

# ESDP Relaxation of Sensor Network Localization

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(Ongoing work with Paul Tseng)

## Talk Outline

- Sensor network localization
- ESDP relaxation: formulation and properties
- A robust version of ESDP for the noisy case
- Numerical simulations

## Sensor Network Localization

- $n$  pts  $\underbrace{x_1, \dots, x_m}_{\text{sensors}}, \underbrace{x_{m+1}, \dots, x_n}_{\text{anchors}}$  in  $\mathbb{R}^2$ .
- Know last  $n - m$  pts ('anchors')  $x_{m+1}, \dots, x_n$  and Eucl. dist. estimate for some pairs of 'neighboring' pts (i.e. within 'radio range')

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

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

- Estimate the first  $m$  pts ('sensors')  $x_1, \dots, x_m$ .

## Optimization Problem Formulation

$$v_p := \min_{x_1, \dots, x_m} \sum_{(i,j) \in \mathcal{A}} \left| \|x_i - x_j\|^2 - d_{ij}^2 \right|$$

- Objective function is nonconvex.  $m$  can be large ( $m > 1000$ ).
- Problem is NP-hard (reduction from PARTITION).
- Use a convex relaxation.

## SDP Relaxation

$$X = [x_1 \cdots x_m];$$

SDP Relaxation (Biswas, Ye '03):

$$\begin{aligned}
 v_{\text{sdp}} := \min_Z & \sum_{(i,j) \in \mathcal{A}, j > m} |y_{ii} - 2x_j^T x_i + \|x_j\|^2 - d_{ij}^2| \\
 & + \sum_{(i,j) \in \mathcal{A}, j \leq m} |y_{ii} - 2y_{ij} + y_{jj} - d_{ij}^2| \\
 \text{s.t. } & Z = \begin{bmatrix} Y & X^T \\ X & I \end{bmatrix} \succeq 0
 \end{aligned}$$

Adding nonconvex constraint  $\text{rank } Z = 2$  yields the original problem.

## ESDP Relaxation

ESDP relaxation (Wang, Zheng, Boyd, Ye '07):

$$\begin{aligned}
 v_{\text{esdp}} := & \min_Z \sum_{(i,j) \in \mathcal{A}, j > m} |y_{ii} - 2x_j^T x_i + \|x_j\|^2 - d_{ij}^2| \\
 & + \sum_{(i,j) \in \mathcal{A}, j \leq m} |y_{ii} - 2y_{ij} + y_{jj} - d_{ij}^2| \\
 \text{s.t. } & Z = \begin{bmatrix} Y & X^T \\ X & I \end{bmatrix} \\
 & \begin{bmatrix} y_{ii} & y_{ij} & x_i^T \\ y_{ij} & y_{jj} & x_j^T \\ x_i & x_j & I \end{bmatrix} \succeq 0 \quad \forall (i,j) \in \mathcal{A}, j \leq m \\
 & \begin{bmatrix} y_{ii} & x_i^T \\ x_i & I \end{bmatrix} \succeq 0 \quad \forall i = 1, \dots, m
 \end{aligned}$$

## ESDP: Testing Solution Accuracy

**Aim:** Determine which sensors are accurately positioned by ESDP.

Consider

$$d_{ij}^2 = \|x_i^{\text{true}} - x_j^{\text{true}}\|^2 + \delta_{ij}$$

Two cases:

- Noiseless case, i.e.  $\delta_{ij} = 0, \forall (i, j) \in \mathcal{A}$ ;
- Noisy case, i.e.  $\delta_{ij} \neq 0$  for some  $(i, j) \in \mathcal{A}$ .

**Defn:**  $x_i$  is *invariant* over  $\text{Sol}(\text{ESDP})$  if  $\forall Z \in \text{Sol}(\text{ESDP})$ , its  $x_i$  entry is the same.

