

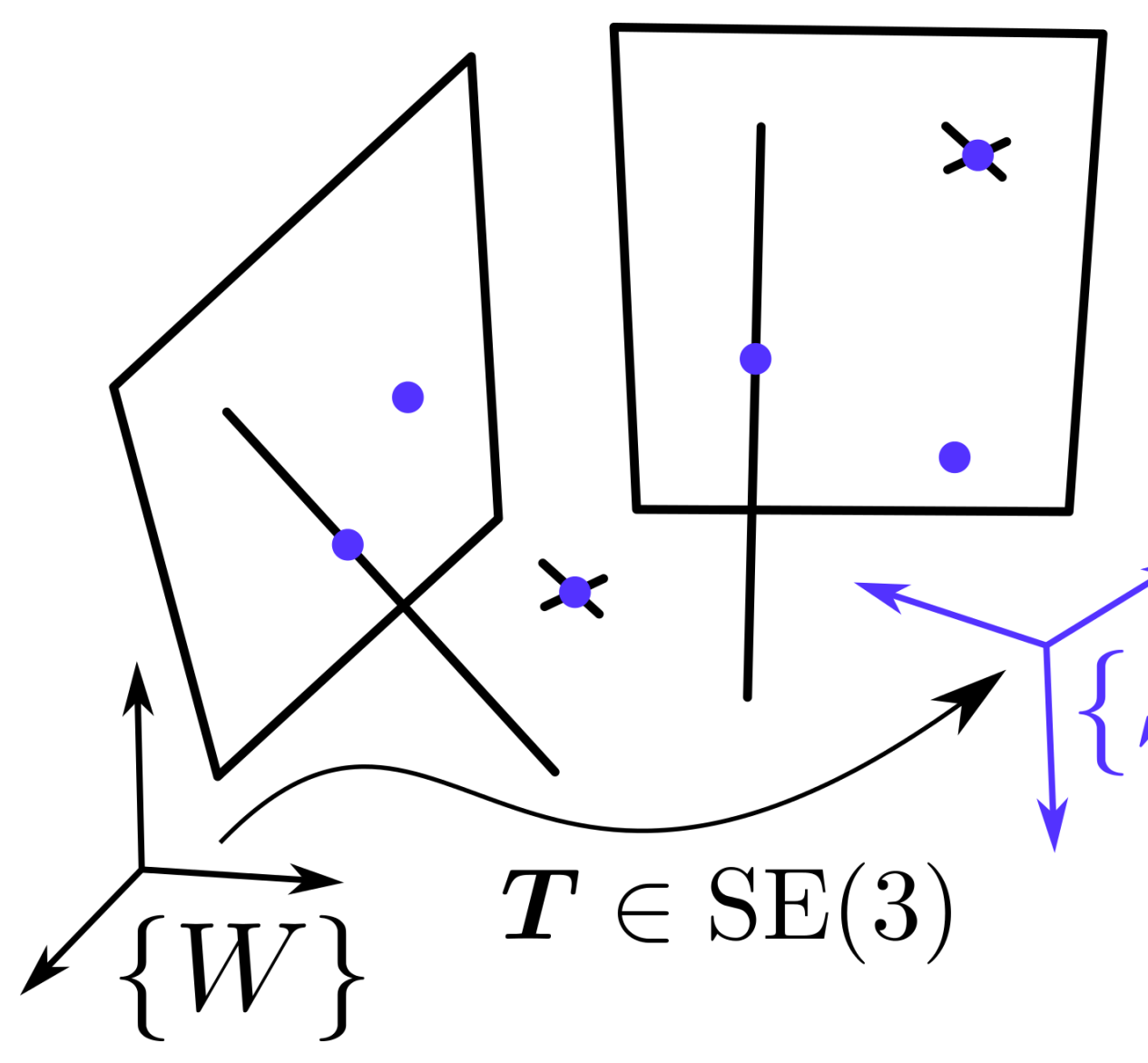
The problem: Rigid registration of points, lines and planes

Given point-to-any correspondences:

