

Interval Robotics

Chapter 5: Robust estimation

L. Jaulin



Exercise. A robot measures its own distance to three marks. The distances and the coordinates of the marks are

mark	x_i	y_i	d_i
1	0	0	[22, 23]
2	10	10	[10, 11]
3	30	-30	[53, 54]

- 1) Define the set \mathbb{X} of all feasible positions.
- 2) Build the contractor associated with \mathbb{X} .
- 3) Build the contractor associated with $\overline{\mathbb{X}}$.

Solution.

$$\mathbb{X} = \bigcap_{i \in \{1,2,3\}} \underbrace{\left\{ (x,y) \mid (x-x_i)^2 + (y-y_i)^2 \in [d_i^-, d_i^+] \right\}}_{\mathbb{X}_i}$$

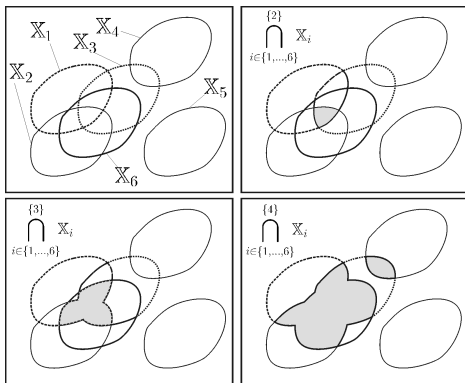
Relaxed intersection

Dealing with outliers

$$\mathcal{C} = (\mathcal{C}_1 \cap \mathcal{C}_2) \cup (\mathcal{C}_2 \cap \mathcal{C}_3) \cup (\mathcal{C}_1 \cap \mathcal{C}_3)$$

Consider m sets $\mathbb{X}_1, \dots, \mathbb{X}_m$ of \mathbb{R}^n . The q -relaxed intersection $\bigcap^{\{q\}} \mathbb{X}_i$ is the set of all $\mathbf{x} \in \mathbb{R}^n$ which belong to all \mathbb{X}_i 's, except q at most. We have

$$\mathbf{x} \in \bigcap^{\{q\}} \mathbb{X}_i \Leftrightarrow \#\{i | \mathbf{x} \in \mathbb{X}_i\} \geq m - q$$



Exercise. Compute

$$\bigcap_{i} \mathbb{X}_i = ?$$

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Solution. we have

$$\bigcap_{\{0\}} \mathbb{X}_i = \emptyset$$

$$\bigcap_{\{1\}} \mathbb{X}_i = \emptyset$$

$$\bigcap_{\{5\}} \mathbb{X}_i = \bigcup \mathbb{X}_i$$

$$\bigcap_{\{6\}} \mathbb{X}_i = \mathbb{R}^2$$

Exercise. Consider 8 intervals: $\mathbb{X}_1 = [1, 4]$,
 $\mathbb{X}_2 = [2, 4]$, $\mathbb{X}_3 = [2, 7]$, $\mathbb{X}_4 = [6, 9]$, $\mathbb{X}_5 = [3, 4]$, $\mathbb{X}_6 = [3, 7]$.
 Compute

$$\begin{aligned} \bigcap_{i \in \{0\}} \mathbb{X}_i &= ?, & \bigcap_{i \in \{1\}} \mathbb{X}_i &= ?, & \bigcap_{i \in \{2\}} \mathbb{X}_i &= ?, \\ \bigcap_{i \in \{3\}} \mathbb{X}_i &= ?, & \bigcap_{i \in \{4\}} \mathbb{X}_i &= ?, \\ \bigcap_{i \in \{5\}} \mathbb{X}_i &= ?, & \bigcap_{i \in \{6\}} \mathbb{X}_i \mathbb{X}_i &= ?. \end{aligned}$$