## ESDP: Noiseless Case

- $v_{\text{esdp}} = 0 \Rightarrow \text{Sol}(P) \subseteq \left\{ X : \begin{bmatrix} Y & X^T \\ X & I \end{bmatrix} \in \text{Sol}(\text{ESDP}) \right\}$
- **Observation:** If  $x_i$  is invariant over  $\text{Sol}(\text{ESDP})$ , then  $x_i$  is invariant over  $\text{Sol}(P)$  and hence  $x_i = x_i^{\text{true}}$ .



## ESDP: Noiseless Case

Consider the **individual trace** for each sensor location  $i$ :

$$\text{tr}_i[Z] := y_{ii} - \|x_i\|^2, \quad i = 1, \dots, m. \quad (\text{Biswas, Ye '03})$$

- **(Wang et al '07)** If  $\text{tr}_i[Z] = 0$  for some  $Z \in \text{ri}(\text{Sol}(\text{ESDP}))$ , then  $x_i$  is invariant over  $\text{Sol}(\text{ESDP})$ .

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How necessary is “ $\text{tr}_i(Z) = 0$  for some  $Z \in \text{ri}(\text{Sol}(\text{ESDP}))$ ” for invariance?

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**New:** **(P, Tseng '08)** If  $x_i$  is invariant over  $\text{Sol}(\text{ESDP})$ , then  $\text{tr}_i[Z] = 0$  for all  $Z \in \text{Sol}(\text{ESDP})$ .

## ESDP: Noisy Case

Suppose  $\delta_{ij} \neq 0$  for some  $(i, j) \in \mathcal{A}$ .

Does  $\text{tr}_i[Z] = 0$  (with  $Z \in \text{ri}(\text{Sol}(\text{ESDP}))$ ) imply  $x_i$  is near the true position of sensor  $i$ ?

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No.

**New:** (P, Tseng '08): For  $|\delta_{ij}| \approx 0$ ,

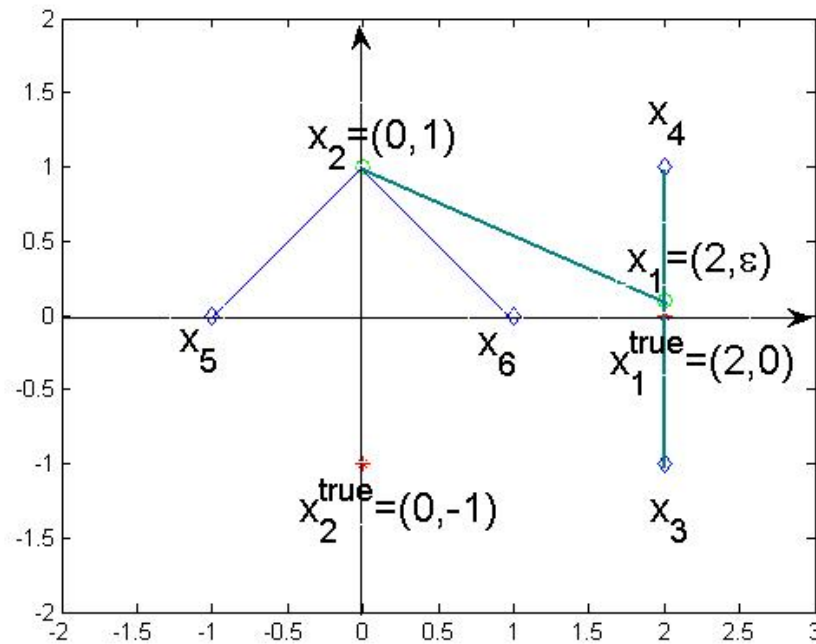
$$\text{tr}_i[Z] = 0 \quad \text{for some } Z \in \text{ri}(\text{Sol}(\text{ESDP})) \not\Rightarrow \|x_i - x_i^{\text{true}}\| \approx 0.$$

Proof is by counterexample.

