

# Guaranteed Assessment of the Area Explored by an AUV

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October 12, 2016



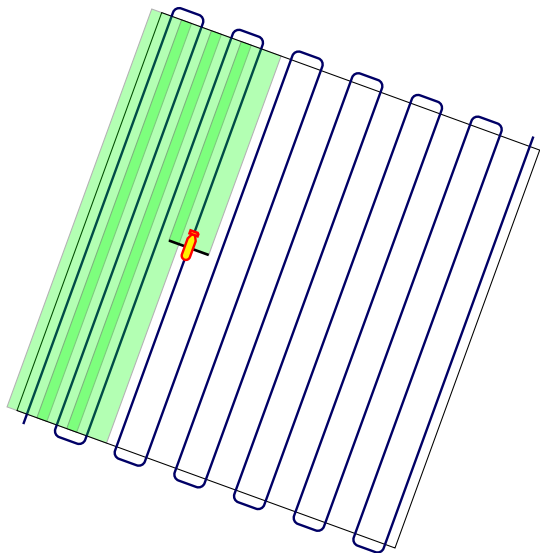
# Outlines

- 1 Problem Statement
- 2 Possible Methods
- 3 Interval Method

# Problem Statement

## Scenario with SSS

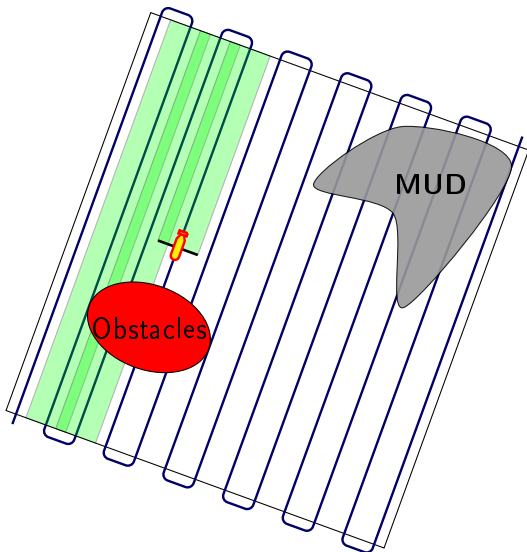
- long mission
- position drift
- unknown environment
- mission re-planning
- no communication



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- long mission
- position drift
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## Issues

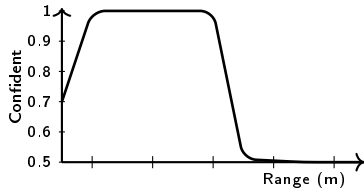
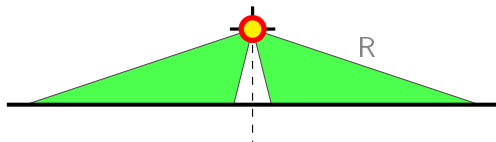
- area explored ?
- data quality ?
- number of view ?



# Side scan sonar

## Hypothesis

- side scan sonar with gapfiller

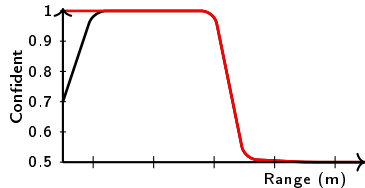
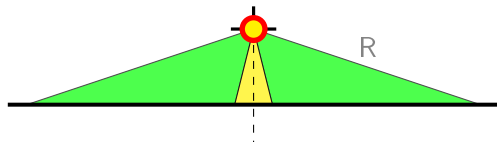


Sonar performance

# Side scan sonar

## Hypothesis

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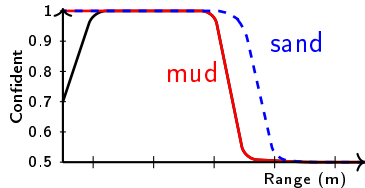
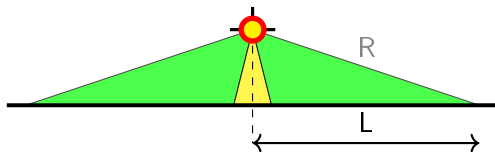


