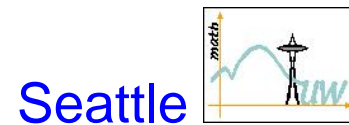


On p -Order Cone Relaxation of Sensor Network Localization

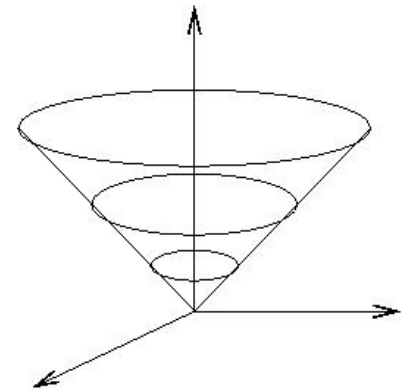
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Talk Outline

- Problem description
- p OCP relaxations
- Properties of p OCP relaxations
- Performance of SOCP relaxation and efficient solution methods
- Conclusions & Future Directions



Sensor Network Localization

Basic Problem:

- n pts in \mathbb{R}^d ($d = 1, 2, 3$).
- Know last $n - m$ pts ('anchors') x_{m+1}, \dots, x_n and ℓ_p -dist. ($1 < p < \infty$) estimate for pairs of 'neighboring' pts

$$d_{ij} \geq 0 \quad \forall (i, j) \in \mathcal{A}$$



with $\mathcal{A} \subseteq \{(i, j) : 1 \leq i < j \leq n\}$.

- Estimate first m pts ('sensors').

History? Graph realization, position estimation in wireless sensor network, determining protein structures, ...

Optimization Problem Formulation

$$v_{\text{opt}} := \min_{x_1, \dots, x_m} \sum_{(i,j) \in \mathcal{A}} \left| \|x_i - x_j\|_p - d_{ij} \right|$$

- Objective function is nonconvex. 
- Problem is NP-hard (reduction from PARTITION). 
- Use a convex (but not SDP) relaxation.

Other formulations: “ $\| \|x_i - x_j\|_p - d_{ij} \|^2$ ” or “ $\| \|x_i - x_j\|_p^p - d_{ij}^p \|^2$ ” or ...

p -Order Cone Program Relaxation

$$\begin{aligned} v_{\text{opt}} = & \min_{x_1, \dots, x_m, y_{ij}} \sum_{(i,j) \in \mathcal{A}} |y_{ij} - d_{ij}| \\ \text{s.t.} \quad & y_{ij} = \|x_i - x_j\|_p \quad \forall (i, j) \in \mathcal{A} \end{aligned}$$

Relax “=” to “ \geq ” constraint:

$$\begin{aligned} v_{\text{socp}} := & \min_{x_1, \dots, x_m, y_{ij}} \sum_{(i,j) \in \mathcal{A}} |y_{ij} - d_{ij}| \\ \text{s.t.} \quad & y_{ij} \geq \|x_i - x_j\|_p \quad \forall (i, j) \in \mathcal{A} \end{aligned}$$

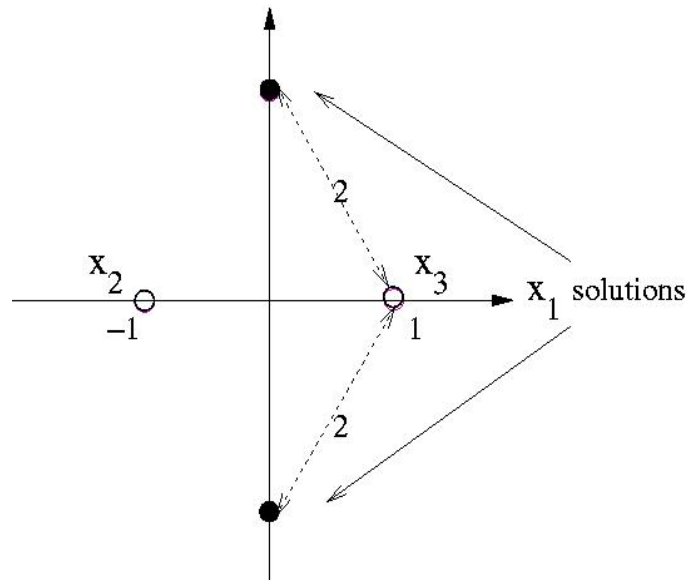
($p=2$: Doherty, Pister, El Ghaoui '03)

