

# Learning a 2-Manifold with a Boundary in $\mathbb{R}^3$

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# Reconstructing surfaces with a boundary is not trivial but efficiently solvable

- How to reconstruct an unknown surface (without a boundary) from a point set?
- Why is handling a boundary not trivial?
- How can the problems arising from the presence of a boundary can be solved efficiently?

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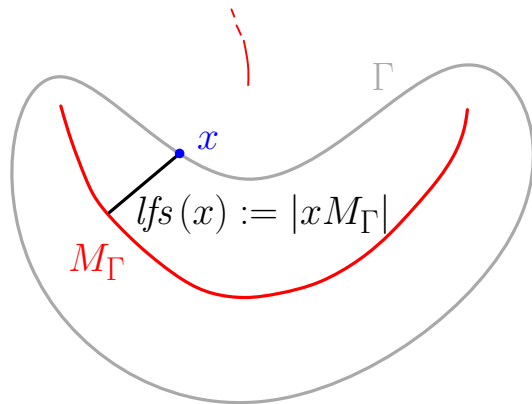
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# Reconstructing surfaces with a boundary is not trivial but *efficiently solvable*

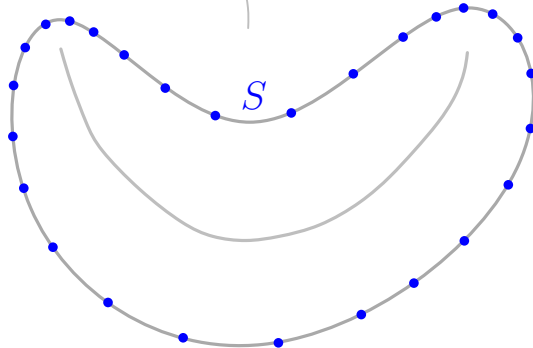
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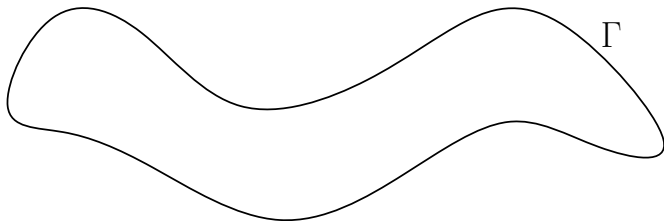
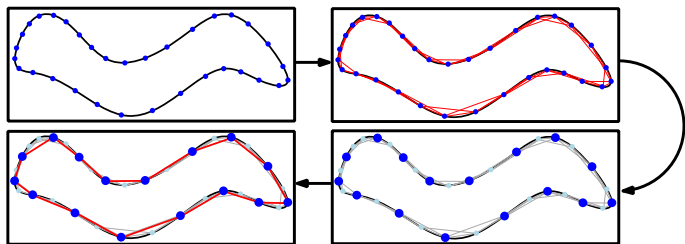
## Local feature size



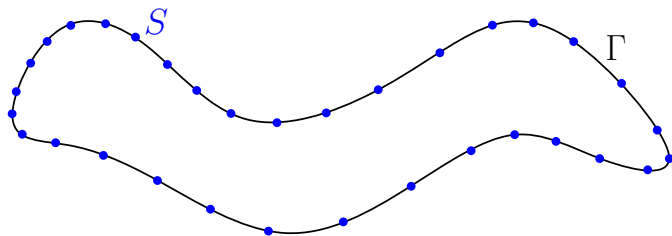
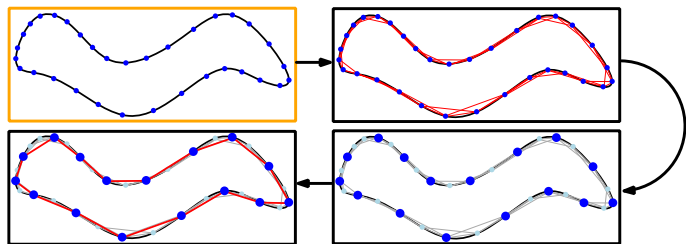
$\varepsilon$ -sample