Solution. For $\mathbb{X}_1 = [1, 4]$,
 $\mathbb{X}_2 = [2, 4]$, $\mathbb{X}_3 = [2, 7]$, $\mathbb{X}_4 = [6, 9]$, $\mathbb{X}_5 = [3, 4]$, $\mathbb{X}_6 = [3, 7]$, we
 have

$$\bigcap_{\{0\}} \mathbb{X}_i = \emptyset, \quad \bigcap_{\{1\}} \mathbb{X}_i = [3, 4], \quad \bigcap_{\{2\}} \mathbb{X}_i = [3, 4],$$

$$\bigcap_{\{3\}} \mathbb{X}_i = [2, 4] \cup [6, 7], \quad \bigcap_{\{4\}} \mathbb{X}_i = [2, 7],$$

$$\bigcap_{\{5\}} \mathbb{X}_i = [1, 9], \quad \bigcap_{\{6\}} \mathbb{X}_i = \mathbb{R}.$$

If \mathbb{X}_i 's are intervals, the relaxed intersection can be computed with a complexity of $m \log m$.

Take all bounds of all intervals with their brackets.

Bounds	1	4	2	4	2	7	6	9	3	4	3	7
Brackets	[]	[]	[]	[]	[]	[]

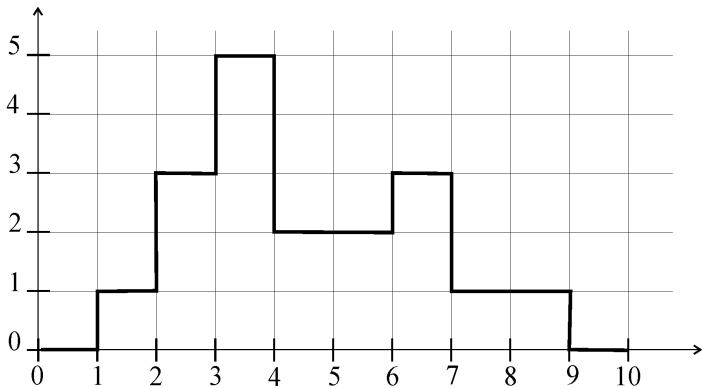
Sort the columns with respect the bounds:

Bounds	1	2	2	3	3	4	4	4	6	7	7	9
Brackets	[[[[[]]]	[]]]

Scan from left to right, counting +1 for '[' and -1 for ']':

Bounds	1	2	2	3	3	4	4	4	6	7	7	9
Brackets	[[[[[]]]	[]]]
Sum	1	2	3	4	5	4	3	2	3	2	1	0

Read the q -intersections



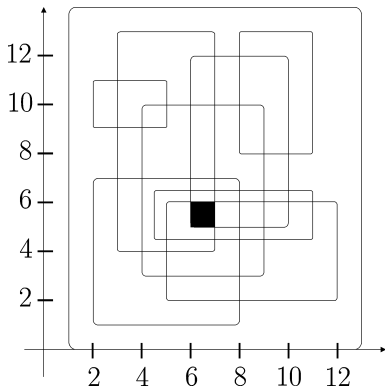
Computing the q relaxed intersection of m boxes is tractable.

Relaxed intersection

Probabilistic set estimation

Application to localization

With real data



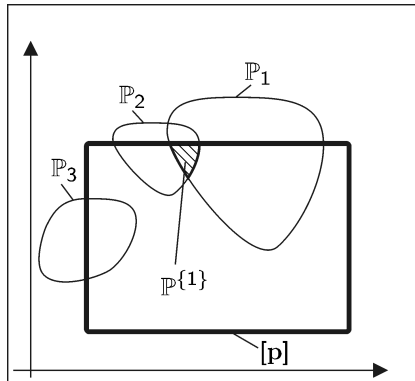
Relaxation of contractors

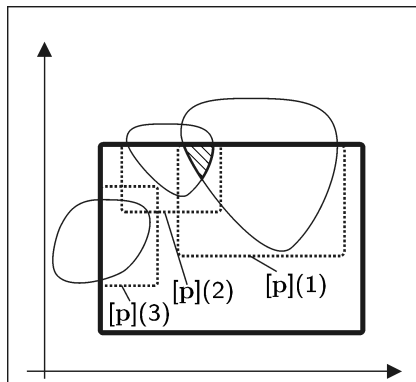
We define the q -relaxed intersection between m contractors