Sonar performance

# Side scan sonar

## Hypothesis

- side scan sonar with gapfiller
- effective range depends of environmental parameters



Sonar performance



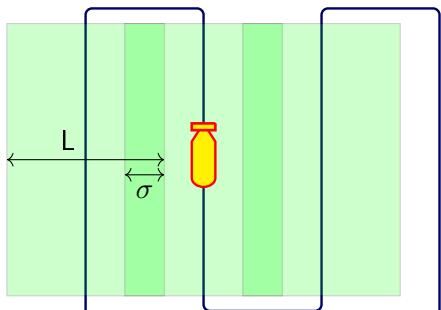
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# Possible Methods

## Geometrical approach

Geometrical footprint of sensors are used to take into account position drift / effective range



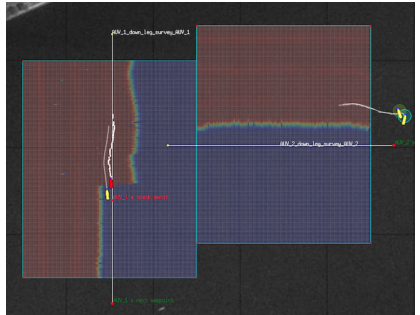
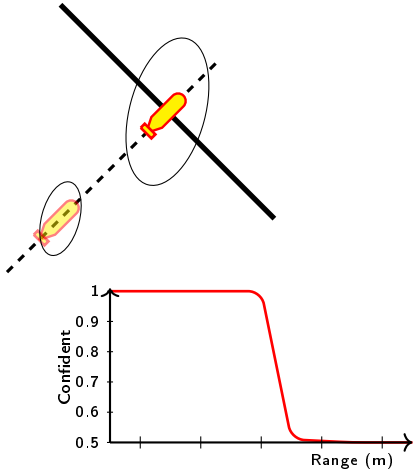
### Remarks

- geometrical assessment
- need a priori knowledge
- easy to use
- not resilient

# Possible methods

Probabilistic (see Liam Paull [PSL14])

) Use probabilistic framework to build an occupancy grid [PSL14]



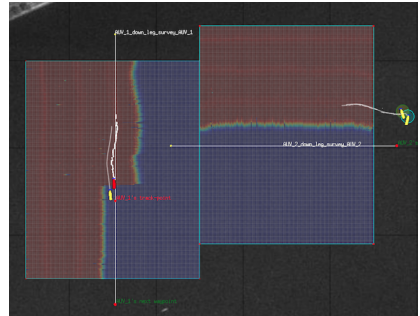
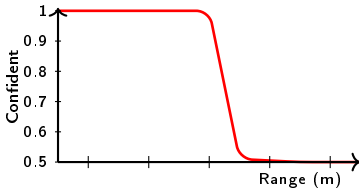
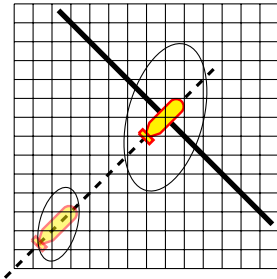
## Remarks

- Tackles with the "mean" case
- Threshold to make decision ?

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## Remarks

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# Possible methods

## Sum up

### Main issues

- manual parameters need to be set
- difficulties to work with multiple views
- no guaranteed

### Proposed approach

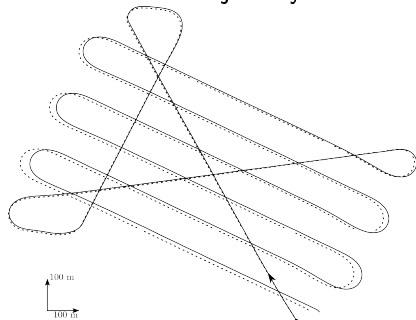
- guaranteed (based on interval arithmetics)
- simple/relevant parameters
- can be used online
- deals with multiple views

# Outlines

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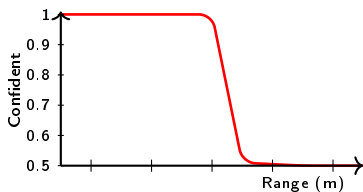
# Bounded error context