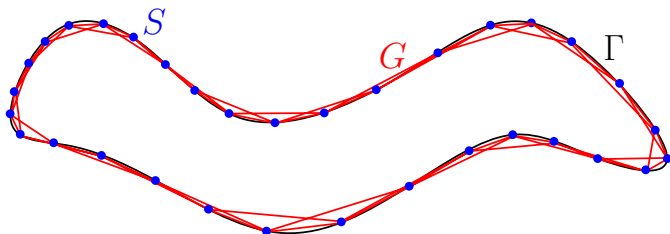
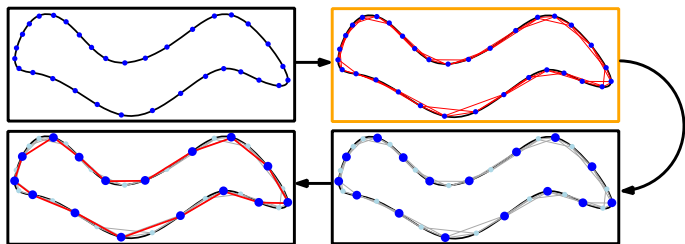
$$\forall x \in \Gamma : \exists s \in S : |xs| \leq \varepsilon \cdot lfs(x)$$

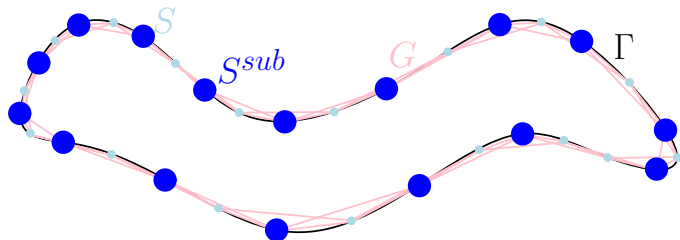
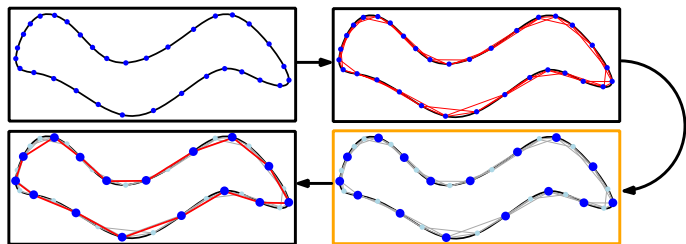


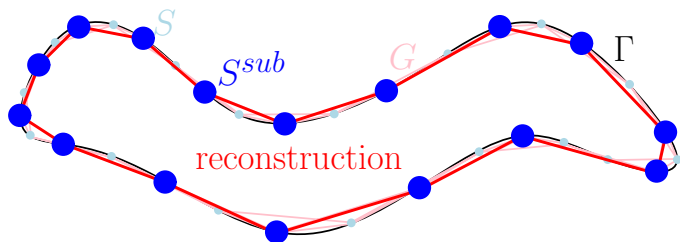
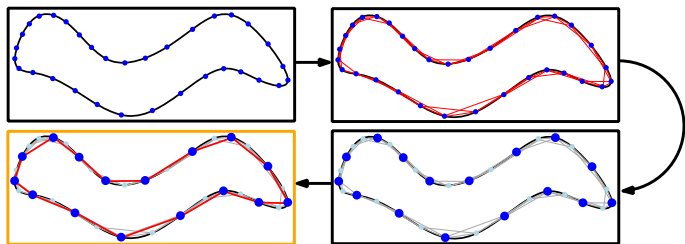
Comb. Reconstruction algorithm from Dumitriu *et al.*