Properties of p OCP Relaxation

$$d = 2, n = 3, m = 1, d_{12} = d_{13} = 2$$

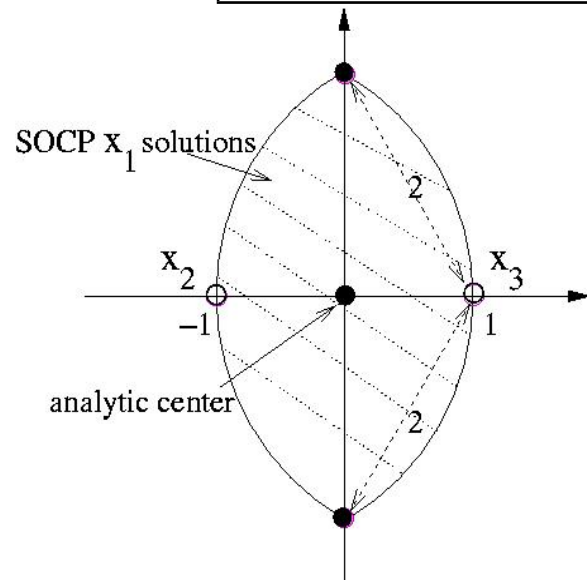
Problem:

$$0 = \min_{x_1 \in \mathbb{R}^2} \left| \|x_1 - (0, 1)\|_p - 2 \right| + \left| \|x_1 - (0, -1)\|_p - 2 \right|$$



***p*OCP Relaxation:**

$$\begin{aligned}
 0 = & \min_{\substack{x_1 \in \mathbb{R}^2 \\ y_{12}, y_{13} \in \mathbb{R}}} |y_{12} - 2| + |y_{13} - 2| \\
 \text{s.t. } & y_{12} \geq \|x_1 - (0, 1)\|_p \\
 & y_{13} \geq \|x_1 - (0, -1)\|_p
 \end{aligned}$$



If solve p OCP by IP method, then likely get analy. center of soln set.

Analytic Center Solution

Define

$$\mathcal{A}_{\text{active}} := \left\{ (i, j) \in \mathcal{A} \mid y_{ij} = \|x_i - x_j\|_p \ \forall \text{ solns } x_1, \dots, x_m, (y_{ij})_{(i,j) \in \mathcal{A}} \text{ of } p\text{OCP} \right\}$$

$$\text{Relative - interior soln} := \text{soln } x_1, \dots, x_m, (y_{ij})_{(i,j) \in \mathcal{A}} \text{ of } p\text{OCP} \text{ with} \\ y_{ij} > \|x_i - x_j\|_p \ \forall (i, j) \in \mathcal{A} \setminus \mathcal{A}_{\text{active}}$$

$$\text{Analytic center soln} := \arg \min_{\substack{\text{rel.-int. soln} \\ x_1, \dots, x_m, (y_{ij})_{(i,j) \in \mathcal{A}}}} \sum_{(i,j) \in \mathcal{A} \setminus \mathcal{A}_{\text{active}}} -\log (y_{ij}^p - \|x_i - x_j\|_p^p)$$

Properties of p OCP Relaxations

Key Property: If $p = 2$ or $d = 2$, then

$$\arg \min_x \sum_{i=1}^k \|x - x_i\|_p \in \text{conv} \{x_1, \dots, x_k\}$$

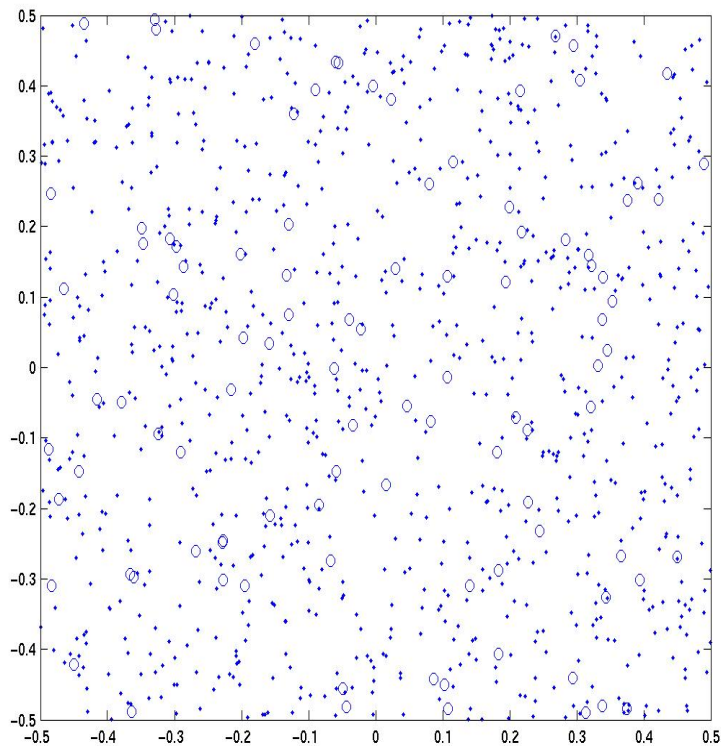
for any k and $x_1, \dots, x_k \in \mathbb{R}^d$. (Remains true if $\|x - x_i\|_p$ is raised to p th power.)

