

# Towards Multi-rigid Body Localization

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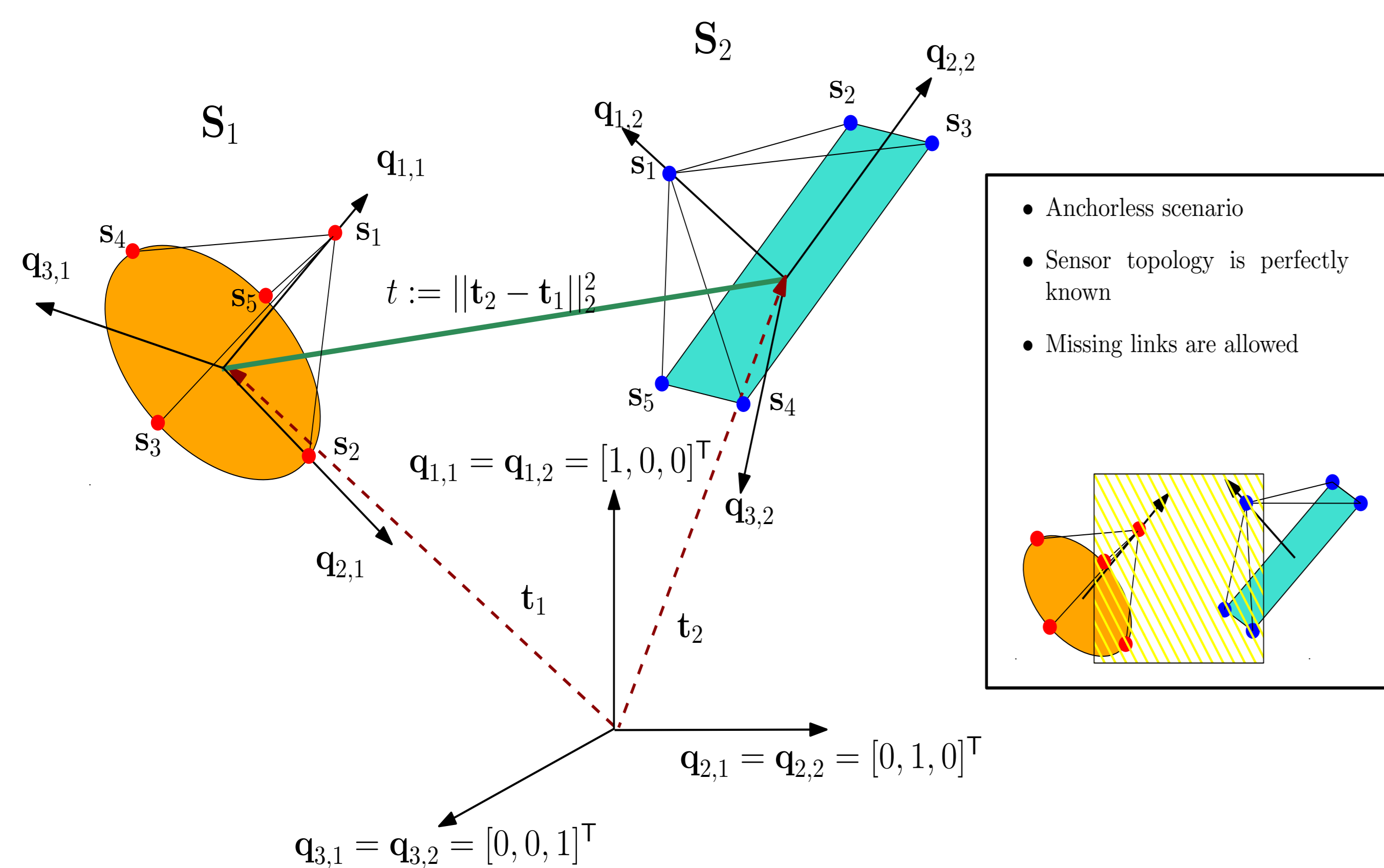
## Problem statement

- Jointly estimate the **relative position** and **orientation** of multiple rigid bodies in  $\mathbb{R}^3$ .
- Based on **range-only** measurements between the sensor pairs.
- **Without** any inertial measurements (e.g., accelerometers) and/or anchors.