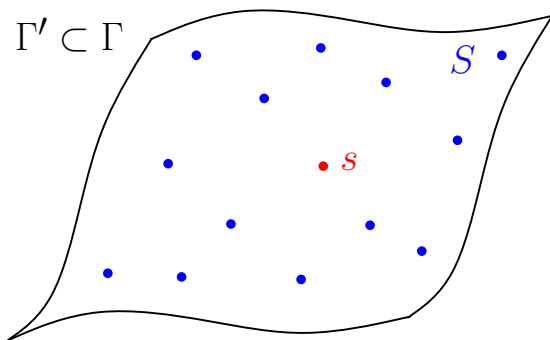


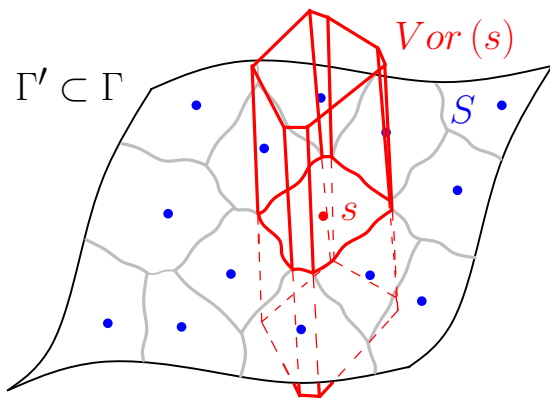
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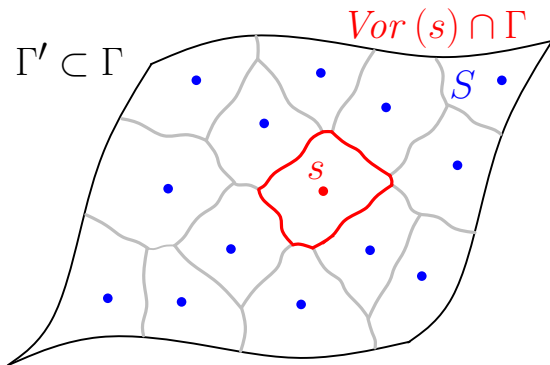
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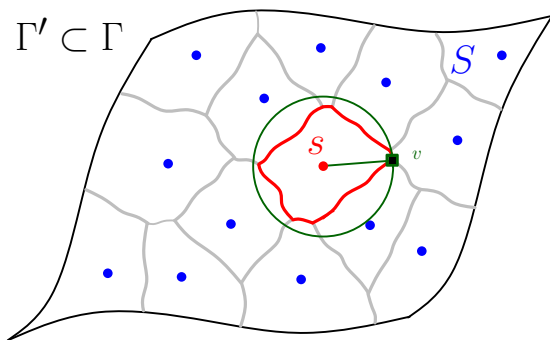
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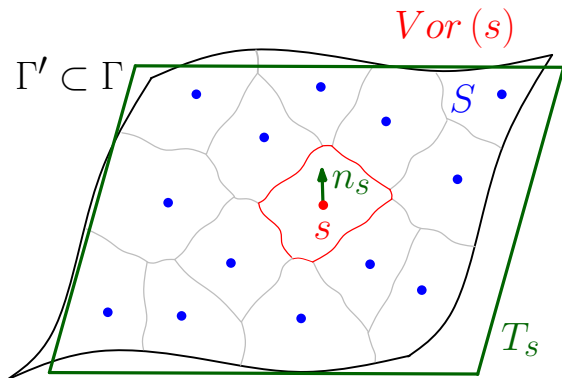
Computing the neighbourhoodgraph  $G$  - w/o boundary

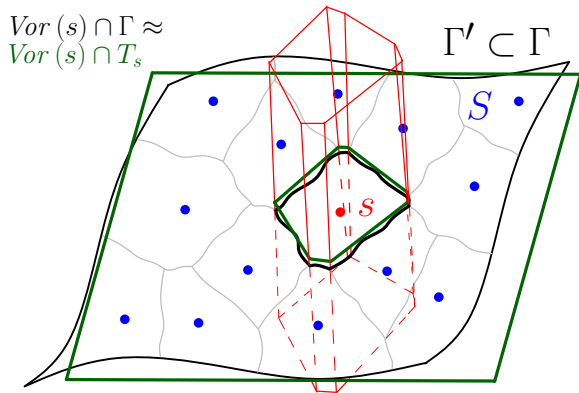
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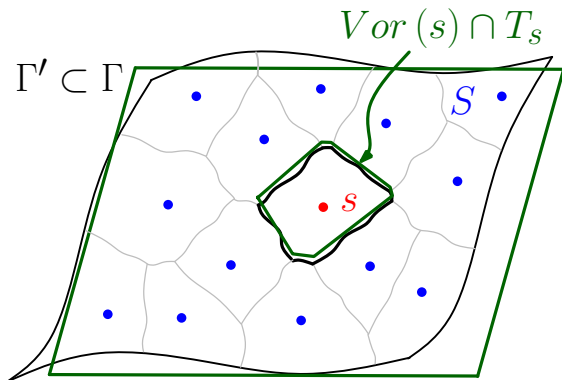
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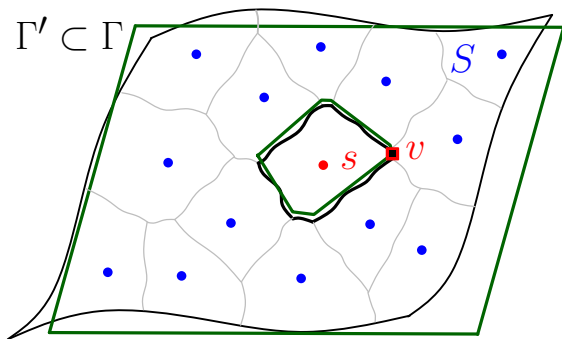
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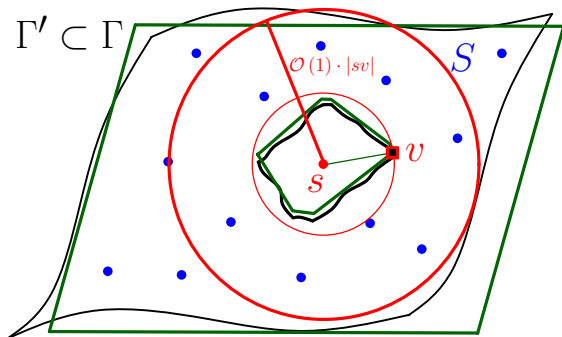