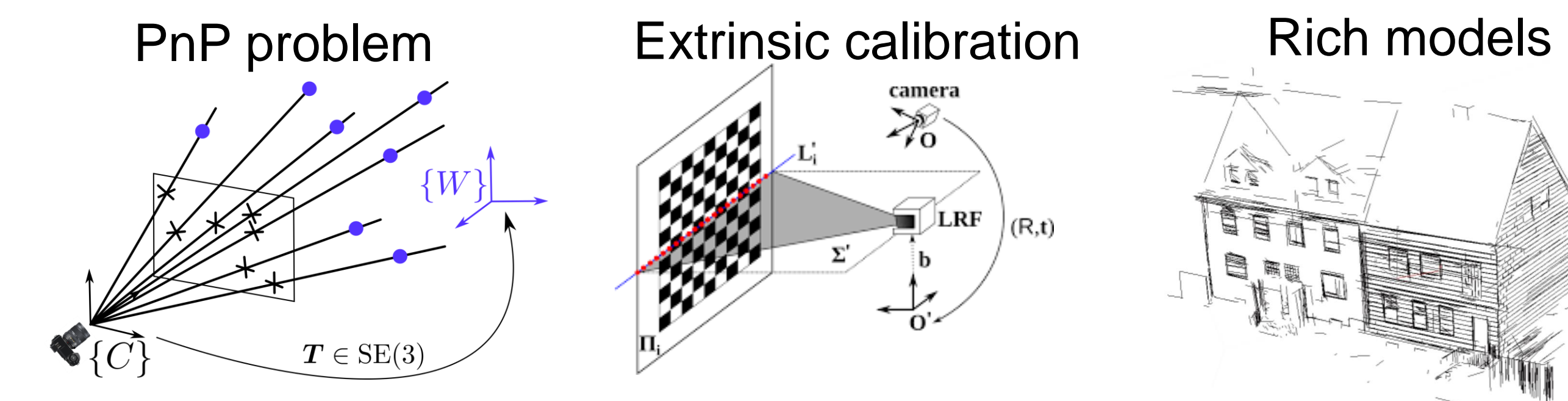
$$T^* = \arg \min_{T \in SE(3)} \sum_{i=1}^m \min_{y_i \in P_i} \|T \oplus x_i - y_i\|_2^2$$

Unknowns:

$$T = [R|t] \in SE(3) \equiv SO(3) \times \mathbb{R}^3$$

$$R = [c_1|c_2|c_3] = \begin{bmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{bmatrix}$$


Examples: Lines and planes are pervasive in Computer Vision



Goal: Find globally optimal pose, fast

- Alternatives are either slow (BnB [1]) or suboptimal (SDP [2])

Equivalent formulation: Quadratic objective

- Generalized distance: $d_P(x)^2 = \|x - y\|_C^2$
- Linear transformation: $T \oplus x_i = X_i \text{vec}(T)$
- Marginalization of translation: $t \in \mathbb{R}^3$

$$f^* = \min_{R \in SO(3)} \tilde{r}^T \tilde{Q} \tilde{r}, \quad \tilde{r} = \begin{bmatrix} r \\ 1 \end{bmatrix}, \quad r = \text{vec}(R)$$

Rotation constraint:
 $R \in SO(3) \Leftrightarrow \{c_i(r) = 0\}$

How to solve this non-convex problem globally?

Non-convex!

