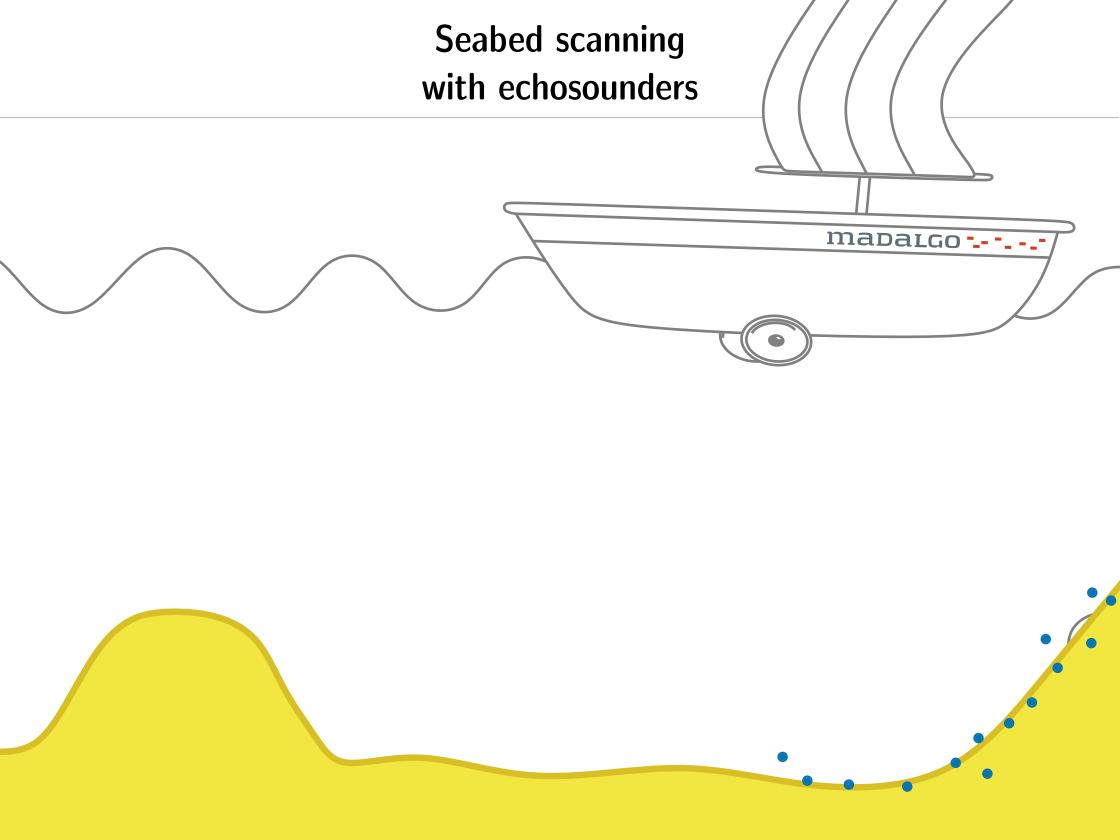
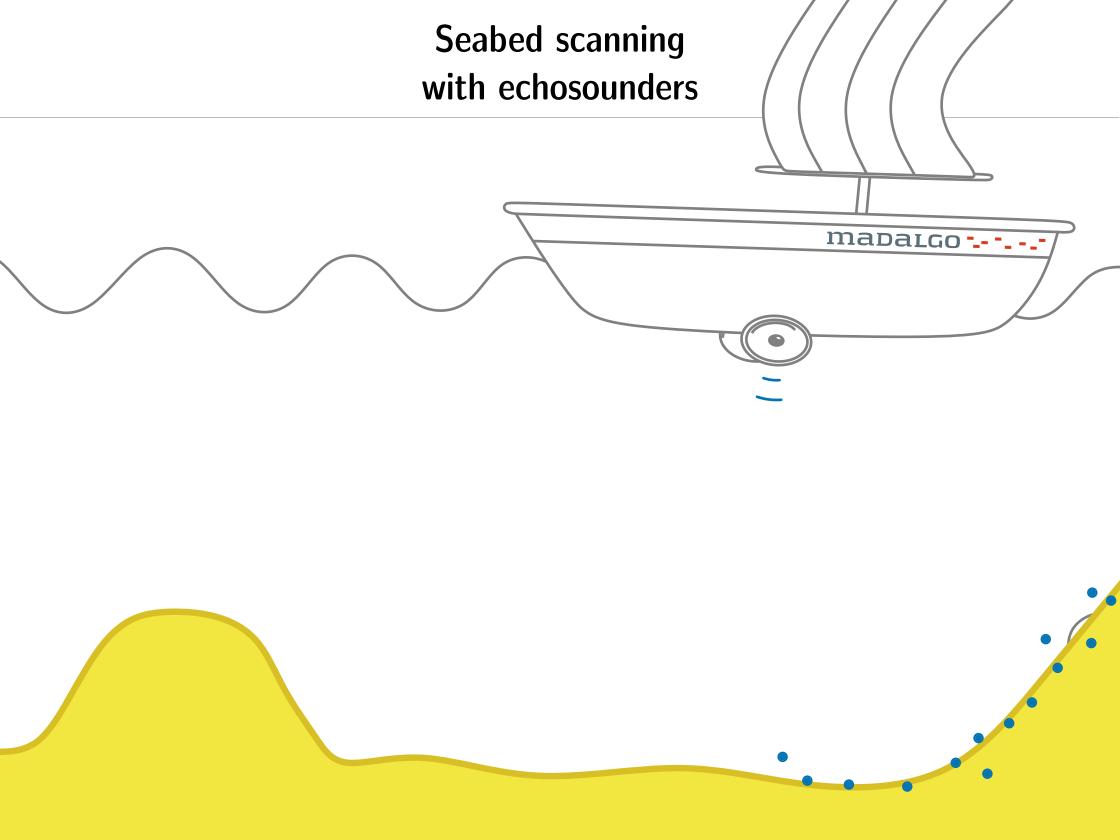
