

Computational Enhancements for Certifiably Correct SLAM

David M. Rosen¹ Luca Carlone²

¹Oculus Research

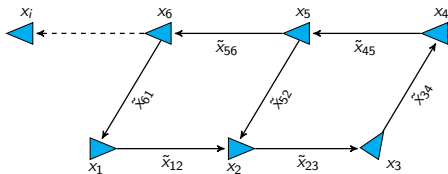
²Massachusetts Institute of Technology

Workshop on Introspective Methods for Reliable Autonomy
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A lightning review of pose-graph SLAM

Given:

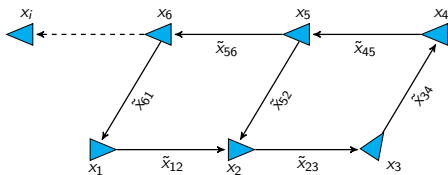
- Unknown poses $x_1, \dots, x_n \in SE(d)$
- A network of noisy relative measurements $\tilde{x}_{ij} \sim p_{ij}(\cdot | x_i^{-1} x_j)$:



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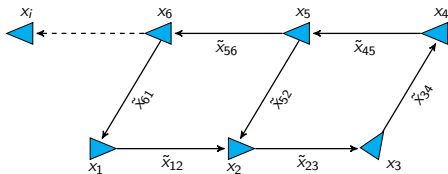
Find: The *maximum-likelihood estimate* $\hat{x} = (\hat{x}_1, \dots, \hat{x}_n) \in SE(d)$:

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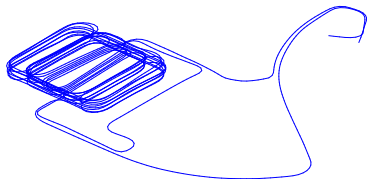


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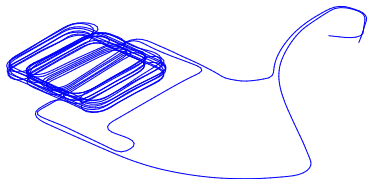
This is a *high-dimensional, nonconvex* optimization problem

The Problem: Local minima

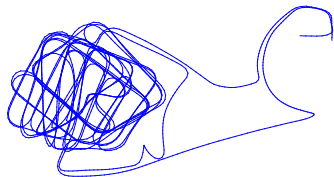


Correct (optimal) estimate

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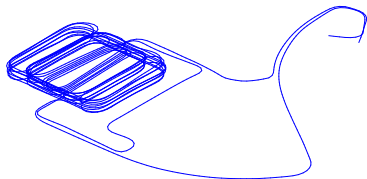


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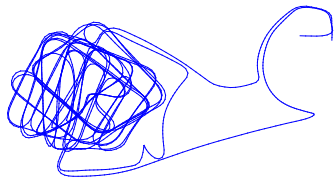


Suboptimal critical point

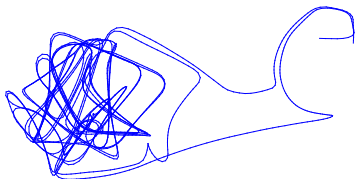
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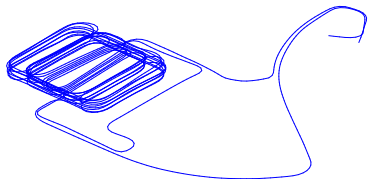


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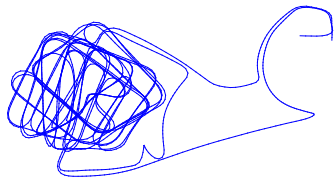


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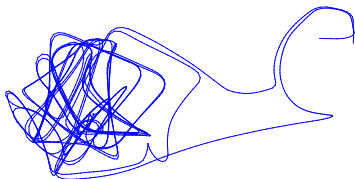
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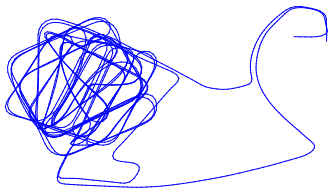
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- Directly searches for *global optima* via *convex relaxation*
- Generates *certificates of correctness* for global optima

¹D.M. Rosen et al. "A Certifiably Correct Algorithm for Synchronization over the Special Euclidean Group". In: *Intl. Workshop on the Algorithmic Foundations of Robotics (WAFR)*. San Francisco, CA, Dec. 2016 (Best Paper)

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However: These computations can be expensive

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In this work:

- *Computational enhancements* for accelerating SE-Sync:
 - Riemannian optimization
 - Truncated-Newton methods
 - Structure-exploiting matrix decompositions
 - Krylov-subspace linear algebra techniques
- *First C++ implementation* of SE-Sync
- *Experimental evaluation*: SE-Sync $\approx 2\times$ faster than GTSAM

Code: <https://github.com/david-m-rosen/SE-Sync>