An example of sensitivity of ESDP solns to measurement noise:

Input distance data:  $\epsilon > 0$

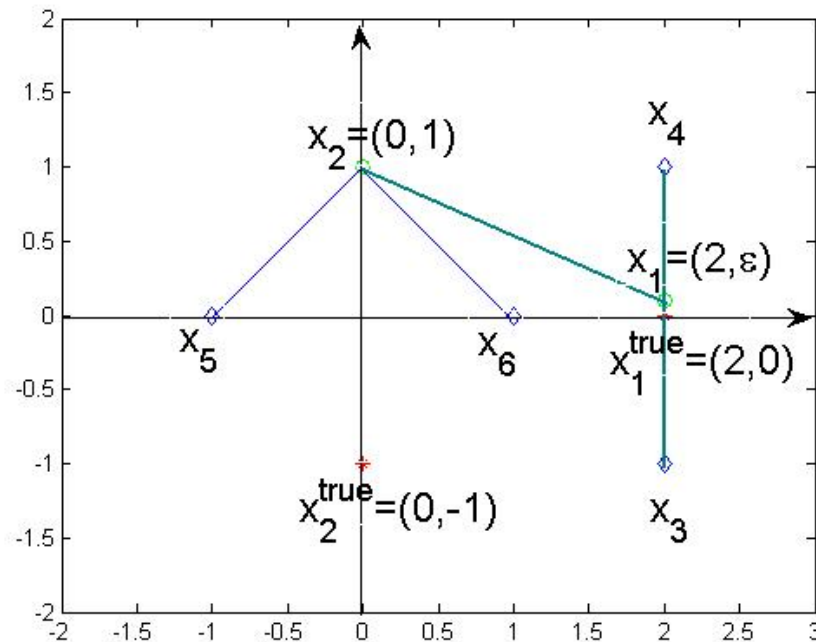
$d_{12} = \sqrt{4 + (1 - \epsilon)^2}$ ,  $d_{13} = 1 + \epsilon$ ,  $d_{14} = 1 - \epsilon$ ,  $d_{25} = d_{26} = \sqrt{2}$ ;  $m = 2$ ,  $n = 6$ .



An example of sensitivity of ESDP solns to measurement noise:

Input distance data:  $\epsilon > 0$

$$d_{12} = \sqrt{4 + (1 - \epsilon)^2}, d_{13} = 1 + \epsilon, d_{14} = 1 - \epsilon, d_{25} = d_{26} = \sqrt{2}; m = 2, n = 6.$$



Thus, even when  $Z \in \text{Sol}(\text{ESDP})$  is unique,  $\text{tr}_i[Z] = 0$  fails to certify accuracy of  $x_i$  in the noisy case!

## Robust ESDP

For each  $(i, j) \in \mathcal{A}$ , fix  $\rho_{ij} > |\delta_{ij}|$ . ( $\rho > |\delta|$ )

Sol( $\rho$ ESDP) denotes the set of  $Z = \begin{bmatrix} Y & X^T \\ X & I \end{bmatrix}$  satisfying

$$\begin{aligned} & \begin{bmatrix} y_{ii} & y_{ij} & x_i^T \\ y_{ij} & y_{jj} & x_j^T \\ x_i & x_j & I \end{bmatrix} \succeq 0 \quad \forall (i, j) \in \mathcal{A}, j \leq m \\ & \begin{bmatrix} y_{ii} & x_i^T \\ x_i & I \end{bmatrix} \succeq 0 \quad \forall i = 1, \dots, m \\ & |y_{ii} - 2x_j^T x_i + \|x_j\|^2 - d_{ij}^2| \leq \rho_{ij} \quad \forall (i, j) \in \mathcal{A}, j > m \\ & |y_{ii} - 2y_{ij} + y_{jj} - d_{ij}^2| \leq \rho_{ij} \quad \forall (i, j) \in \mathcal{A}, j \leq m \end{aligned}$$