AUV trajectory



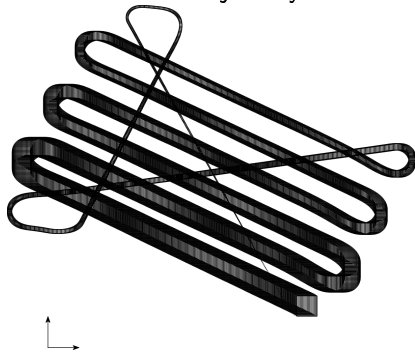
$$\mathbf{x}(\cdot) \in [\mathbf{x}](\cdot)$$

Range of the SSS



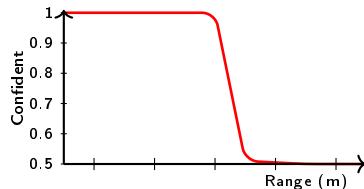
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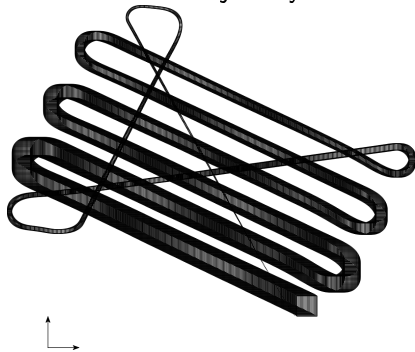
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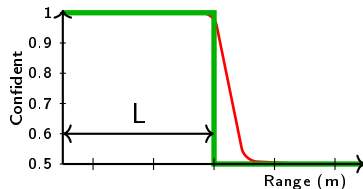
# Bounded error context

AUV trajectory



$$\mathbf{x}(\cdot) \in [\mathbf{x}](\cdot)$$

Range of the SSS



# Thick set inversion problem

Consider :

- an uncertain trajectory  $\mathbf{x}(\cdot) \in [\mathbf{x}](\cdot)$
- $\varphi : \mathbb{R}^p \times \mathbb{R}^n \rightarrow \mathbb{R}$  the *visibility function* (continuous)
- $\psi : \mathbb{R}^p \times \mathbb{R}^n \rightarrow \mathbb{R}$  models the scope of the sensor

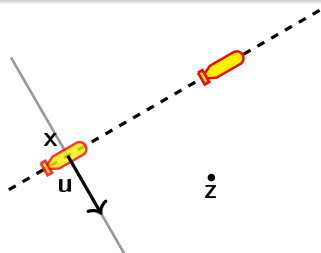
Given  $\mathbf{x} \in [\mathbf{x}](\cdot)$ , we aim at finding the set:

$$\mathbb{Z}(\mathbf{x}) = \{z \in \mathbb{R}^q \mid \varphi(z, \mathbf{x}) = 0 \text{ and } \psi(z, \mathbf{x}) \leq 0\}.$$

# Thick set inversion problem

## Example SSS

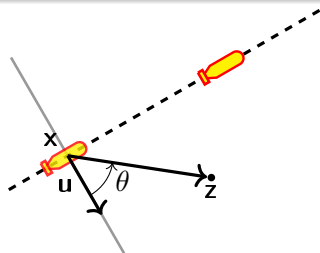
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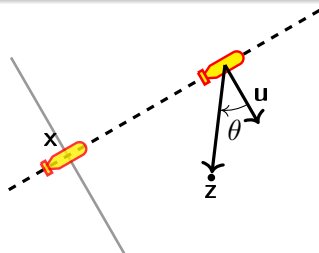
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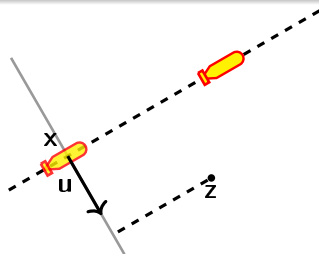