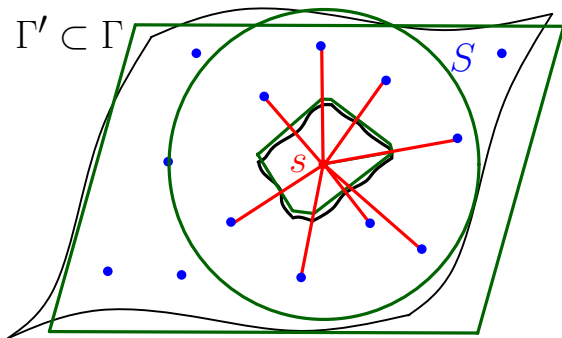
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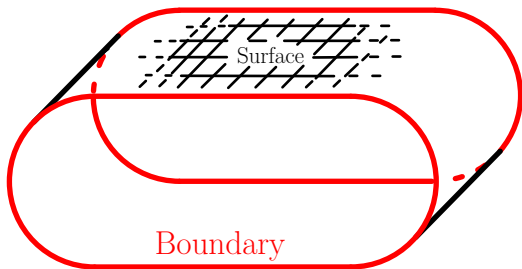
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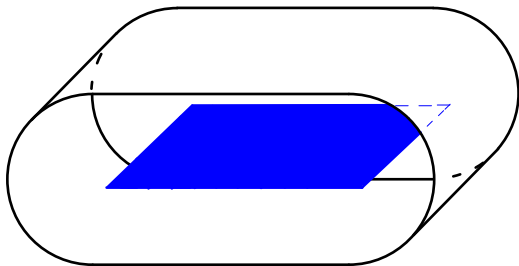
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## Extending the definition of the local feature size



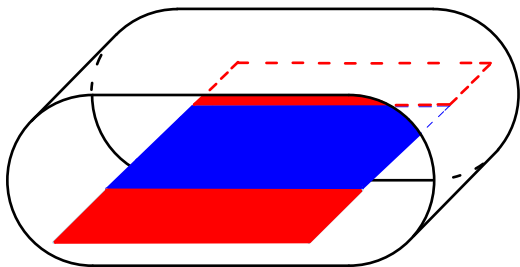
## Extending the definition of the local feature size



$$lfs(x) = \text{dist}(x, M_\Gamma)$$



# Extending the definition of the local feature size

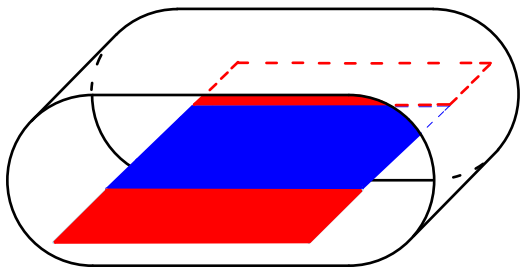


$$lfs(x) = \min \{ \text{dist}(x, M_\Gamma), lfs_{\partial\Gamma}(x) \},$$

with

$$lfs_{\partial\Gamma} = \left\{ \begin{array}{ll} \min_{y \in \partial\Gamma} \{ \text{dist}(y, M_{\partial\Gamma}) + |xy| \} & x \in \Gamma^\circ \\ \text{dist}(x, M_{\partial\Gamma}) & x \in \partial\Gamma \end{array} \right\}$$

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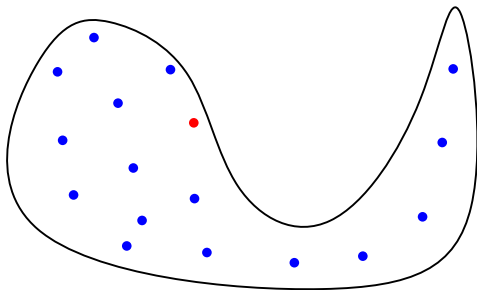