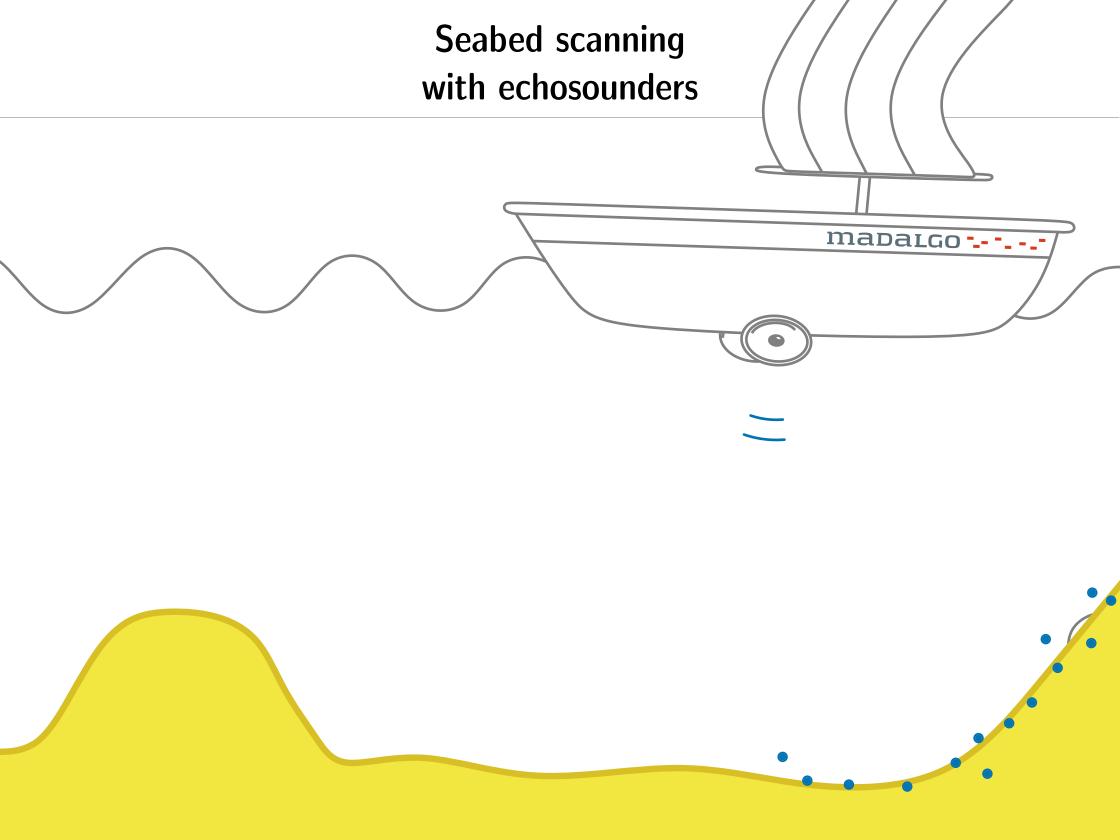


