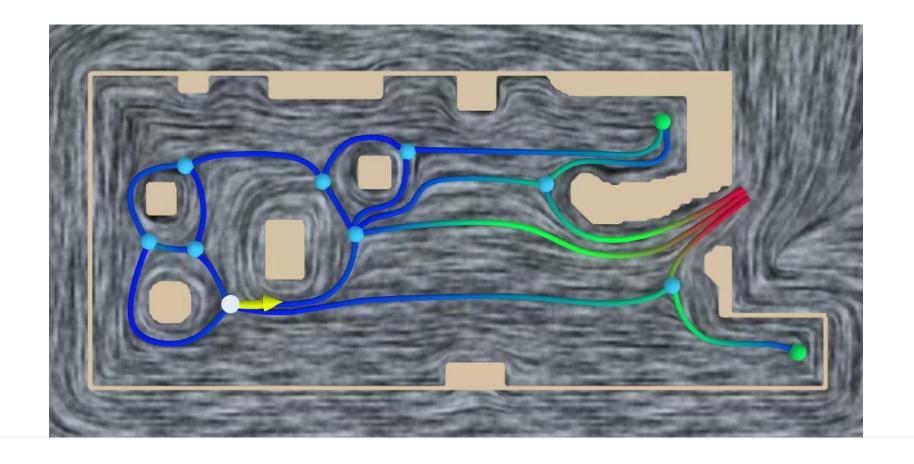


- Path routing with field topology



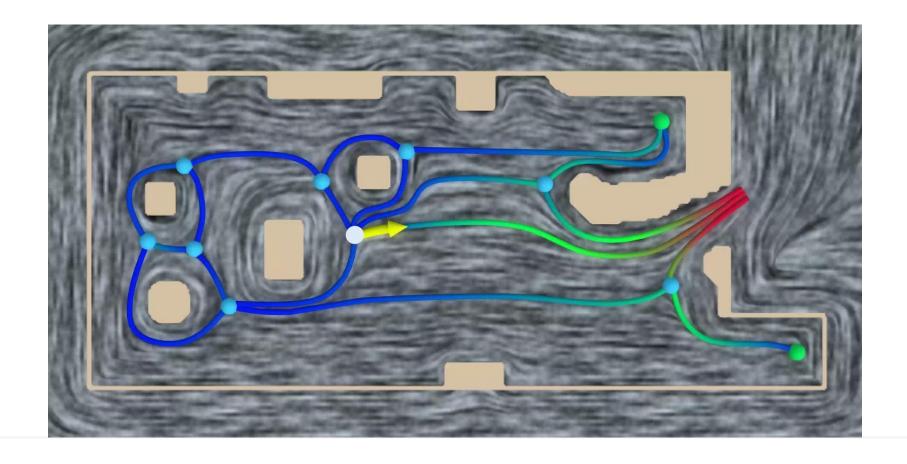


- Path routing with field topology





- Path routing with field topology



## **Key Points**

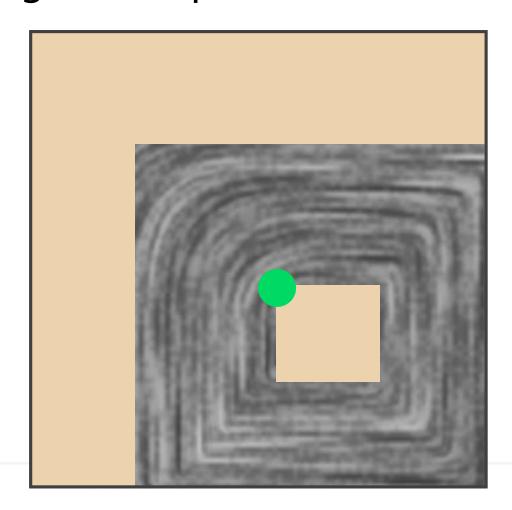


- Geometry-aware tensor field update
  - 2D tensor field
  - Time-varying tensor fields update
- Field guided path planning
  - Local path generation by particle advection
  - Global path finding by field topology
  - Field topology control
- Path-constrained camera trajectory estimation