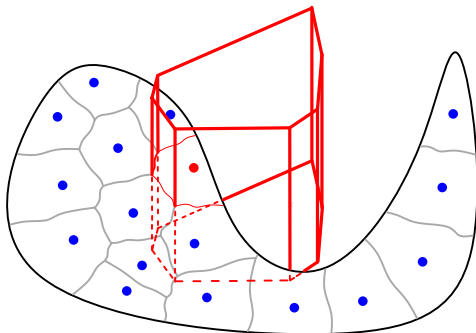
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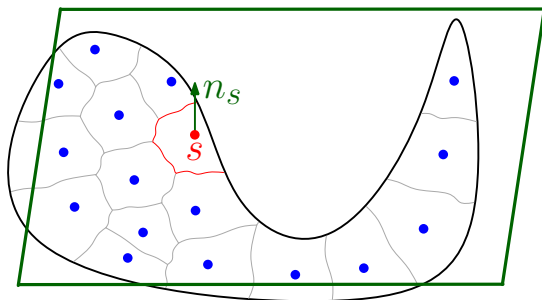
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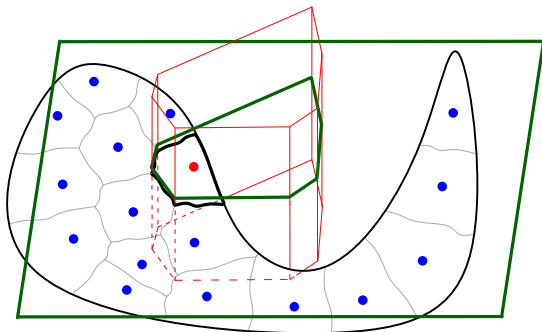
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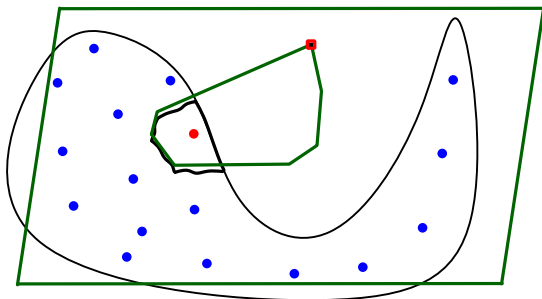
“good” properties are maintained

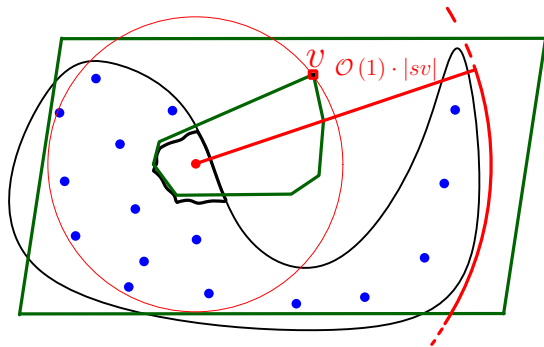
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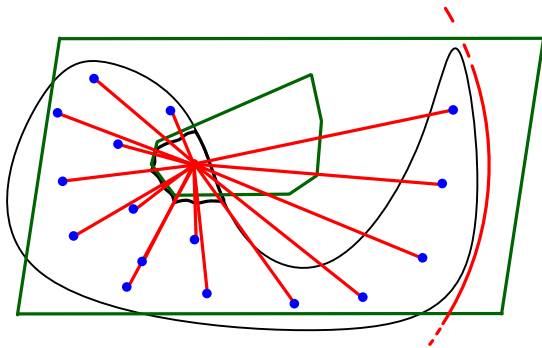
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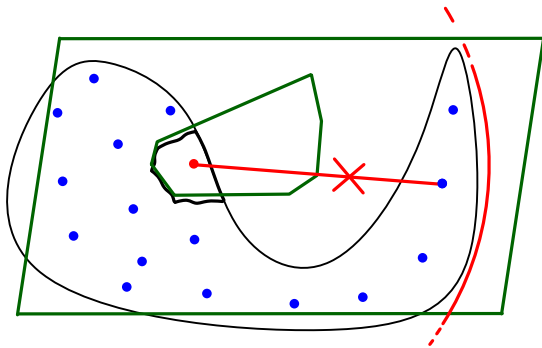
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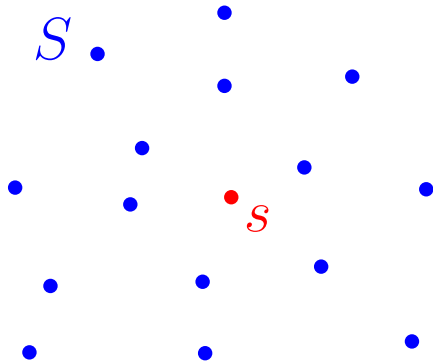
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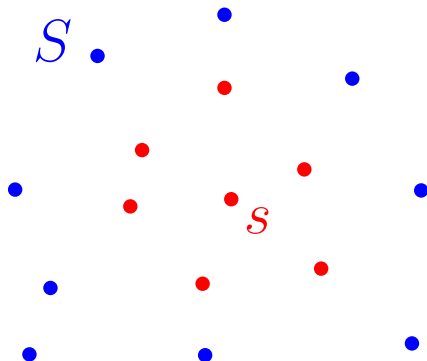
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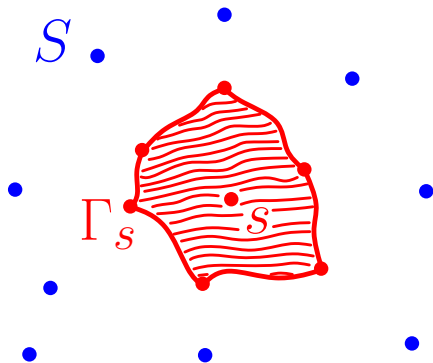
# Approach - high level view



