

Microphone Array Position Self-Calibration from Reverberant Speech Input

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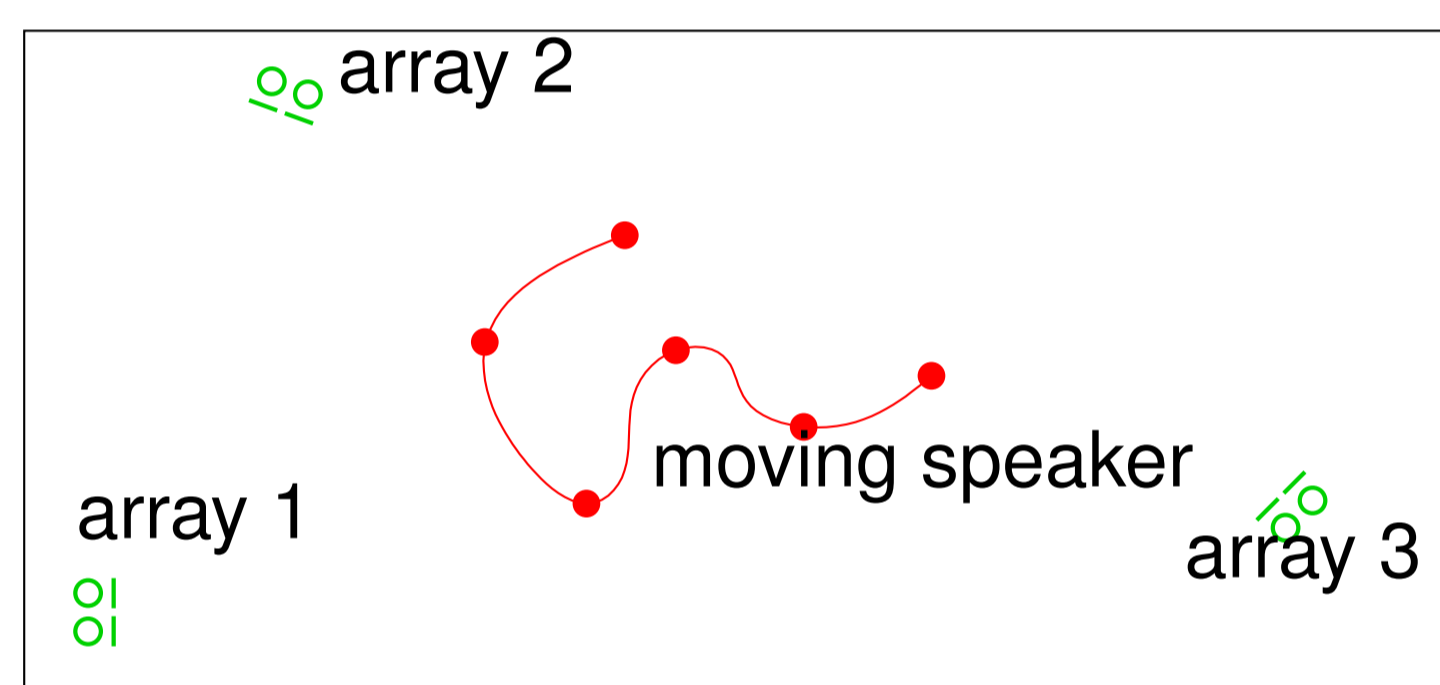
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Introduction

- The geometry of an acoustic sensor network is required for many signal processing applications
- Automatic sensor position estimation preferable to error-prone manual measurement process
- Existing approaches often use artificial calibration signals or special hardware to achieve high positioning accuracy
- Goal: Relative geometry calibration based on reverberant speech input

Problem statement (2D)

- Each sensor node consists of a microphone array
- Array configuration within sensor node known
- Measurements: Direction of Arrival (DoA) from each sensor node
- Unknown parameters:
 - ▶ Sensor positions: $[x_j^S, y_j^S]$
 - ▶ Sensor orientations: Θ_j
 - ▶ Speaker positions: $[x_i^P, y_i^P]$