# Field topology control



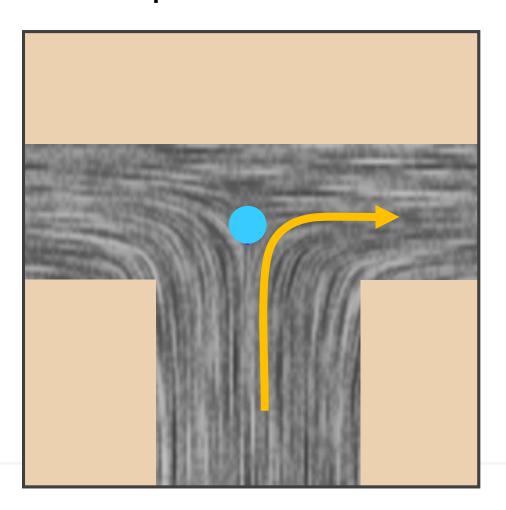
- Movement of a degenerate point



# Field topology control



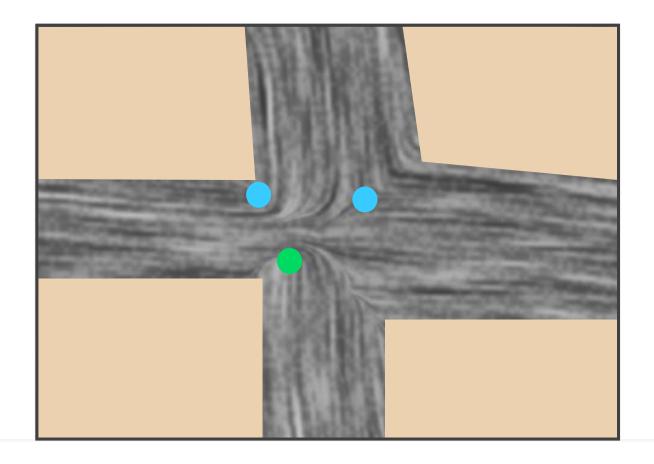
- Movement of degenerate points



# Field topology control



- Cancellation of degenerate pairs



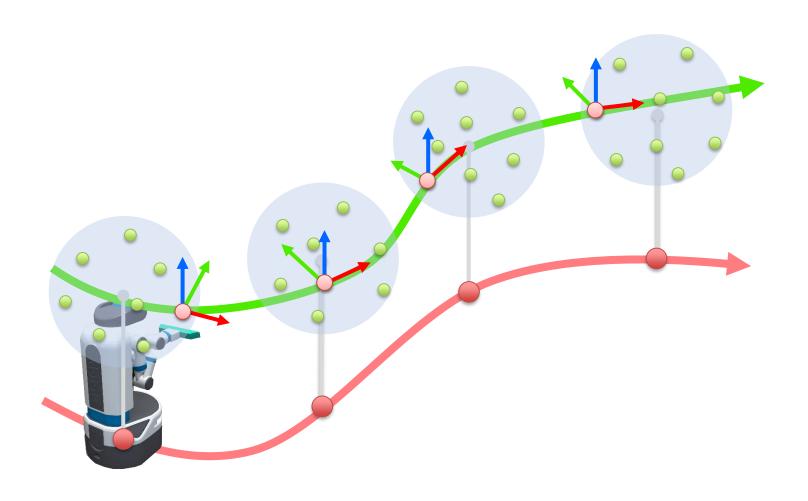
# **Key Points**



- Geometry-aware tensor field update
  - 2D tensor field
  - Time-varying tensor fields update
- Field guided path planning
  - Local path generation by particle advection
  - Global path finding by field topology
  - Field topology control
- Path-constrained camera trajectory estimation