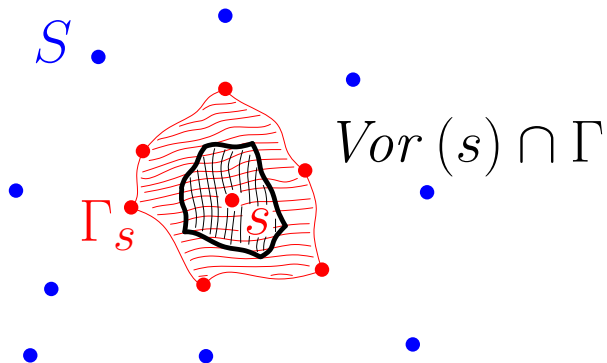
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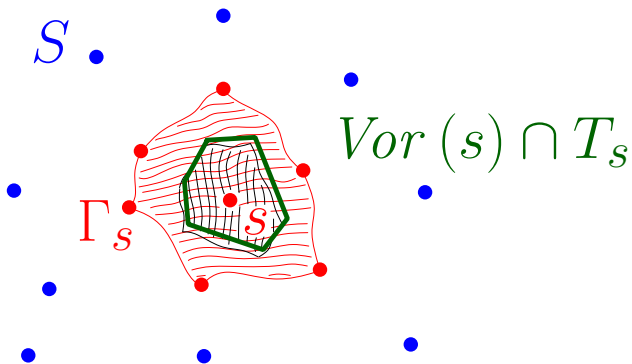
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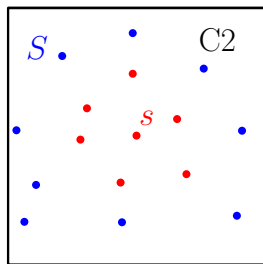
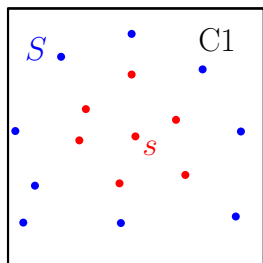
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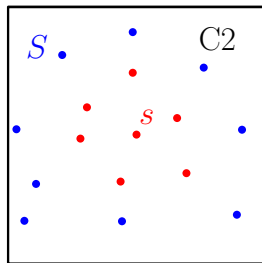
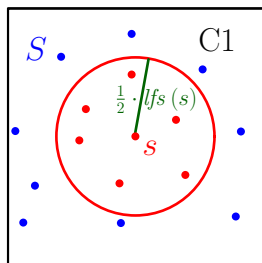