Proposed cost function

- ϕ_{ij} : DoA of i -th speaker position measured by j -th sensor:

$$\mathbf{v}_{ij} = [\cos(\phi_{ij}) \sin(\phi_{ij})]^T$$

- DoA vector predicted by current geometry estimates:

$$\tilde{\mathbf{v}}_{ij} = \begin{bmatrix} \cos(\tilde{\phi}_{ij} - \hat{\theta}_j) \\ \sin(\tilde{\phi}_{ij} - \hat{\theta}_j) \end{bmatrix} = \underbrace{\begin{bmatrix} \cos(\hat{\theta}_j) & \sin(\hat{\theta}_j) \\ -\sin(\hat{\theta}_j) & \cos(\hat{\theta}_j) \end{bmatrix}}_{\mathbf{R}(-\hat{\theta}_j)} \underbrace{\frac{1}{\sqrt{\Delta x_{ij}^2 + \Delta y_{ij}^2}}}_{1/|\hat{\mathbf{v}}_{ij}|} \underbrace{\begin{bmatrix} \Delta x_{ij} \\ \Delta y_{ij} \end{bmatrix}}_{\hat{\mathbf{v}}_{ij}}$$

- Cost function:

$$J_C = \sum_{i=1}^N \sum_{j=1}^K \left\{ |\hat{\mathbf{v}}_{ij}| \left[1 - \cos(\angle(\mathbf{v}_{ij}, \tilde{\mathbf{v}}_{ij})) \right] \right\}^2$$

- Iterative cost function minimization using Newton's method gives estimates: $[\hat{x}_i^S, \hat{y}_i^S, \hat{\theta}_j, \hat{x}_i^P, \hat{y}_i^P]$

Previous cost function

- Geometric relation between sensor and observation:

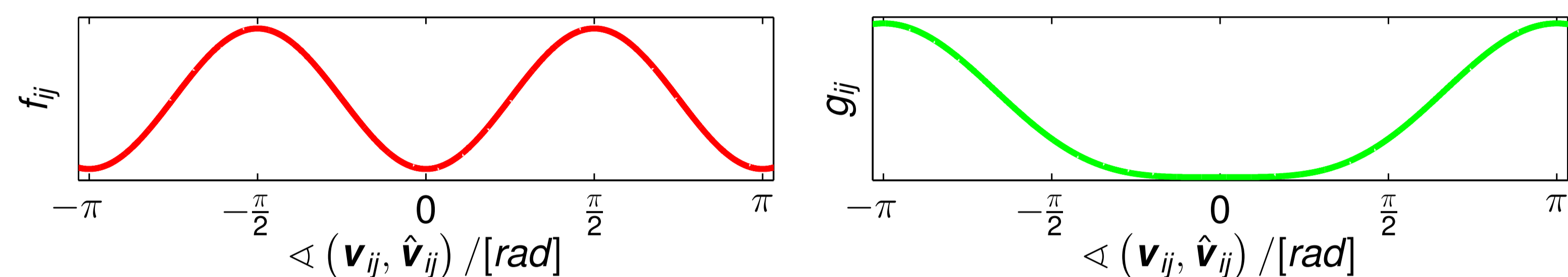
$$\tan(\hat{\theta}_j + \phi_{ij}) = \frac{\sin(\hat{\theta}_j + \phi_{ij})}{\cos(\hat{\theta}_j + \phi_{ij})} = \frac{\Delta y_{ij}}{\Delta x_{ij}}$$

- Resulting cost function

$$J_S = \sum_{i=1}^N \sum_{j=1}^K \left\{ |\hat{\mathbf{v}}_{ij}| \sin(\angle(\mathbf{v}_{ij}, \hat{\mathbf{v}}_{ij})) \right\}^2$$