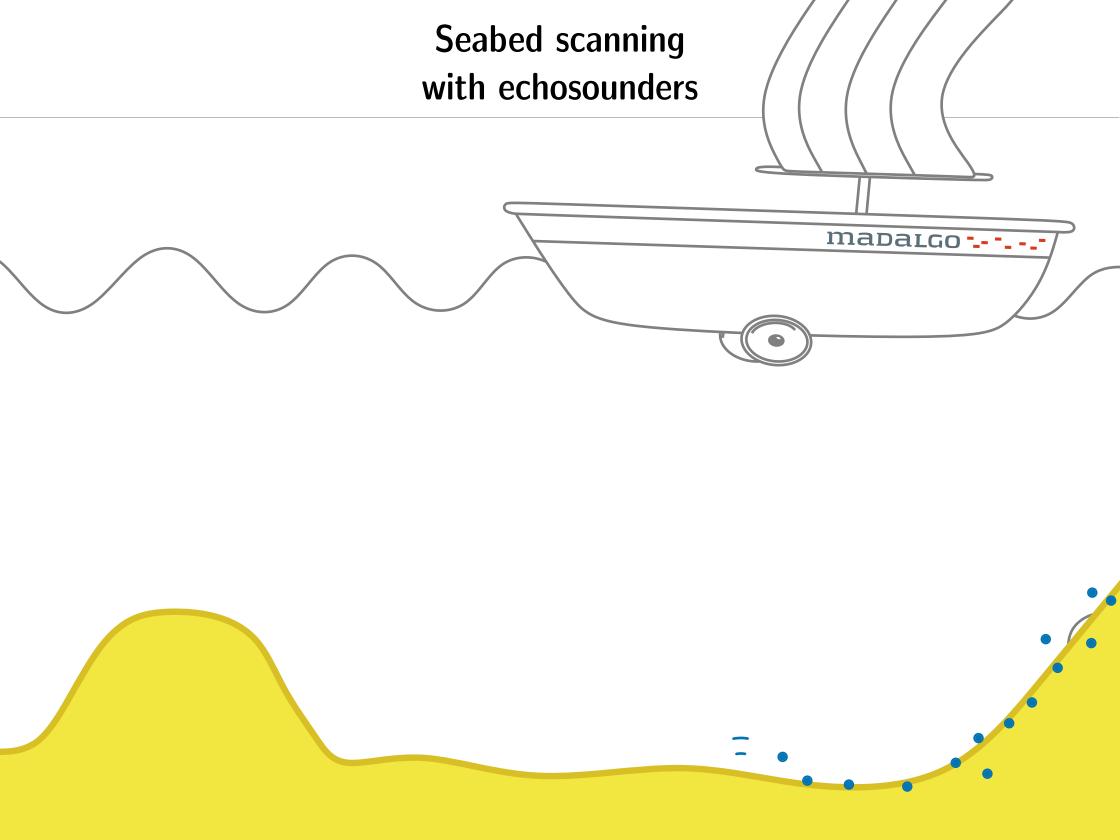
Cleaning massive sonar