# Thick set inversion problem

## Example SSS

$$\varphi(z, x) = \det(\mathbf{u}, z - x)$$

$$\psi(z, x) = \|\langle \mathbf{u}, (z - x) \rangle\| - L$$



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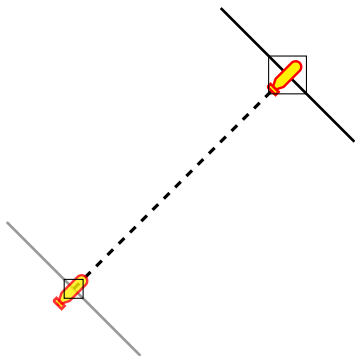
Using interval analysis, can be enclosed by [DJZ13, DJ16]:

$$\mathbb{Z}^- \subset \mathbb{Z} \subset \mathbb{Z}^+.$$

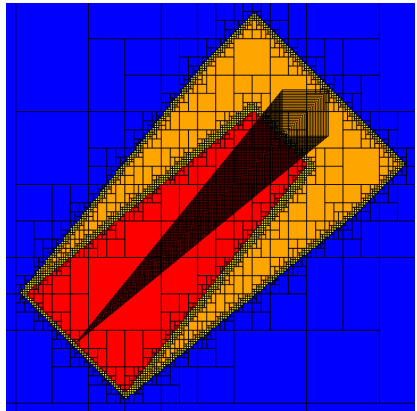
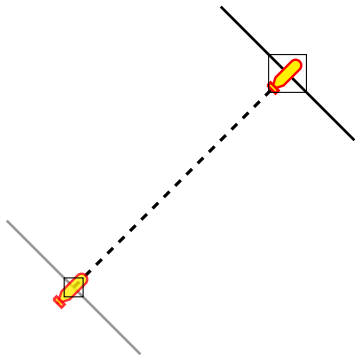
$$\mathbb{Z}^- = \bigcap_{\mathbf{x}(\cdot) \in [\mathbf{x}](\cdot)} \bigcup_{t \geq 0} \mathbb{Z}(\mathbf{x}(t)) \text{ and } \mathbb{Z}^+ = \bigcup_{\mathbf{x}(\cdot) \in [\mathbf{x}](\cdot)} \bigcup_{t \geq 0} \mathbb{Z}(\mathbf{x}(t)).$$



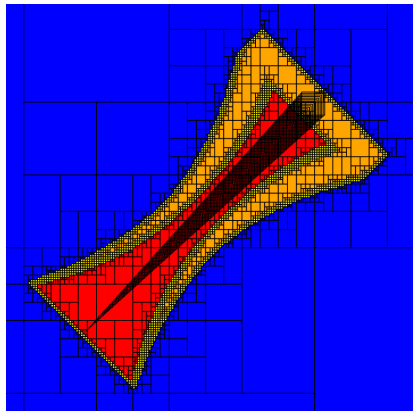
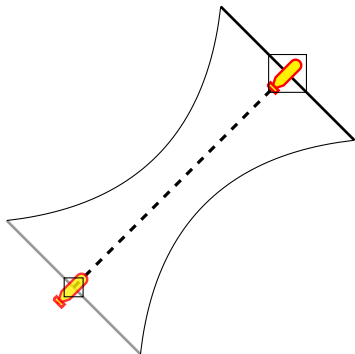
# Example



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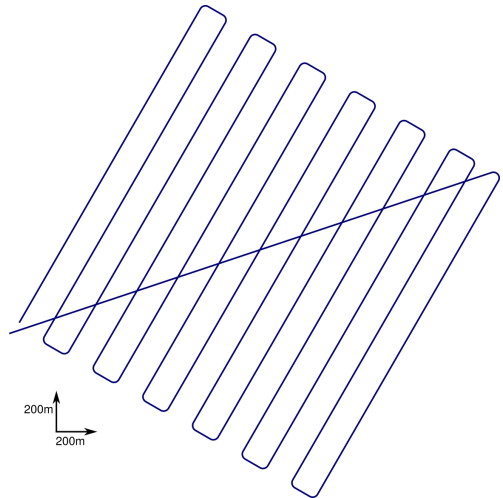


# Bigger example

- 5 hours of mission
- error model  $\delta_x = 0.2 * \sqrt{t}$
- track to track dist. = 150 m
- effective range 120 m



*Daurade*

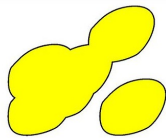
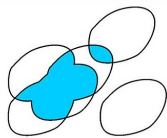
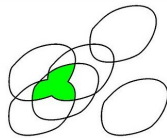
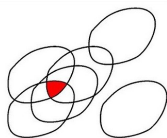


# Relaxed Intersection

## Definition

Given  $m$  subsets  $\mathbb{X}_1, \dots, \mathbb{X}_m$  of  $\mathbb{R}^n$ .