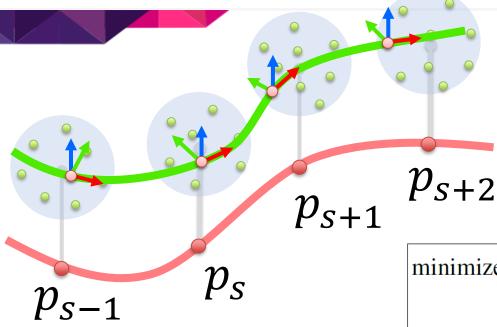
# **Camera Trajectory Optimization**





## **Camera Trajectory Optimization**





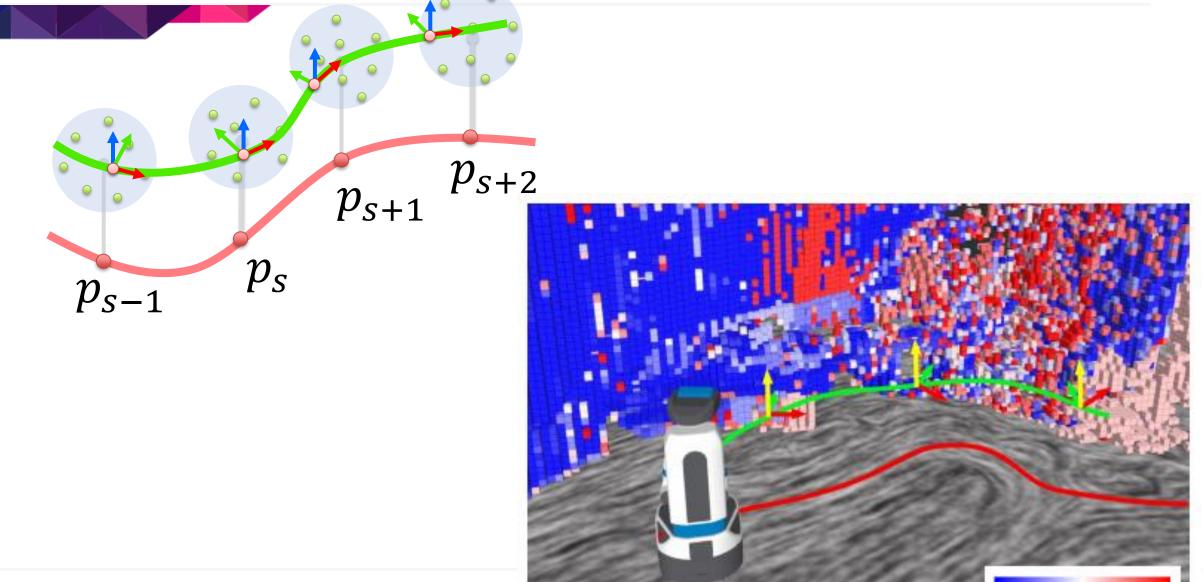
- Visibility to unknown
- Linear speed
- Angular speed

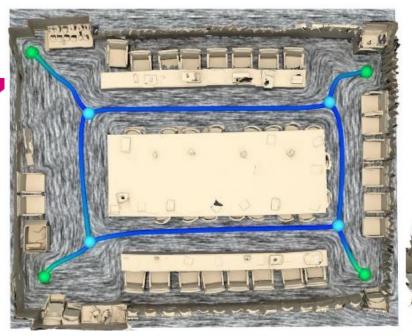
0-1 integer programming

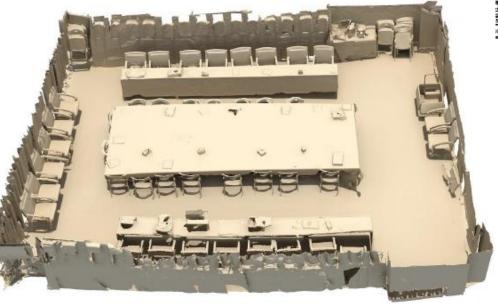
$$\begin{aligned} & \text{minimize} \quad E = - \ \omega_{\text{V}} \sum_{s} \sum_{i,j,k} V(q_{s}^{i}, a_{s}^{i}) x_{s}^{i} \\ & + \omega_{\text{L}} \sum_{s} \sum_{i,j,k} (q_{s-1}^{i} - 2q_{s}^{j} + q_{s+1}^{k}) x_{s-1}^{i} x_{s}^{j} x_{s+1}^{k} \\ & + \omega_{\text{A}} \sum_{s} \sum_{i,j,k} (a_{s-1}^{i} - 2a_{s}^{j} + a_{s+1}^{k}) x_{s-1}^{i} x_{s}^{j} x_{s+1}^{k}, \\ & \text{subject to} & (q_{s+1}^{i} - q_{s-1}^{j}) x_{s+1}^{i} x_{s-1}^{j} < 2v_{\text{m}}, \\ & (a_{s+1}^{i} - a_{s-1}^{j}) x_{s+1}^{i} x_{s-1}^{j} < 2a_{\text{m}}, \\ & x_{s}^{i} \in \{0,1\}, \quad \sum_{k} x_{s}^{k} = 1, \\ & s = 1, \dots, S-1, \quad i,j = 1, \dots, C. \end{aligned}$$