point clouds

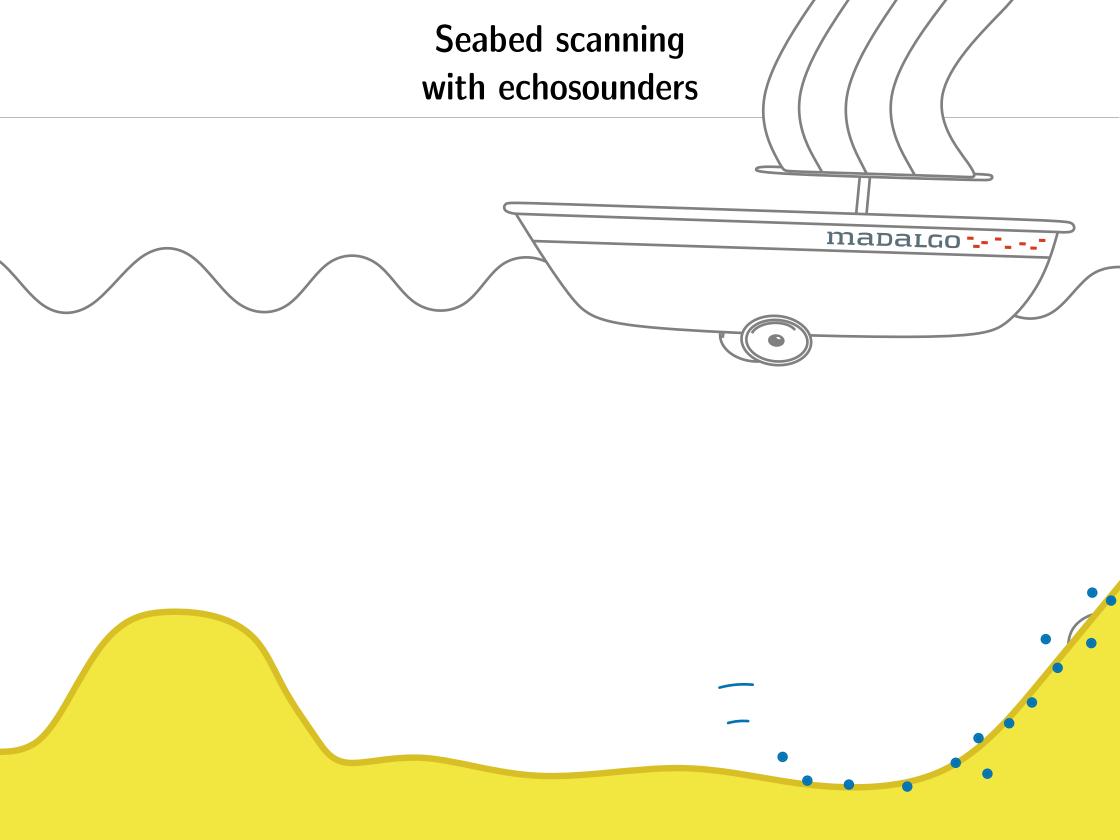