Q: What if $d \geq 3$ and $p \neq 2$? (False if $\|x - x_i\|_p$ is raised to p th power. (S. Zhang '05, C.-K. Sim '03))

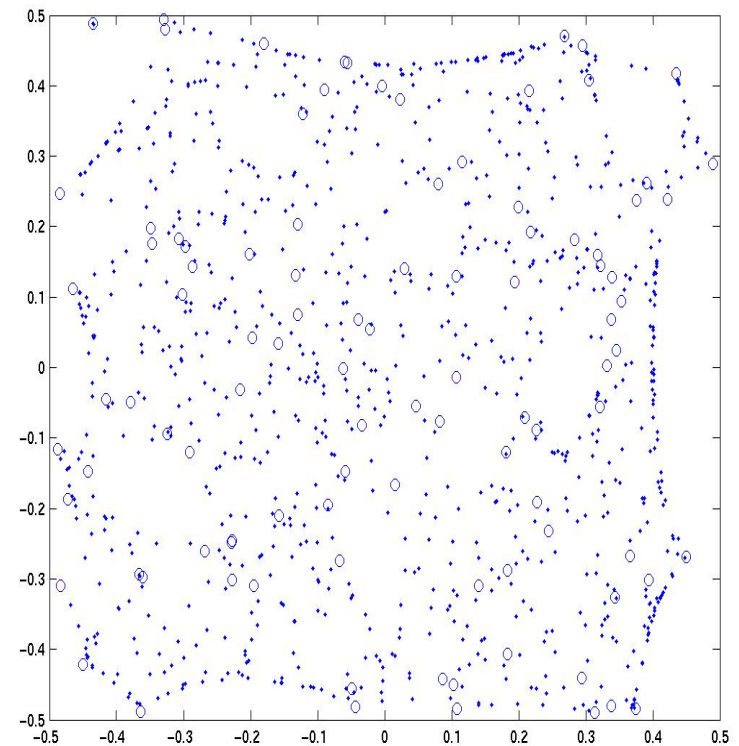
Fact 1: If $(x_1, \dots, x_m, y_{ij})_{(i,j) \in \mathcal{A}}$ is the analytic center soln of p OCP, then

$$x_i \in \text{conv} \{x_j\}_{j \in \mathcal{N}(i)} \quad \forall i \leq m$$

with $\mathcal{N}(i) := \{j \mid (i, j) \in \mathcal{A}\}$.



True soln ($m = 900$, $n = 1000$,
nhbrs if ℓ_2 -dist $< .06$)



SOCP soln found by IP method
(SeDuMi)

Fact 2: If $(x_1, \dots, x_m, y_{ij})_{(i,j) \in \mathcal{A}}$ is a relative-interior p OCP soln (e.g., analytic center), then for each $i \leq m$,

$$\|x_i - x_j\|_p = y_{ij} \quad \text{for some } j \in \mathcal{N}(i) \quad \iff \quad x_i \text{ appears in every } p\text{OCP soln.}$$

Error Bounds

What if distances have errors?