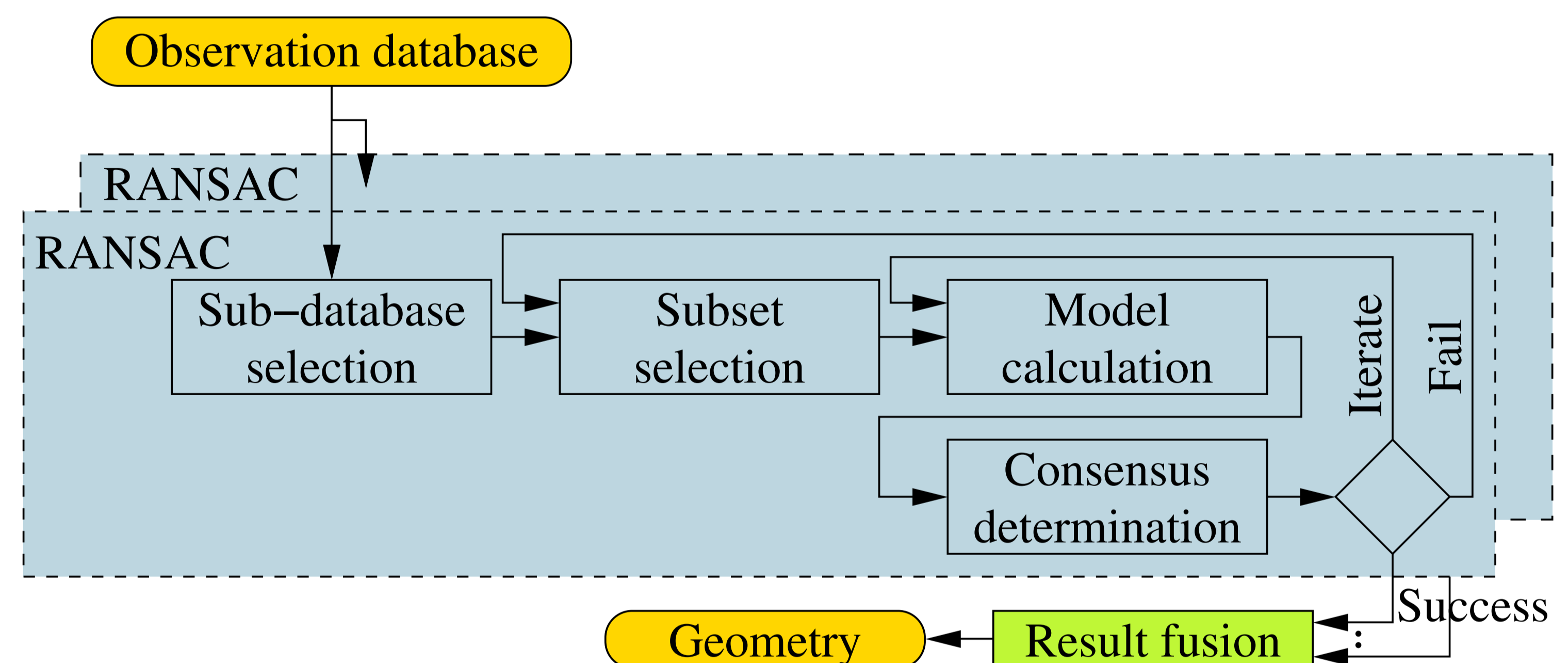
Comparison

- **Previous cost function:** Local minima, that correspond to wrong sensor orientations
- **Proposed cost function:** Avoids local minima, that correspond to wrong sensor orientations



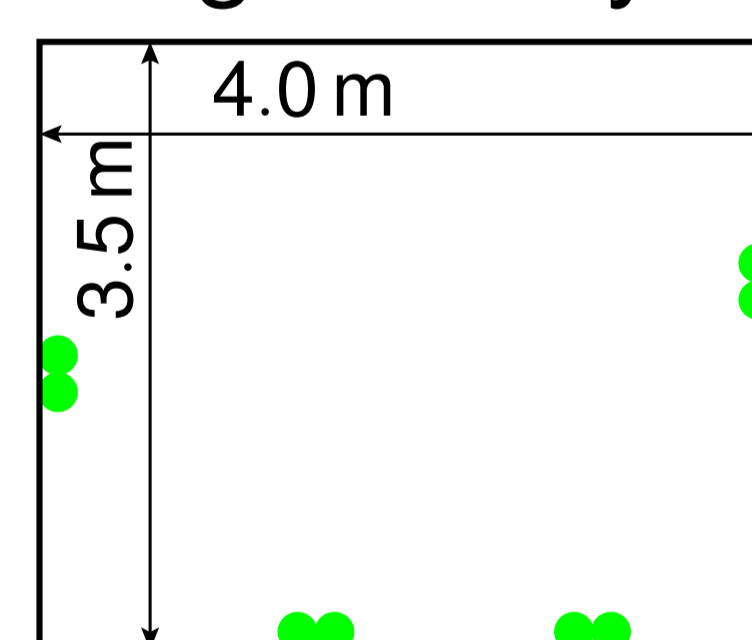
Random Sample Consensus (RANSAC)

- Precision of automatic geometry calibration highly depends on the quality of the DoA estimates
- Calibration embedded into RANSAC for outlier rejection