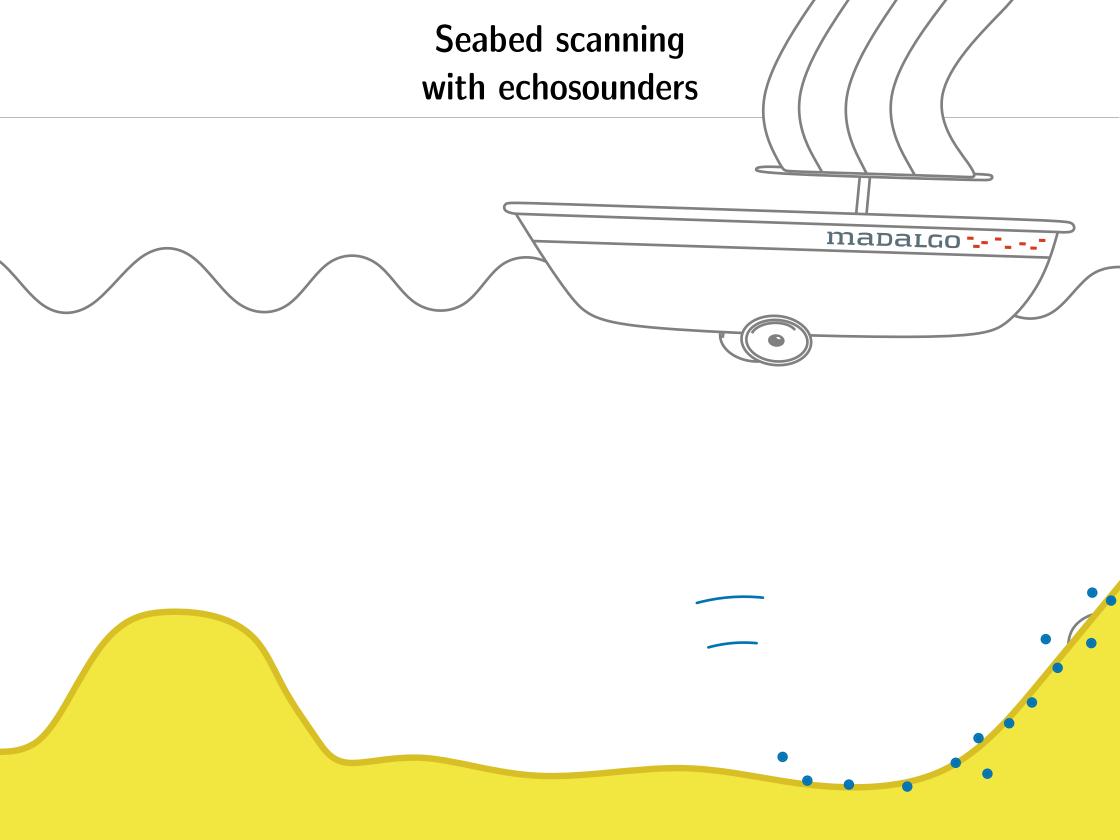


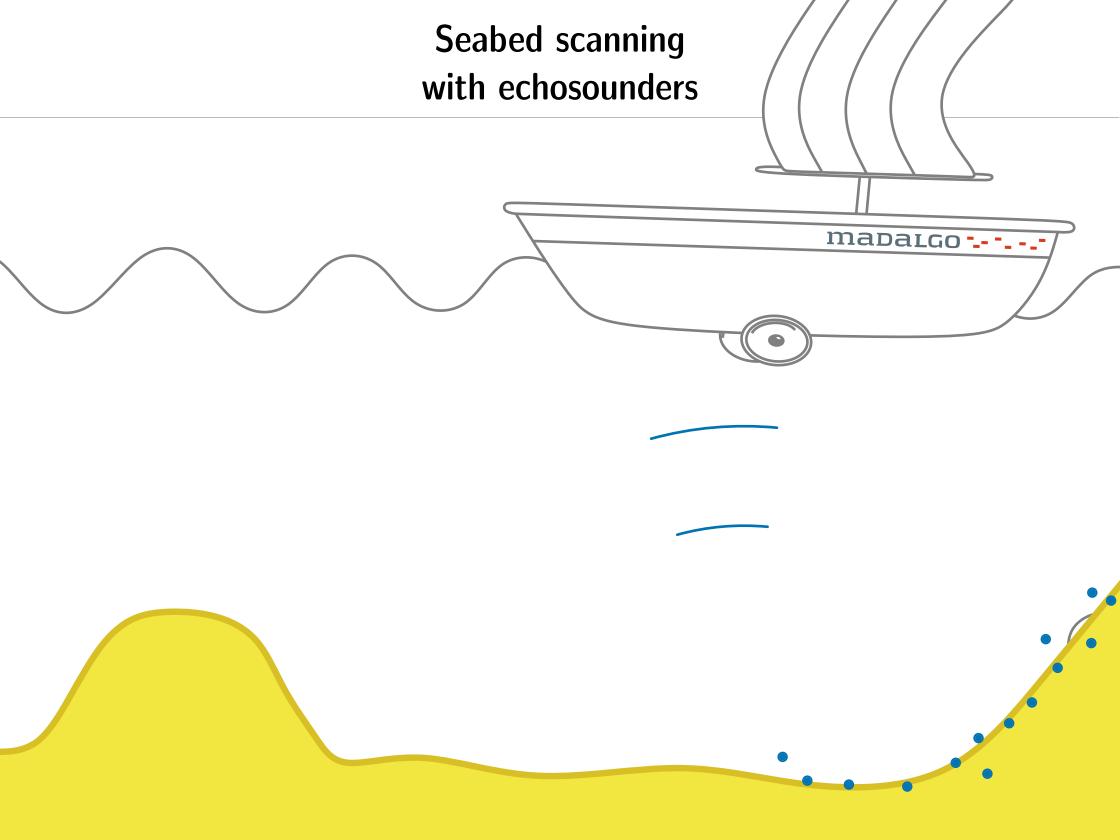