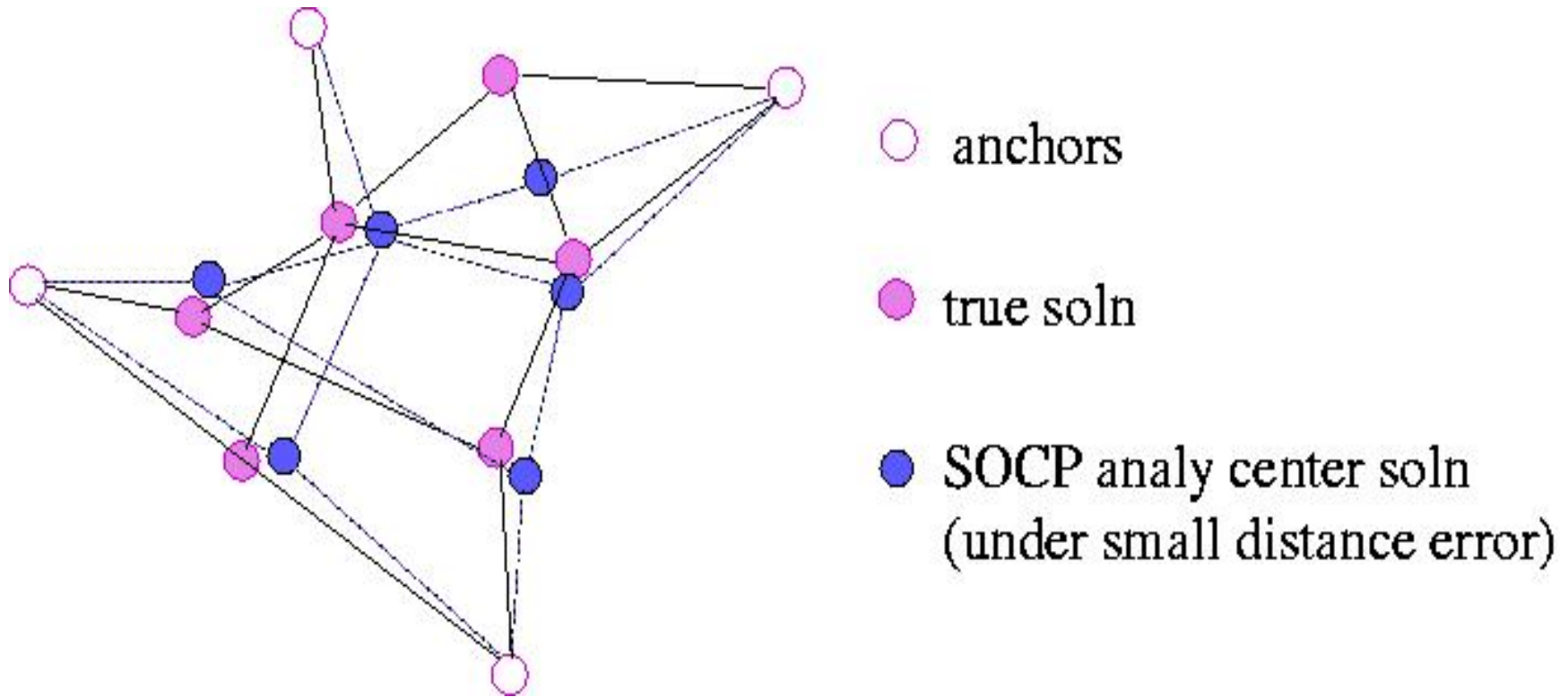
$$d_{ij} = d_{ij}^{\text{true}} + \delta_{ij},$$

where $\delta_{ij} \in \mathfrak{R}$ and $d_{ij}^{\text{true}} := \|x_i^{\text{true}} - x_j^{\text{true}}\|_p$ ($x_i^{\text{true}} = x_i \forall i > m$).

Fact 3: If $(x_1, \dots, x_m, y_{ij})_{(i,j) \in \mathcal{A}}$ is a relative-interior p OCP soln corresp. $(d_{ij})_{(i,j) \in \mathcal{A}}$ and $\sum_{(i,j) \in \mathcal{A}} |\delta_{ij}| \leq \delta$, then for each i ,

$$\|x_i - x_j\|_p = y_{ij} \quad \text{for some } j \in \mathcal{N}(i) \quad \implies \quad \|x_i - x_i^{\text{true}}\|_p = O\left(\sqrt[p]{\sum_{(i,j) \in \mathcal{A}} |\delta_{ij}|}\right).$$

Fact 4: As $\sum_{(i,j) \in \mathcal{A}} |\delta_{ij}| \rightarrow 0$, (analytic center p OCP soln corresp. $(d_{ij})_{(i,j) \in \mathcal{A}}$)
 \rightarrow (analytic center p OCP soln corresp. $(d_{ij}^{\text{true}})_{(i,j) \in \mathcal{A}}$).



Error bounds for the analytic center p OCP soln when distances have small errors.