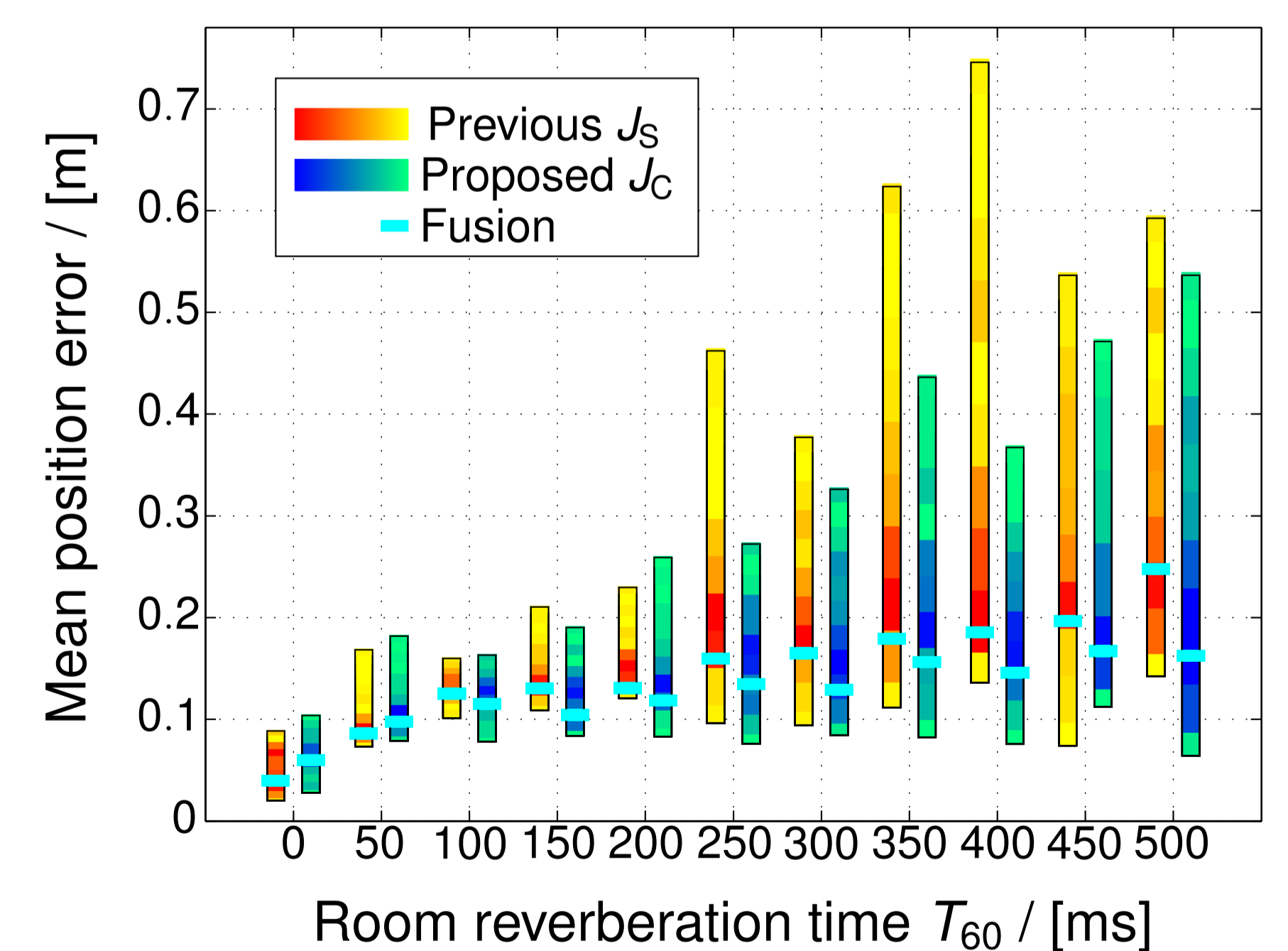


Experiments

- Room geometry:



- Simulated audio database, based on image method
- Reverberation times from 0 ms up to 500 ms



Comparison of the mean positioning error between the existing cost function and the proposed cost function for different reverberation times.

Conclusions

- New formulation avoids solutions that correspond to mirrored sensor orientations
- RANSAC increases robustness against reverberation
- Mean positioning error smaller than 0.25 m for reverberation times up to 500 ms