$$\mathcal{C} = \left(\bigcap_{i \in \{1, \dots, m\}}^{\{q\}} \mathcal{C}_i \right) \Leftrightarrow \forall [\mathbf{x}] \in \mathbb{IR}^n, \mathcal{C}([\mathbf{x}]) = \bigcap^{\{q\}} \mathcal{C}_i([\mathbf{x}]).$$

Relaxed intersection

Probabilistic set estimation
Application to localization
With real data



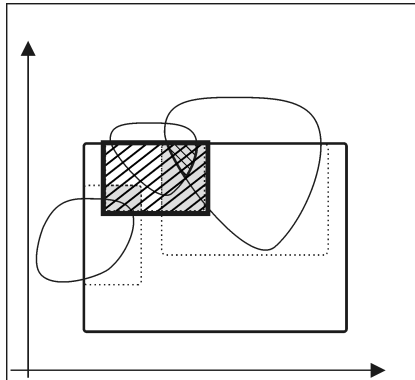


Relaxed intersection

Probabilistic set estimation

Application to localization

With real data

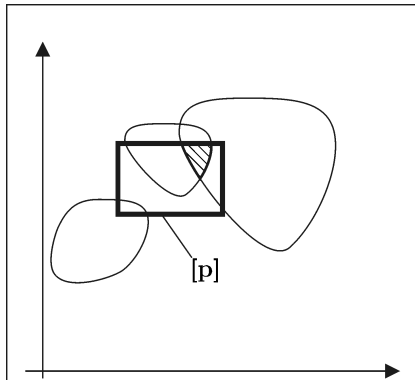


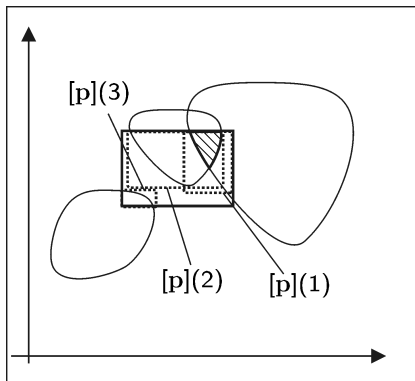
Relaxed intersection

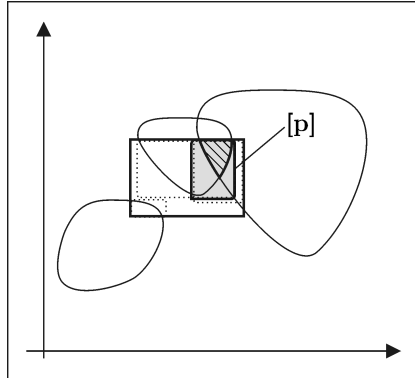
Probabilistic set estimation

Application to localization

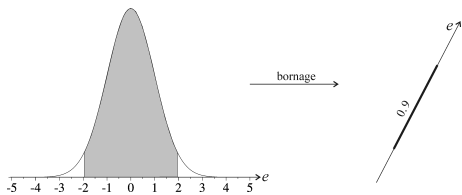
With real data

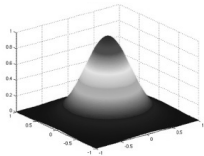




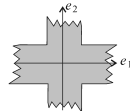
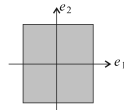
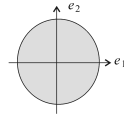


Probabilistic set estimation



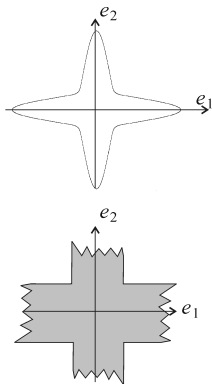


bornage \rightarrow



$$\Pi(\mathbf{e}) \propto \left(\exp(-e_1^2) + \exp\left(-\frac{e_1^2}{10}\right) \right) \\ * \left(\exp(-e_2^2) + \exp\left(-\frac{e_2^2}{10}\right) \right)$$