Solving p OCP Relaxation I: IP Method

$$\begin{aligned} \min_{x_1, \dots, x_m, y_{ij}} \quad & \sum_{(i,j) \in \mathcal{A}} |y_{ij} - d_{ij}| \\ \text{s.t.} \quad & y_{ij} \geq \|x_i - x_j\|_p \quad \forall (i, j) \in \mathcal{A} \end{aligned}$$

Put into conic form:

$$\begin{aligned} \min \quad & \sum_{(i,j) \in \mathcal{A}} u_{ij} \\ \text{s.t.} \quad & x_i - x_j - w_{ij} = 0 \quad \forall (i, j) \in \mathcal{A} \\ & y_{ij} - u_{ij} = d_{ij} \quad \forall (i, j) \in \mathcal{A} \\ & u_{ij} \geq 0, \quad (y_{ij}, w_{ij}) \in p\text{-order cone} \quad \forall (i, j) \in \mathcal{A}. \end{aligned}$$

Solve by an IP method, e.g., SeDuMi or Mosek for $p = 2$.

Solving p OCP Relaxation II: Smoothing + Coordinate Gradient Descent

$$\min_{y \geq z} |y - d| = \max\{0, z - d\}$$

So p OCP relaxation:

$$\min_{x_1, \dots, x_m} \sum_{(i,j) \in \mathcal{A}} \max\{0, \|x_i - x_j\|_p - d_{ij}\}$$

This is an unconstrained nonsmooth convex program.

- Smooth approximation:

$$\max\{0, t\} \approx h_\mu(t) := \frac{(t^2 + \mu^2)^{1/2} + t}{2} \quad (\mu > 0)$$

p OCP approximation:

$$\min \quad f_\mu(x_1, \dots, x_m) := \sum_{(i,j) \in \mathcal{A}} h_\mu(\|x_i - x_j\|_p - d_{ij} + \mu)$$

Add a smoothed log-barrier term $-\mu \sum_{(i,j) \in \mathcal{A}} \log(\mu + h_\mu(d_{ij} - \mu - \|x_i - x_j\|_p))$

Solve the smooth approximation using coordinate gradient descent (SCGD):

- If $\|\nabla_{x_i} f_\mu\| = \Omega(\mu)$, then update x_i by moving it along the Newton direction $-\left[\nabla_{x_i x_i}^2 f_\mu\right]^{-1} \nabla_{x_i} f_\mu$, with Armijo stepsize rule, and re-iterate.
- Decrease μ when $\|\nabla_{x_i} f_\mu\| = O(\mu) \forall i$.

$\mu^{\text{init}} = 1e - 5$. $\mu^{\text{end}} = 1e - 9$. Decrease μ by a factor of 10.

Code in Fortran. Computation easily distributes.

Simulation Results for $p = 2$

- Uniformly generate $x_1^{\text{true}}, \dots, x_n^{\text{true}}$ in $[0, 1]^2$, $m = .9n$,
two pts are nhbrs if $\text{dist} < \text{radio}R$.

Set

$$d_{ij} = \|x_i^{\text{true}} - x_j^{\text{true}}\|_2 \cdot |1 + \epsilon_{ij} \cdot nf|,$$

$$\epsilon_{ij} \sim N(0, 1)$$

(Biswas, Ye '03)