Convex relaxation via Lagrangian duality

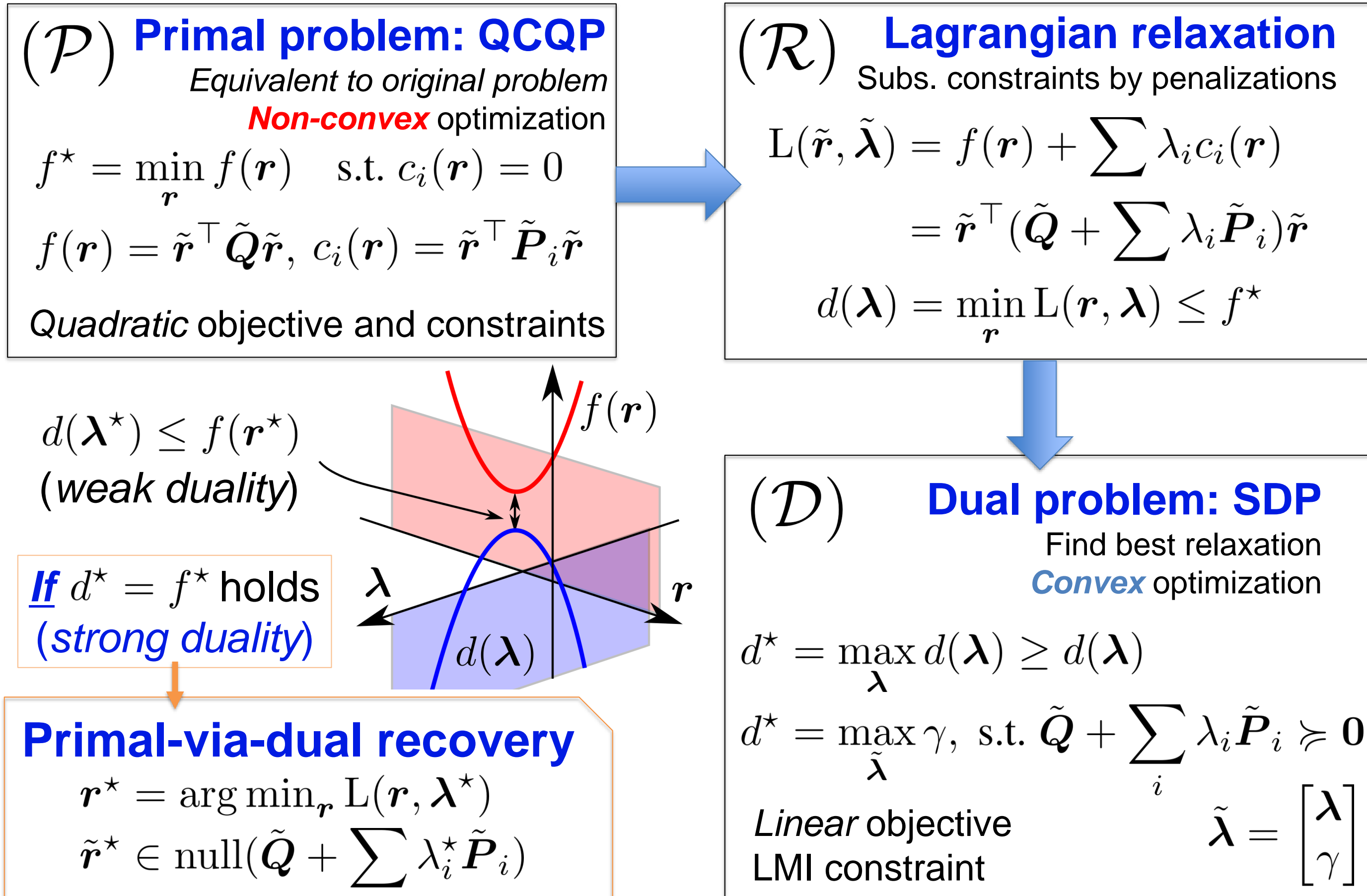
(P) Primal problem: QCQP
Equivalent to original problem
Non-convex optimization
 $f^* = \min_r f(r) \quad \text{s.t. } c_i(r) = 0$
 $f(r) = \tilde{r}^T \tilde{Q} \tilde{r}, \quad c_i(r) = \tilde{r}^T \tilde{P}_i \tilde{r}$
Quadratic objective and constraints

(R) Lagrangian relaxation
Subs. constraints by penalizations
 $L(\tilde{r}, \tilde{\lambda}) = f(r) + \sum \lambda_i c_i(r)$
 $= \tilde{r}^T (\tilde{Q} + \sum \lambda_i \tilde{P}_i) \tilde{r}$
 $d(\lambda) = \min_r L(r, \lambda) \leq f^*$

(D) Dual problem: SDP
Find best relaxation
Convex optimization
 $d^* = \max_{\lambda} d(\lambda) \geq d(\lambda)$
 $d^* = \max_{\tilde{\lambda}} \gamma, \quad \text{s.t. } \tilde{Q} + \sum \lambda_i \tilde{P}_i \succeq 0$
Linear objective
LMI constraint
 $\tilde{\lambda} = \begin{bmatrix} \lambda \\ \gamma \end{bmatrix}$

Primal-via-dual recovery
 $r^* = \arg \min_r L(r, \lambda^*)$
 $\tilde{r}^* \in \text{null}(\tilde{Q} + \sum \lambda_i^* \tilde{P}_i)$