## Approach - low level view

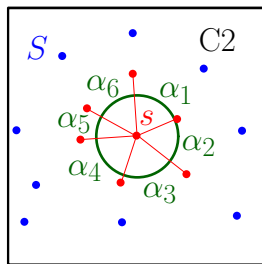
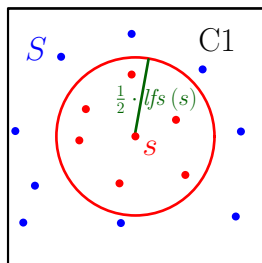




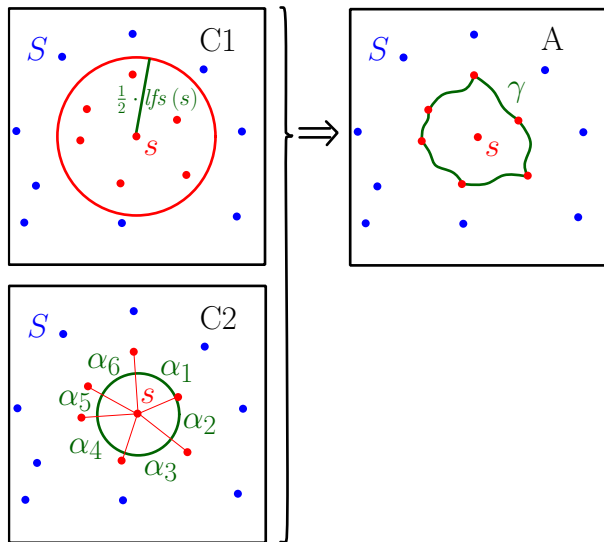
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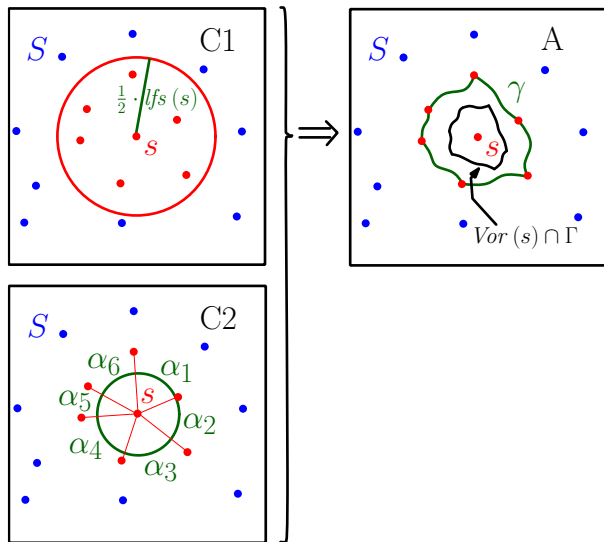
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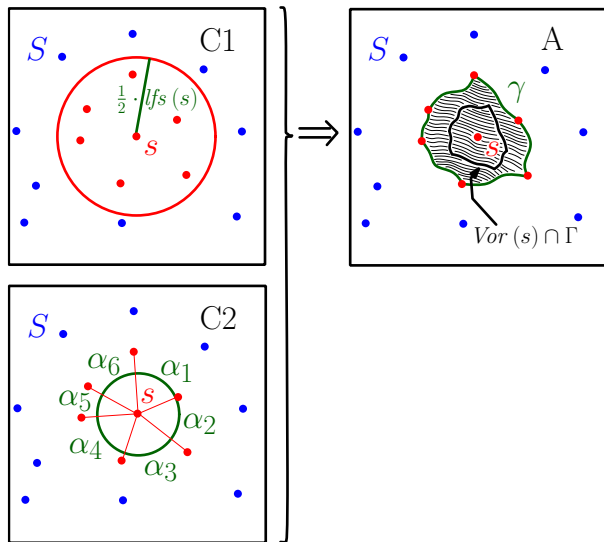
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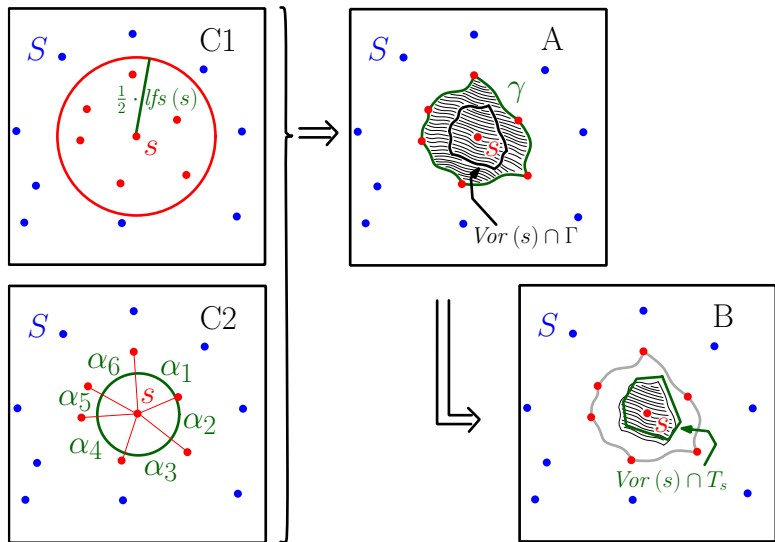
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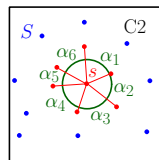
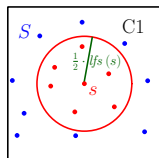
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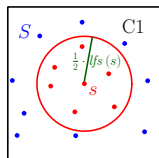
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# Recognition of C1 and C2