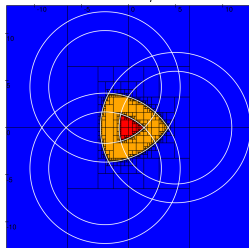
The  $q$ -relaxed intersection, denoted by  $\mathbb{X}^{\{q\}} = \bigcap^{\{q\}} \mathbb{X}_i$ , is the set of all  $x \in \mathbb{R}^n$  which belong to all  $\mathbb{X}_i$ 's, except  $q$  at most.



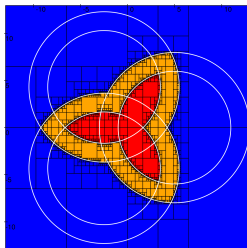
# Relaxed Intersection

## Examples

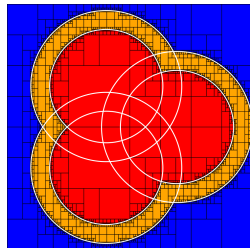
Given 3 disks,



$[[X\{0\}]]$

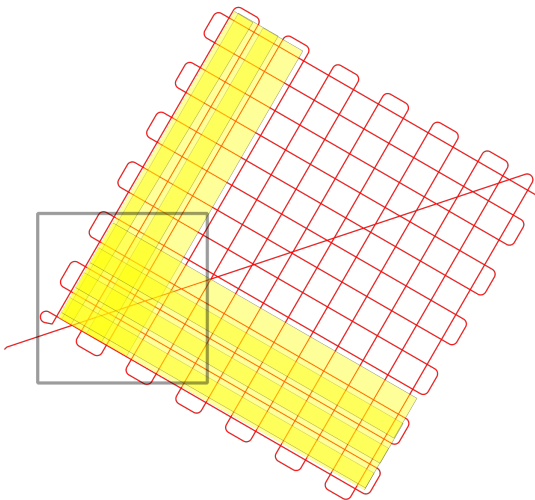


$[[X\{1\}]]$

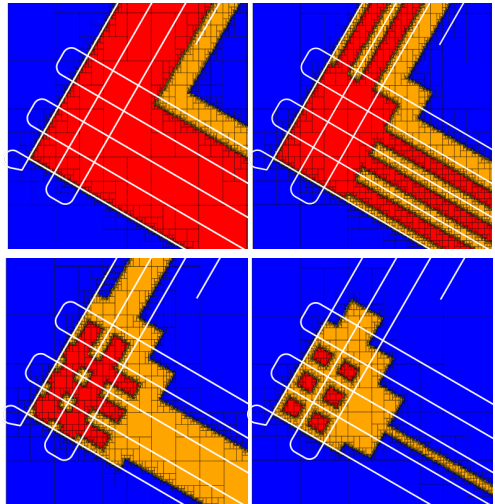
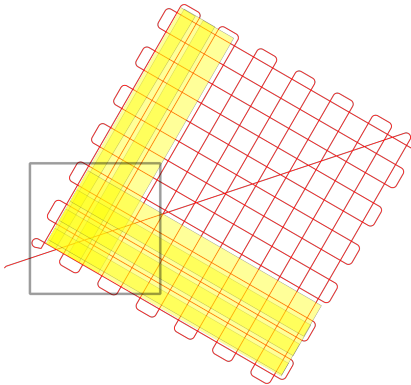


$[[X\{2\}]]$

# Illustration



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## Sum up

Interval methods are powerful to tackle with uncertainty

The proposed interval method

- real time computing
- simple hypothesis
- continuous representation of the space
- useful for small AUV with strong pose uncertainty

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Thank you !

## References



B. Desrochers and L. Jaulin.

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