Weak duality: $d(\lambda^*) \leq f(r^*)$
Strong duality: If $d^* = f^*$ holds



Main contribution: Exploit redundant rotation constraints

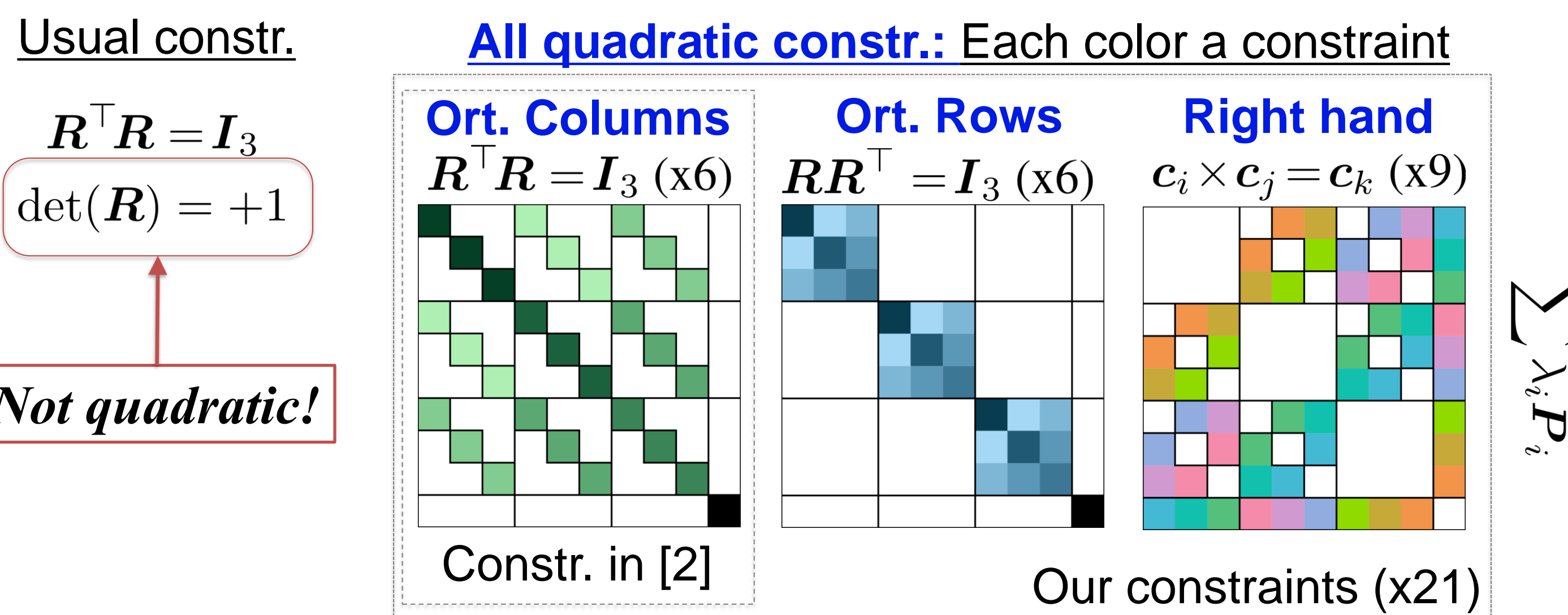
Usual constr. $R^T R = I_3$
 $\det(R) = +1$

All quadratic constr.: Each color a constraint

Ort. Columns: $R^T R = I_3$ (x6)
Ort. Rows: $RR^T = I_3$ (x6)
Right hand: $c_i \times c_j = c_k$ (x9)

Constr. in [2] vs Our constraints (x21)