**Note:**  $[X^{\text{true}} \ I][X^{\text{true}} \ I]^T \in \text{Sol}(\rho\text{ESDP})$ .



Let

$$Z^{\rho, \delta} := \arg \min_{Z \in \text{Sol}(\rho \text{ESDP})} - \sum_{(i,j) \in \mathcal{A}, j \leq m} \ln \det \left( \begin{bmatrix} y_{ii} & y_{ij} & x_i^T \\ y_{ij} & y_{jj} & x_j^T \\ x_i & x_j & I \end{bmatrix} \right) \\ - \sum_{i \leq m} \ln \det \left( \begin{bmatrix} y_{ii} & x_i^T \\ x_i & I \end{bmatrix} \right)$$

**New:** (P, Tseng '08)  $\exists \bar{\tau} > 0, \forall \tau \leq \bar{\tau}, \exists \bar{\rho} > 0$  such that for each  $i$ ,

$$\text{tr}_i[Z^{\rho, \delta}] < \tau \text{ for some } 0 \leq |\delta| < \rho \leq \bar{\rho} \\ \implies \lim_{|\delta| < \rho \rightarrow 0} x_i^{\rho, \delta} = x_i^{\text{true}}$$

Moreover,

$$\|x_i^{\rho, \delta} - x_i^{\text{true}}\| \leq \sqrt{2|\mathcal{A}| + m} (\text{tr}_i[Z^{\rho, \delta}])^{\frac{1}{2}} \quad 0 \leq |\delta| < \rho.$$

## Algorithm: CGD-barrier

Let  $h_a(t) := \frac{1}{2}(t - a)_+^2 + \frac{1}{2}(-t - a)_+^2$ .

$$\begin{aligned}
 f_\mu(Z) := & \sum_{(i,j) \in \mathcal{A}, j > m} h_{\rho_{ij}}(y_{ii} - 2x_j^T x_i + \|x_j\|^2 - d_{ij}^2) \\
 & + \sum_{(i,j) \in \mathcal{A}, j \leq m} h_{\rho_{ij}}(y_{ii} - 2y_{ij} + y_{jj} - d_{ij}^2) \\
 & - \mu \sum_{(i,j) \in \mathcal{A}, j \leq m} \ln \det \left( \begin{bmatrix} y_{ii} & y_{ij} & x_i^T \\ y_{ij} & y_{jj} & x_j^T \\ x_i & x_j & I \end{bmatrix} \right) \\
 & - \mu \sum_{i \leq m} \ln \det \left( \begin{bmatrix} y_{ii} & x_i^T \\ x_i & I \end{bmatrix} \right)
 \end{aligned}$$

## Algorithm: CGD-barrier

- For each  $(i, j) \in \mathcal{A}$ ,

$$h_{\rho_{ij}}(y_{ii} - 2x_j^T x_i + \|x_j\|^2 - d_{ij}^2) = 0 \iff |y_{ii} - 2x_j^T x_i + \|x_j\|^2 - d_{ij}^2| \leq \rho_{ij}$$

$$h_{\rho_{ij}}(y_{ii} - 2y_{ij} + y_{jj} - d_{ij}^2) = 0 \iff |y_{ii} - 2y_{ij} + y_{jj} - d_{ij}^2| \leq \rho_{ij}.$$

- For each  $\rho > |\delta|$ ,  $\operatorname{argmin} f_\mu \rightarrow Z^{\rho, \delta}$  as  $\mu \rightarrow 0$ .
- In the noiseless case ( $\delta = 0$ ), setting  $\rho = 0$ , then, as  $\mu \rightarrow 0$ ,  $\operatorname{argmin} f_\mu \rightarrow Z$  for some  $Z \in \operatorname{ri}(\operatorname{Sol}(\operatorname{ESDP}))$ .