→ bornage



$$\mathbf{y} = \boldsymbol{\psi}(\mathbf{p}) + \mathbf{e},$$

where

$\mathbf{e} \in \mathbb{E} \subset \mathbb{R}^m$ is the error vector,

$\mathbf{y} \in \mathbb{R}^m$ is the collected data vector,

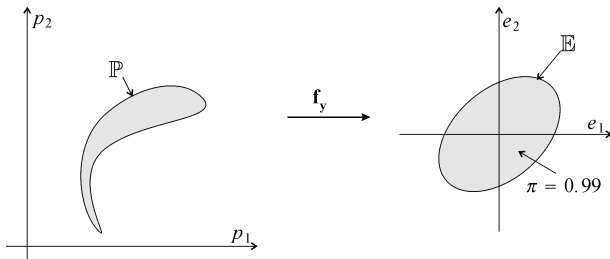
$\mathbf{p} \in \mathbb{R}^n$ is the parameter vector to be estimated.

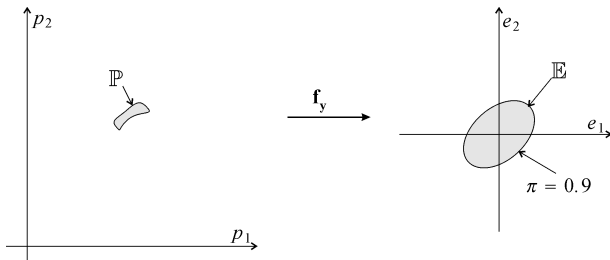
Or equivalently

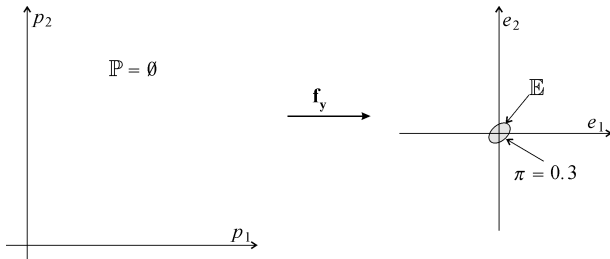
$$\mathbf{e} = \mathbf{y} - \psi(\mathbf{p}) = \mathbf{f}_y(\mathbf{p}),$$

The *posterior feasible set* for the parameters is

$$\mathbb{P} = \mathbf{f}_{\mathbf{y}}^{-1}(\mathbb{E}).$$







Consider the error model

$$\underbrace{\begin{pmatrix} e_1 \\ \vdots \\ e_m \end{pmatrix}}_{=\mathbf{e}} = \underbrace{\begin{pmatrix} y_1 - \psi_1(\mathbf{p}) \\ \vdots \\ y_m - \psi_m(\mathbf{p}) \end{pmatrix}}_{=\mathbf{f}_y(\mathbf{p})}$$

The data y_i is an *inlier* if $e_i \in [e_i]$ and an *outlier* otherwise. We assume that

$$\forall i, \Pr(e_i \in [e_i]) = \pi$$

and that all e_i 's are independent.

Equivalently,

$$\left\{ \begin{array}{ll} y_1 - \psi_1(\mathbf{p}) \in [e_1] & \text{with a probability } \pi \\ \vdots & \vdots \\ y_m - \psi_m(\mathbf{p}) \in [e_m] & \text{with a probability } \pi \end{array} \right.$$

The probability of having k inliers is

$$\frac{m!}{k!(m-k)!} \pi^k \cdot (1-\pi)^{m-k}.$$

The probability of having strictly more than q outliers is thus

$$\gamma(q, m, \pi) = \sum_{k=0}^{m-q-1} \frac{m!}{k!(m-k)!} \pi^k \cdot (1-\pi)^{m-k}.$$

Denote by $\mathbb{E}^{\{q\}}$ the set of all $\mathbf{e} \in \mathbb{R}^m$ consistent with at least $m - q$ error intervals $[e_i]$.