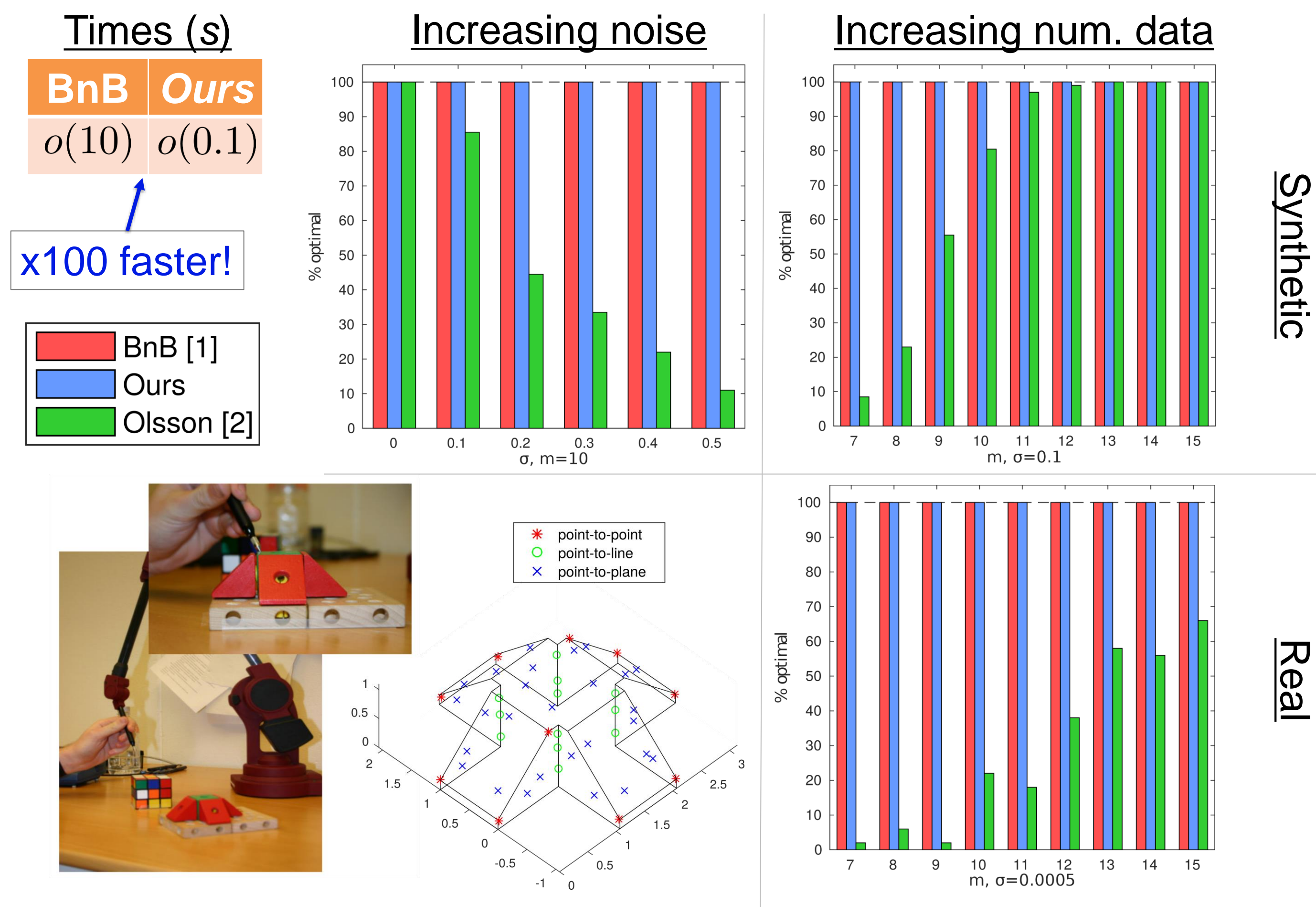
Not quadratic!



Empirical observation: If all (x21) constraints were added in (D), there was *always* strong duality ($d^* = f^*$).

Experiments: % optimal solutions (ours 100%), faster than BnB

We evaluate how often a method attains the globally optimal solution:



Conclusion

Empirically, we show the non-convexity of the constraint $R \in SO(3)$ can be circumvented when solving the studied registration problem.

Ongoing work:

Optimality verification

Theoretical guarantees

Multiple global minima

Faster SDP solver

Robust registration

References:

- [1] C. Olsson, F. Kahl and M. Oskarsson. *Branch-and-Bound Methods for Euclidean Registration Problems*. In IEEE Trans. Pattern Anal. Mach. Intell. (TPAMI), 2009.
- [2] C. Olsson and A. Eriksson. *Solving Quadratically Constrained Geometrical Problems using Lagrangian Duality*. In Intl. Conf. Pattern Recognition (ICPR), 2008.