## Algorithm: CGD-barrier

Use Block Coordinate Gradient Descent:

For each  $i$ , compute gradient  $\nabla_{Z_i} f_\mu$  of  $f_\mu$  w.r.t. the subset of variables  $Z_i := \{x_i, y_{ii}, y_{ij} : (i, j) \in \mathcal{A}\}$ .

- If  $\|\nabla_{Z_i} f_\mu\| \geq \max\{\mu, 1e - 6\}$  for some  $i$ , update  $Z_i$  by moving along the Newton direction  $-\left(\nabla_{Z_i Z_i}^2 f_\mu\right)^{-1} \nabla_{Z_i} f_\mu$  with Armijo stepsize rule.
- Decrease  $\mu$  when  $\|\nabla_{Z_i} f_\mu\| < \max\{\mu, 1e - 6\}$  for all  $i$ .

$\mu_{\text{initial}} = 100$ ,  $\mu_{\text{final}} = 1e - 9$ . Decrease  $\mu$  by a factor of 10.

Coded in Fortran. Computation easily distributes.

## Simulation Results

- Compare  $\rho$ ESDP as solved by CGD-barrier and ESDP as solved by Sedumi (with the interface to Sedumi coded by Wang et al.).
- Uniformly generated  $\{x_1^{\text{true}}, \dots, x_n^{\text{true}}\}$  in  $[-.5, .5]^2$ ,  $m = .9n$ . Two pts are neighbors iff  $\text{dist} < rr$ . Set

$$d_{ij} = \|x_i^{\text{true}} - x_j^{\text{true}}\| \cdot \max\{0, 1 + \epsilon_{ij} \cdot nf\},$$

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

- Sensor  $i$  is judged as “accurately positioned” if

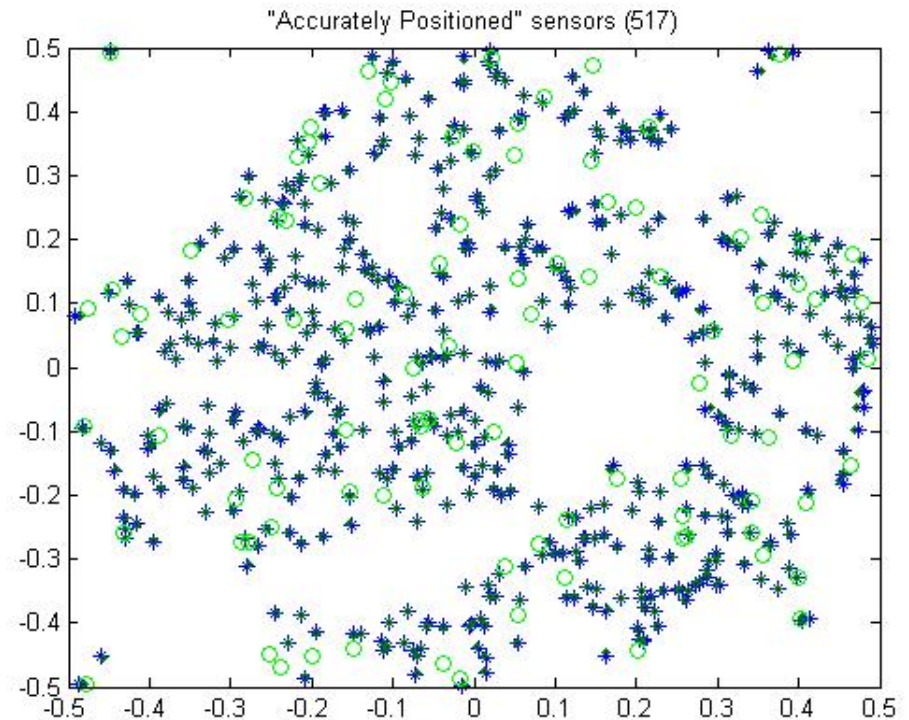
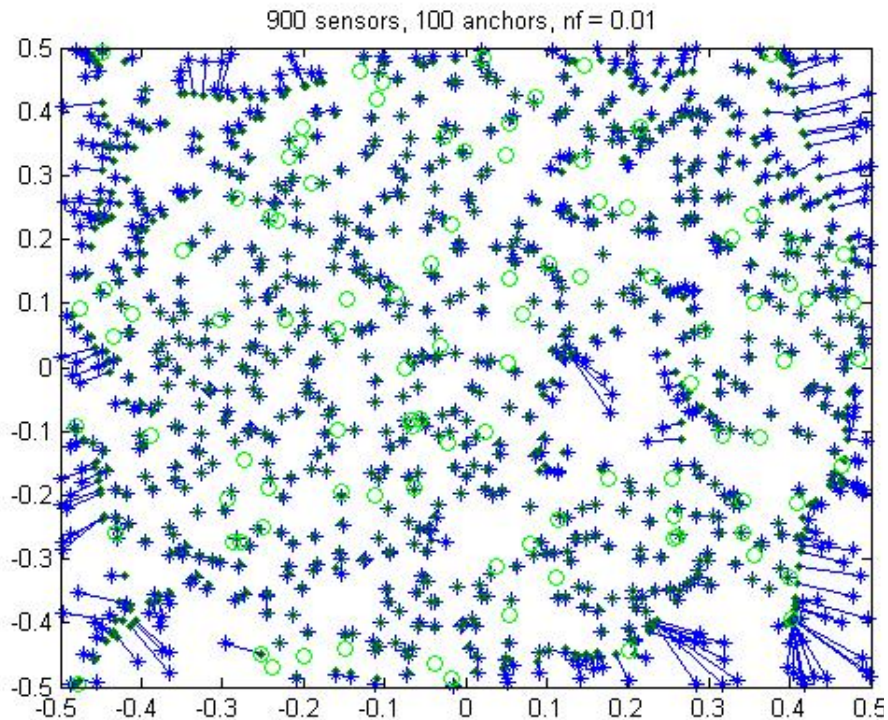
$$\text{tr}_i[Z^{\rho, \delta}] < 5 \times 10^{-6} + 0.02nf.$$