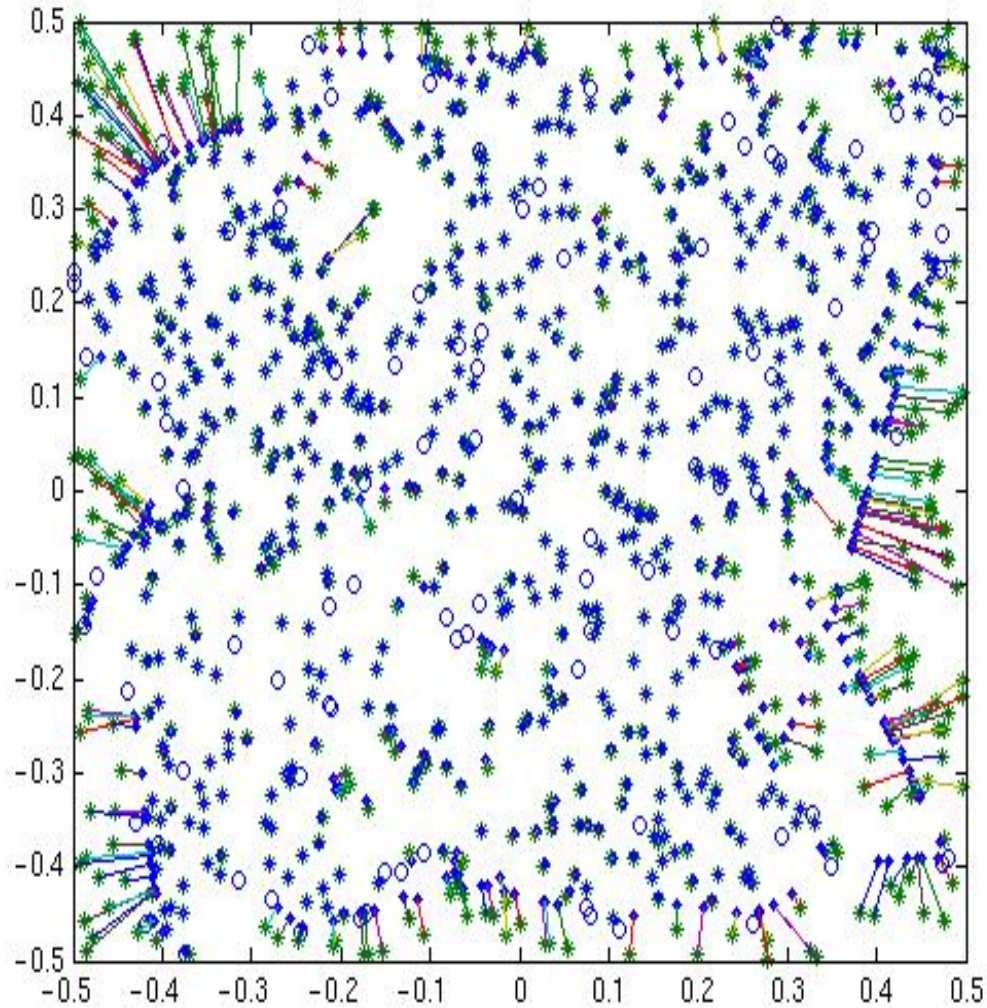
- Solve SOCP using SeDuMi 1.05 or SCGD.
- Sensor i is uniquely positioned if

$$\left| \|x_i - x_j\|_2 - y_{ij} \right| \leq 10^{-7} d_{ij} \quad \text{for some } j \in \mathcal{N}(i).$$

n	m	nf	SeDuMi	“SCGD”
			cpu/ m_{up} / Err_{up}	cpu/ m_{up} / Err_{up}
1000	900	0	1.0/327/4.6e-5	.2/357/3.8e-5
1000	900	.001	0.9/438/1.6e-3	.4/442/1.5e-3
1000	900	.01	1.1/555/1.5e-2	1.6/518/1.1e-2
2000	1800	0	33.7/1497/4.3e-4	0.8/1541/3.3e-4
2000	1800	.001	38.0/1465/3.3e-3	1.8/1466/3.6e-3
2000	1800	.01	16.2/1704/6.3e-2	2.9/1707/5.1e-2
4000	3600	0	17.8/2758/3.0e-4	1.6/2844/3.2e-4
4000	3600	.001	20.4/2907/3.2e-3	5.1/2894/3.0e-3
4000	3600	.01	17.2/3023/9.1e-3	6.1/3020/9.1e-3

Table 1: $radioR = .06(.035)$ for $n = 1000, 2000(4000)$

- cpu (minutes) times are on a HP DL360 workstation, running Linux 3.5.
- m_{up} := number of uniquely positioned sensors.
- $\text{Err}_{\text{up}} := \max_{i \text{ uniq. pos.}} \|x_i - x_i^{\text{true}}\|_2$.



- = anchors * = true positions of sensors
● = SOCP soln found by SCGD ($m = 900$, $n = 1000$, $radioR = .06$, $nf = .01$)

Conclusions & Future Directions

- Finding an analytic center soln of p OCP is key.
- Exploit network structures of p OCP? Distributed computation?
- Additional linear constraints?
- Other objective functions, e.g., $\sum_{(i,j) \in \mathcal{A}} \left| \|x_i - x_j\|_p - d_{ij} \right|^2$?
- How to localize sensors not uniquely positioned? Smoothing + local descent on original objective function...
- Extend error bound result to SDP and ESDP relaxations (Biswas, Ye '04, ..., Wang et al. '06) of $\min_{x_1, \dots, x_m} \sum_{(i,j) \in \mathcal{A}} \left| \|x_i - x_j\|_2^2 - d_{ij}^2 \right|$?