## Applications

- Control and maneuvering of unmanned aircrafts, drones, underwater vehicles, satellites, and robotics.
- **Topology-aware** localization.

## Problem modeling



- $N$  sensors mounted on each rigid body.
- We **know** how the sensors are mounted on each rigid body (e.g., during fabrication).
- Absolute position of the sensors, or the rigid bodies itself is **not known**.

Absolute position of  $i$ th body (**rigid body transformation [1]**):

$$\mathbf{S}_i = \mathbf{Q}_i \mathbf{C}_i + \mathbf{t}_i \mathbf{1}_N^T \quad i = 1, 2.$$

$$\mathbf{S}_i = [s_{1,i}, \dots, s_{N,i}] \in \mathbb{R}^{3 \times N}$$

**Remarks:**

1.  $\mathbf{C}_i = [c_{1,i}, \dots, c_{N,i}]$  is the **known** coordinates of the  $i$ th body in the reference frame.
2.  $\mathbf{Q}_i$  is the **unknown** rotation matrix, i.e.,  $\mathbf{Q}_i^T \mathbf{Q}_i = \mathbf{I}$ . It tells how the  $i$ th rigid body is rotated in the reference frame.
3.  $\mathbf{t}_i \in \mathbb{R}^{3 \times 1}$  is the **unknown** translation of the  $i$ th body.

## Measurements

Squared range between cross-body sensors (noise free):

$$\mathbf{Y}^{\odot 2} = \boldsymbol{\psi}_1 \mathbf{1}_N^T + \mathbf{1}_N \boldsymbol{\psi}_2^T - 2\mathbf{S}_1^T \mathbf{S}_2$$

$$\boldsymbol{\psi}_i = [\|s_{1,i}\|^2, \dots, \|s_{N,i}\|^2]^T \in \mathbb{R}^{N \times 1} \text{ and } [\mathbf{Y}]_{m,n} = \|s_{m,i} - s_{n,i}\|^2, m, n = 1, \dots, N \quad i = 1, 2.$$

## Proposed estimators

For relative localization (imagine multidimensional scaling):

1. Estimate the relative rotation matrix  $\mathbf{Q} = \mathbf{Q}_1^T \mathbf{Q}_2 \in \mathcal{V}_{3,3}$ , where
 
$$\mathcal{V}_{3,3} = \{\mathbf{Q} \in \mathbb{R}^{3 \times 3} : \mathbf{Q}^T \mathbf{Q} = \mathbf{I}_3\},$$
2. Estimate the relative translation  $\mathbf{t} = \|\mathbf{t}_2 - \mathbf{t}_1\|^2 \in \mathbf{R}$ .

## Relative rotation estimator

**Step 1:** Project out the known vectors  $\mathbf{1}_N^T$  and  $\mathbf{1}_N$  from  $\mathbf{Y}$

$$\tilde{\mathbf{Y}} = -1/2 \boldsymbol{\Gamma}_N \mathbf{Y}^{\odot 2} \boldsymbol{\Gamma}_N = \tilde{\mathbf{S}}_1^T \tilde{\mathbf{S}}_2 = \mathbf{C}_1^T \mathbf{Q} \mathbf{C}_2$$

$$\tilde{s}_i = s_i \boldsymbol{\Gamma}_N \text{ with projection matrix } \boldsymbol{\Gamma}_N$$

**Step 2:** Orthogonal Procrustes Problem (OPP) [2]:

$$\hat{\mathbf{Q}} = \underset{\mathbf{Q} \in \mathcal{V}_{3 \times 3}}{\operatorname{argmin}} \|\tilde{\mathbf{Y}} - \mathbf{C}_1^T \mathbf{Q}\|_F^2 = \mathbf{U} \mathbf{V}^T$$

$$\tilde{\mathbf{Y}} = \tilde{\mathbf{Y}} \mathbf{C}_2^\dagger \text{ and } \mathbf{C}_1 \tilde{\mathbf{Y}} =: \mathbf{U} \boldsymbol{\Sigma} \mathbf{V}^T$$

## Relative translation estimator

We can express range in terms of relative translation as:

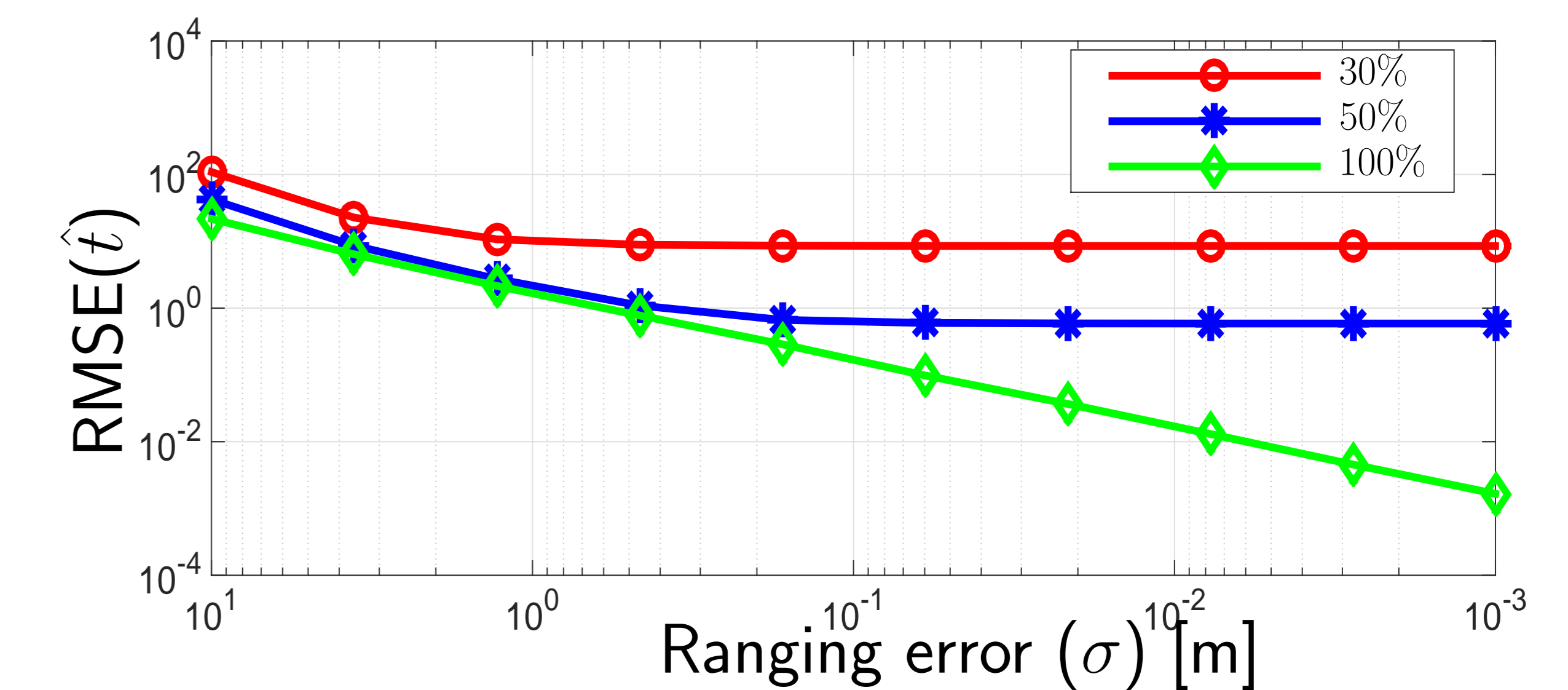
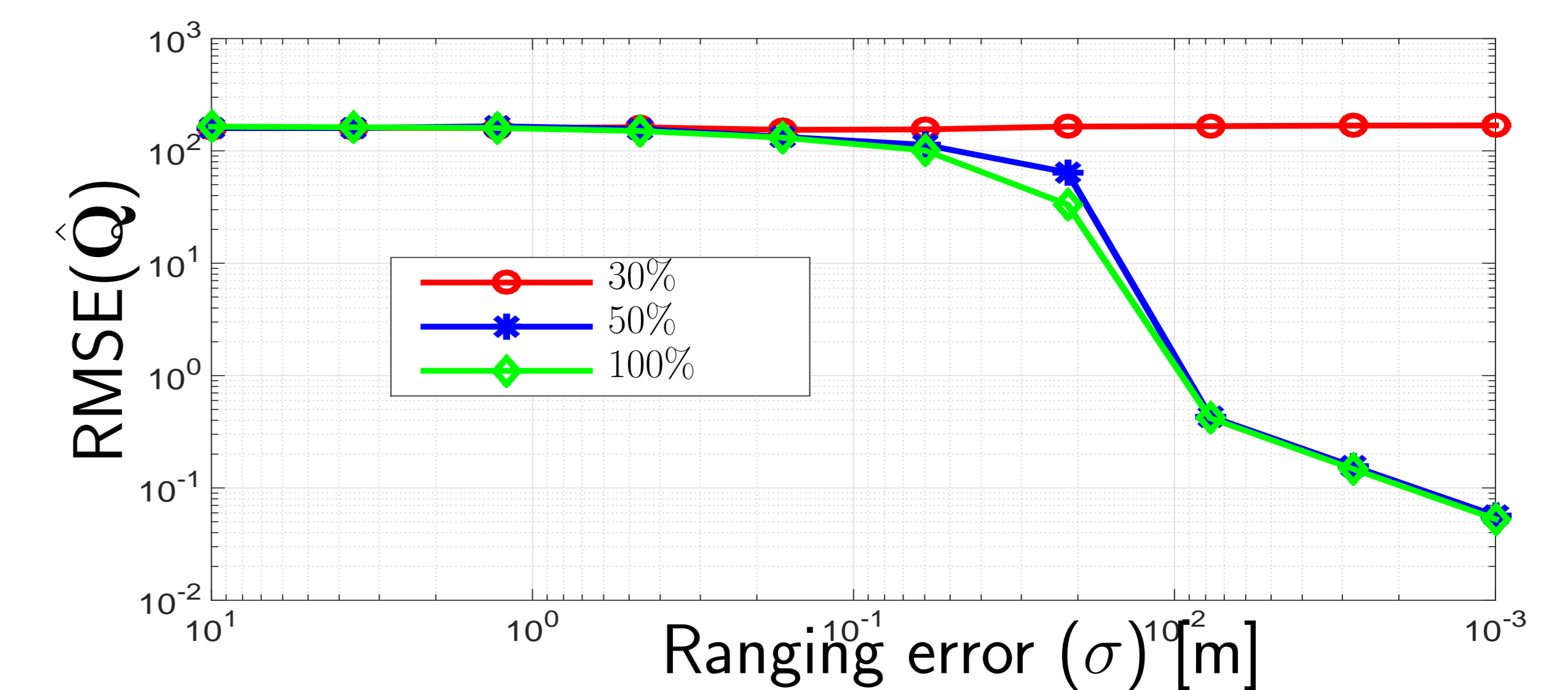
$$\|\mathbf{Y}\|_F^2 = N \sum_{n=1}^N (\|c_{n,1}\|_2^2 + \|c_{n,2}\|_2^2) + N^2 \underbrace{\|\mathbf{t}_1 - \mathbf{t}_2\|_2^2}_{\text{relative translation}}$$

Thus

$$\hat{t} = \frac{1}{N^2} \|\mathbf{Y}\|_F^2 - \frac{1}{N} \left( \sum_{n=1}^N (\|c_{n,1}\|_2^2 + \|c_{n,2}\|_2^2) \right)$$

## Simulation results

**Set-up:**  $N = 10$ ,  $\{\psi_1, \theta_1, \phi_1\} = \{20^\circ, -25^\circ, 30^\circ\}$  and  $\{\psi_2, \theta_2, \phi_2\} = \{40^\circ, 135^\circ, -75^\circ\}$ ,  $\mathbf{t}_1 = [1, -5, 4]^T \text{ m}$ ,  $\mathbf{t}_2 = [-3, 1, 7]^T \text{ m}$ , and  $10^4$  Monte-Carlo experiments.



[1] S. P. Chepuri, G. Leus, and A.-J. van der Veen, "Rigid Body Localization Using Sensor Networks," *IEEE Trans. Sig. Process.*, 2014.

[2] G.H. Golub and C.F. van Loan, *Matrix Computations*, JHU Press, 1996