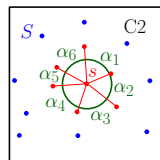


# Recognition of C1 and C2



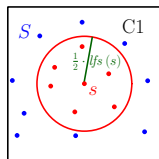
## Theorem

Can compute  $k_\epsilon \in \Omega\left(\frac{1}{\epsilon^2}\right)$ , s.t.  $k_\epsilon$  nearest neighbours of  $s \in S$  are inside  $B\left(x, \frac{1}{2} \cdot lfs(s)\right)$





# Recognition of C1 and C2



## Theorem

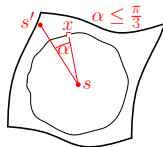
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Require new sampling condition for  $k \leq k_\epsilon$   
 $\rightarrow (\epsilon, k)$ -sample

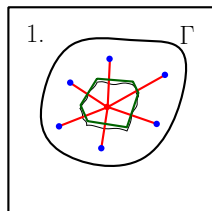
## Definition

$(\epsilon, k)$ -sample is an  $\epsilon$ -sample  $S$  with  $\forall s \in S$ :

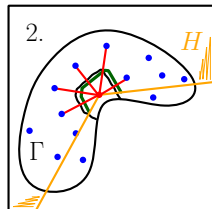
$$\forall x \in \Gamma : |xs| = \frac{\epsilon}{1-\epsilon} \cdot lfs(s) : \\ \exists s' \in \{NN_1, \dots, NN_k\} : \angle(xss') \leq \frac{\pi}{3}$$



# Implications of working with an $(\varepsilon, k)$ -sample

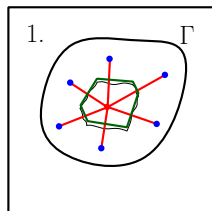


$$|s\partial\Gamma| \geq 2 \cdot \varepsilon \cdot lfs(s) :$$



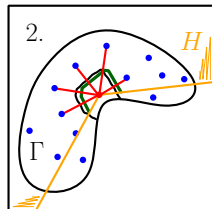
$$|s\partial\Gamma| < 2 \cdot \varepsilon \cdot lfs(s) :$$

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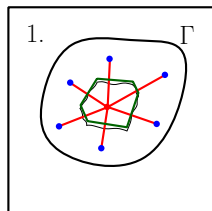
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Can prove: edges are superset of Delaunay-edges restricted to  $\Gamma$



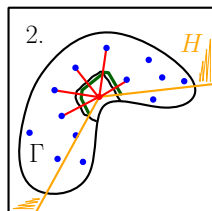
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# Implications of working with an $(\varepsilon, k)$ -sample



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Can prove: edges are superset of Delaunay-edges restricted to  $\Gamma$



$$|s\partial\Gamma| < 2 \cdot \varepsilon \cdot \text{Ifs}(s) :$$

Can compute "wedge"  $H$ , s.t.

$\partial\Gamma$  avoids  $\text{Vor}(s) \cap \Gamma \cap H$

$$\implies \text{Vor}(s) \cap T_s \cap H \approx \text{Vor}(s) \cap \Gamma \cap H$$

## SURFACE RECONSTRUCTION

Boundary leads to problems...

... which can be solved by  
safeguarding from the boundary...

... guaranteed by assuming  
an  $(\varepsilon, k)$ -sample.

In the full paper:

Analysis of the runtime:  $\mathcal{O}(n \cdot \log(n))$

Analysis of the approximation quality:  $|e| \leq 1.135 \cdot c \cdot \varepsilon \cdot lfs$