## Simulation Results

				$\rho$ ESDP <sub>CGD-barrier</sub>	ESDP <sub>Sedumi</sub>
$n$	$m$	$nf$	$rr$	<b>cpu</b> / $m_{ap}$ / $err_{ap}$	<b>cpu(cpus)</b> / $m_{ap}$ / $err_{ap}$
1000	900	0	.06	31/574/3.4e-4/	189(106)/626/2.2e-4
1000	900	.001	.06	23/520/2.8e-3	170(89)/624/3.1e-3
1000	900	.01	.06	14/517/1.1e-2	128(48)/664/1.5e-2
2000	1800	0	.06	63/1626/1.7e-4/	1157(397)/1689/2.7e-4
2000	1800	.001	.06	50/1596/8.5e-4/	1255(503)/1653/1.3e-3
2000	1800	.01	.06	52/1602/6.2e-3/	1374(417)/1689/1.2e-2

- cpu(sec) times are on a HP DL360 workstation, running Linux 3.5. ESDP is solved by Sedumi; cpus:= time taken in running Sedumi.
- Take  $\rho_{ij} = d_{ij}^2 \cdot ((1 - 2 \cdot nf)^{-2} - 1)$ .
- $m_{ap}$  := number of accurately positioned sensors.  
 $err_{ap} := \max_{i \text{ accurate. pos.}} \|x_i - x_i^{\text{true}}\|$ .

900 sensors, 100 anchors,  $rr = 0.06$ ,  $nf = 0.01$ , solving  $\rho$ ESDP by CGD-barrier method.  
 $x_i^{\text{true}}$  denoted by blue asterisks,  $x_i^{\rho, \delta}$  denoted by green dots, anchors denoted by circles.  $x_i^{\text{true}}$   
and  $x_i^{\rho, \delta}$  joined by blue straight line.



Thanks for coming! ☺