For $m = 3$, we have

$$\begin{aligned} \mathbb{E}^{\{0\}} &= [e_1] \times [e_2] \times [e_3] \\ \mathbb{E}^{\{1\}} &= ([e_1] \cap [e_2]) \cup ([e_2] \cap [e_3]) \cup ([e_1] \cap [e_3]) \\ \mathbb{E}^{\{2\}} &= [e_1] \cup [e_2] \cup [e_3] \\ \mathbb{E}^{\{3\}} &= \mathbb{R}^3. \end{aligned}$$

$$\mathbb{P}\{q\} = \mathbf{f}_{\mathbf{y}}^{-1} \left(\mathbb{E}\{q\} \right) = \bigcap_{i \in \{1, \dots, m\}} f_{y_i}^{-1}([e_i]).$$

Application to localization

A robot measures distances to three beacons.

i	x_i	y_i	$[d_i]$
1	1	3	$[1, 2]$
2	3	1	$[2, 3]$
3	-1	-1	$[3, 4]$

The intervals $[d_i]$ contain the true distance with a probability of $\pi = 0.9$.

The feasible sets associated to each data is

$$\mathbb{P}_i = \left\{ \mathbf{p} \in \mathbb{R}^2 \mid \sqrt{(p_1 - x_i)^2 + (p_2 - y_i)^2} - d_i \in [-0.5, 0.5] \right\},$$

where $d_1 = 1.5, d_2 = 2.5, d_3 = 3.5$.

$$\text{prob}(\mathbf{p} \in \mathbb{P}^{\{0\}}) = 0.729$$

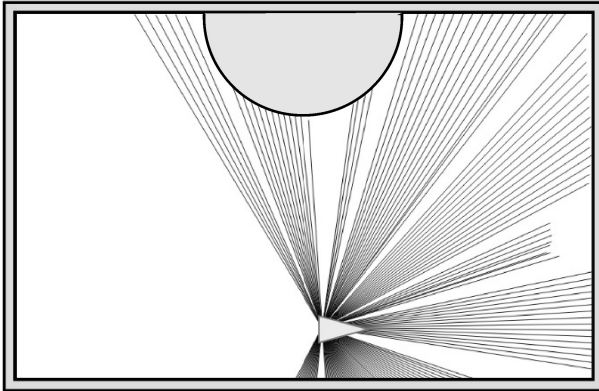
$$\text{prob}(\mathbf{p} \in \mathbb{P}^{\{1\}}) = 0.972$$

$$\text{prob}(\mathbf{p} \in \mathbb{P}^{\{2\}}) = 0.999$$



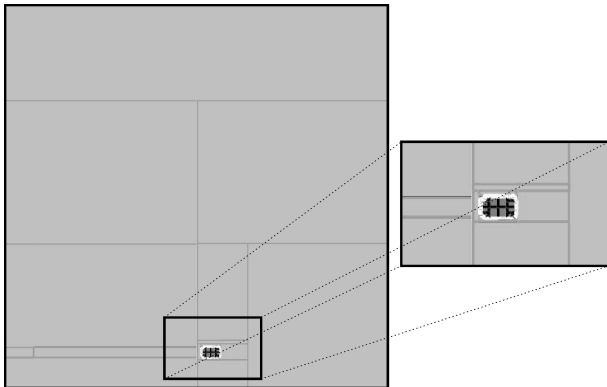
With real data










For $q = 16, m = 143, \pi = 0.95$, the probability of being wrong is

$$\alpha = \gamma(q, m, \pi) = 8.46 \times 10^{-4}.$$



References

- 1 Interval analysis [4, 1, 2]
- 2 Localization with intervals : [3]
- 3 IAMOOC [2]

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