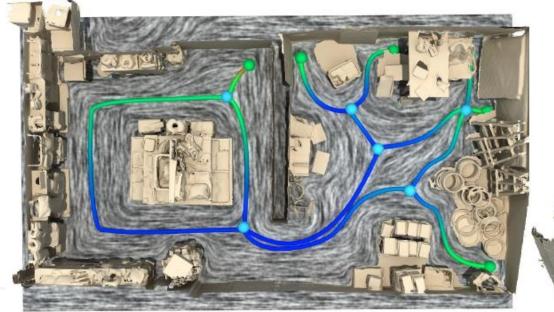
# **Camera Trajectory Optimization**

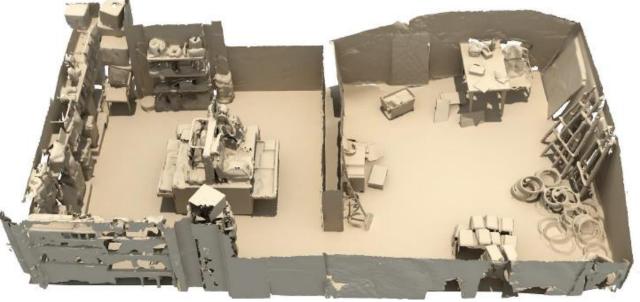




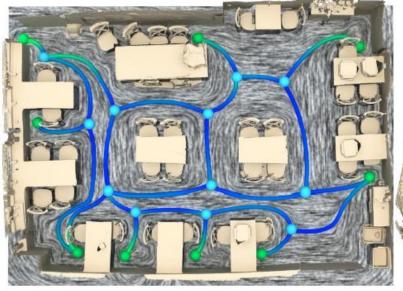


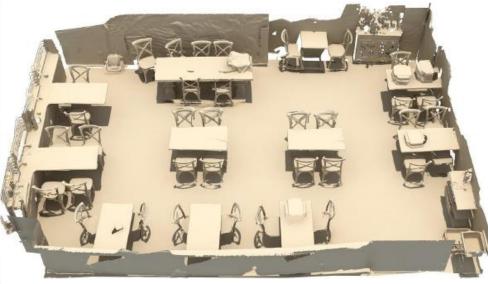


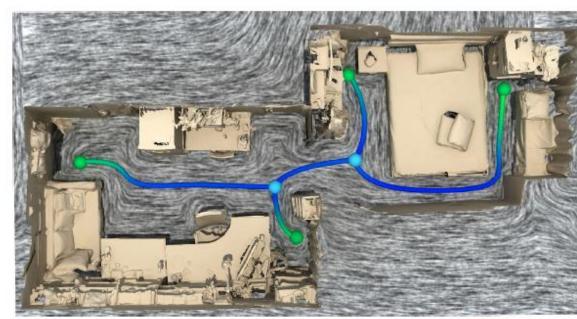


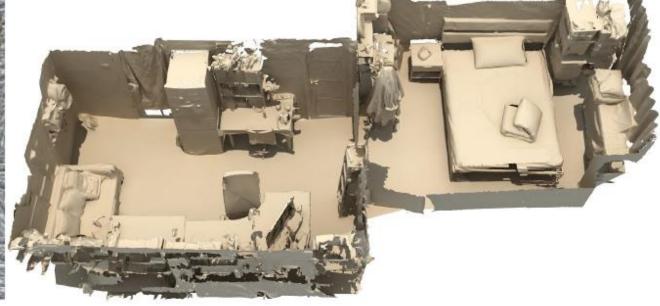






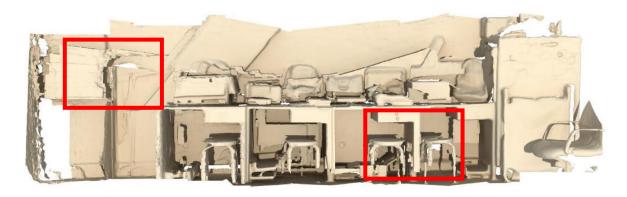








- Scanning quality



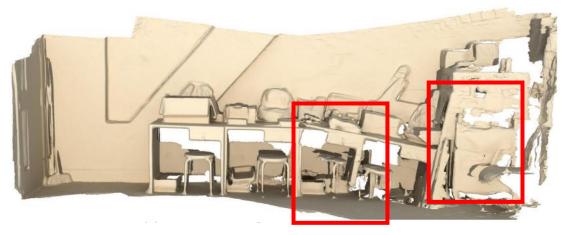
Scanned along potential field path



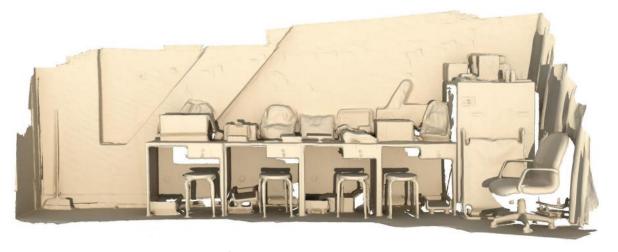
Scanned along **tensor** field path



- Scanning quality



Non-smooth camera trajectory

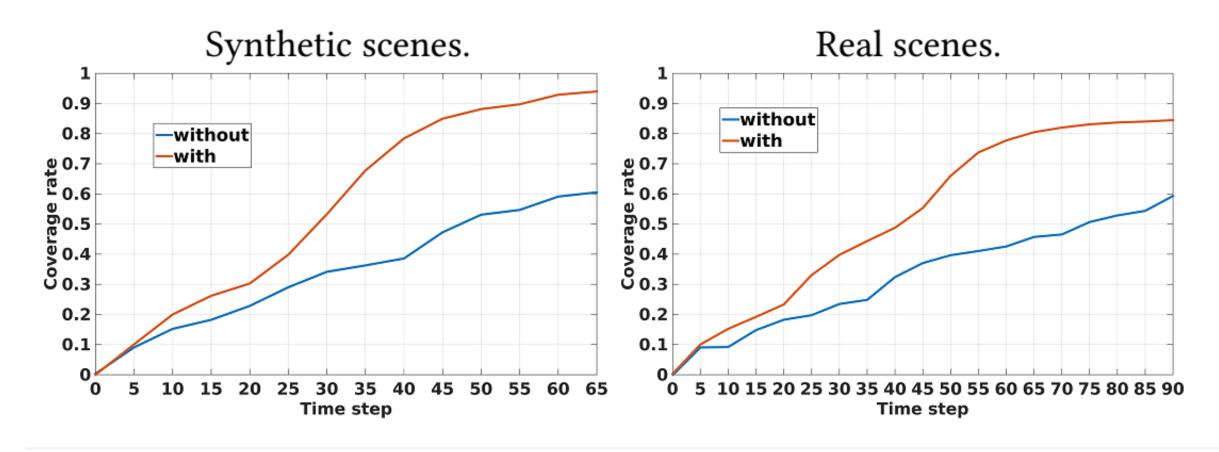


**Optimized** camera trajectory

#### **Evaluation**



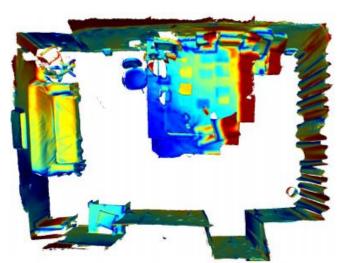
- Effect of global path planning



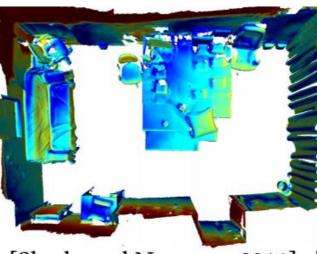
### Comparison

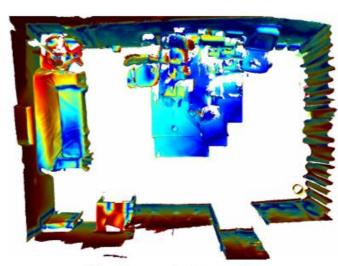


Pseudo-ground-truth

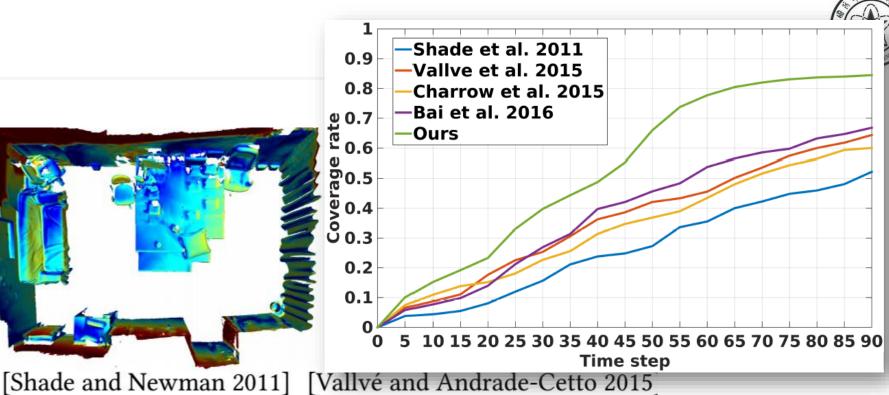


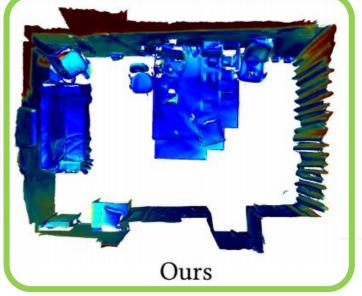
[Charrow et al. 2015]





[Bai et al. 2016]





#### **Future work**





Field guidance over **non-planar ground surfaces**, such as terrains?

#### **Future work**





Use **3D tensor fields** to guide robot grasping in complicated 3D environment?

# Thank you for attention!

Package available on ROS!

Try google:

"tensor field ros"



About | Support | Discussion Forum | Service Status | Q&A answers.ros.org

Documentation

**Browse Software** 

News

#### tensor\_field\_nav

Documentation Status

#### Contents

- 1. Introduction
- 2. Overview
- 3. Contributors and Publication
- Download and Insta
  - i. Dependencies
- From source
- 5. Run
- Acknowledgements

#### 1. Introduction

Tensor\_field\_nav package is developed for autonomous mapping of unknown indoor scenes by a mobile robot holding an RGBD camera. The key idea is to utilize 2D directional fields to guide robot movement. We compute and update a geometry-aware tensor field constrained by the currently reconstructed scene. The 3D scene geometry (i.e., the known surfaces) is projected to the floor plane. A set of 2D tangential constraints along the projected boundaries is extracted and used to compute/update the tensor field. The robot path is formed by particle advection over the tensor field, which is inherently obstacle avoiding.

During online scanning, the tensor field is updated in real-time, conforming to the incrementally reconstructed scene. To ensure a smooth robot path when advecting over the time-varying field, we propose a space-time optimization of tensor fields via imposing both spatial smoothness and temporal coherence. There are several important advantages of tensor field guided navigation. First, tensor fields are orientation-free and thus contain much less singularities (degenerate points), as compared to vector fields which are predominantly used in the literature. Fewer singularities lead not only to smoother path advection, which is critical for quality reconstruction, but also to more efficient navigation due to less ambiguity. In addition, tensor fields are sink-free, avoiding the issue of local trapping. Most importantly, the topological skeleton of a tensor field, comprised of all degenerate points and the separatrices connecting them, can be viewed as a routing graph. With this global structure, one can achieve global path planning for efficient scene scanning.