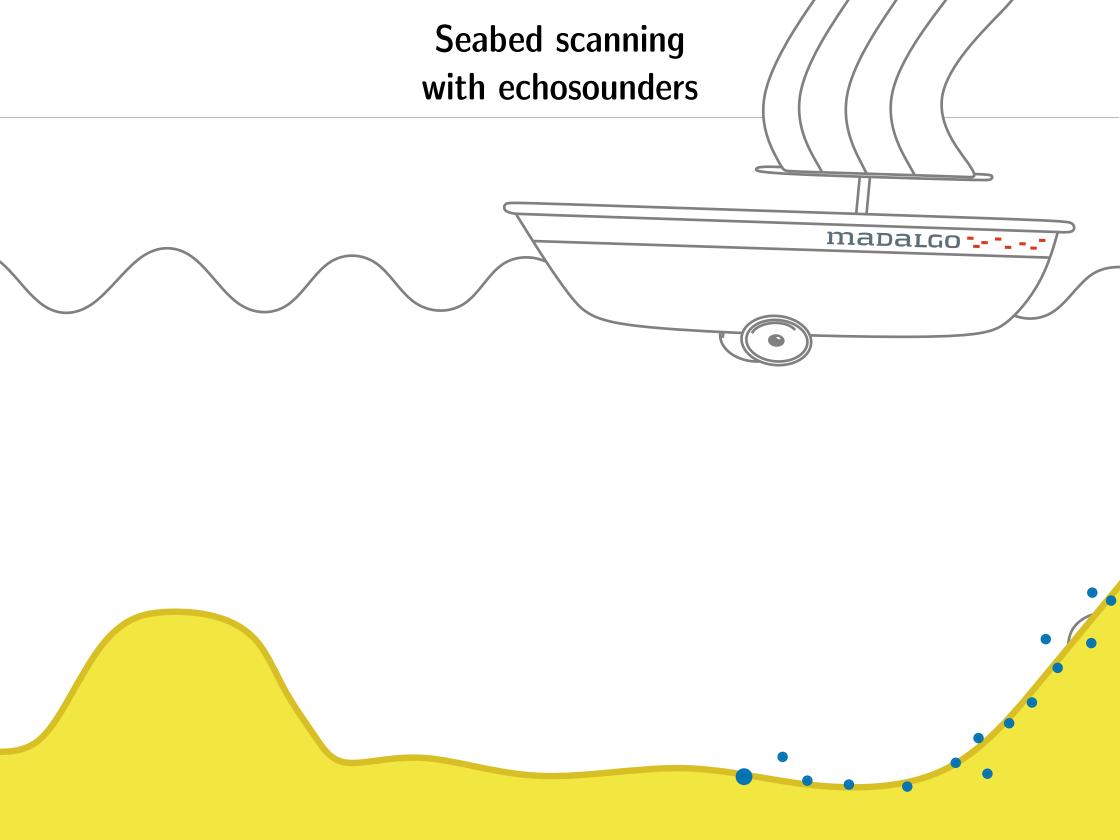


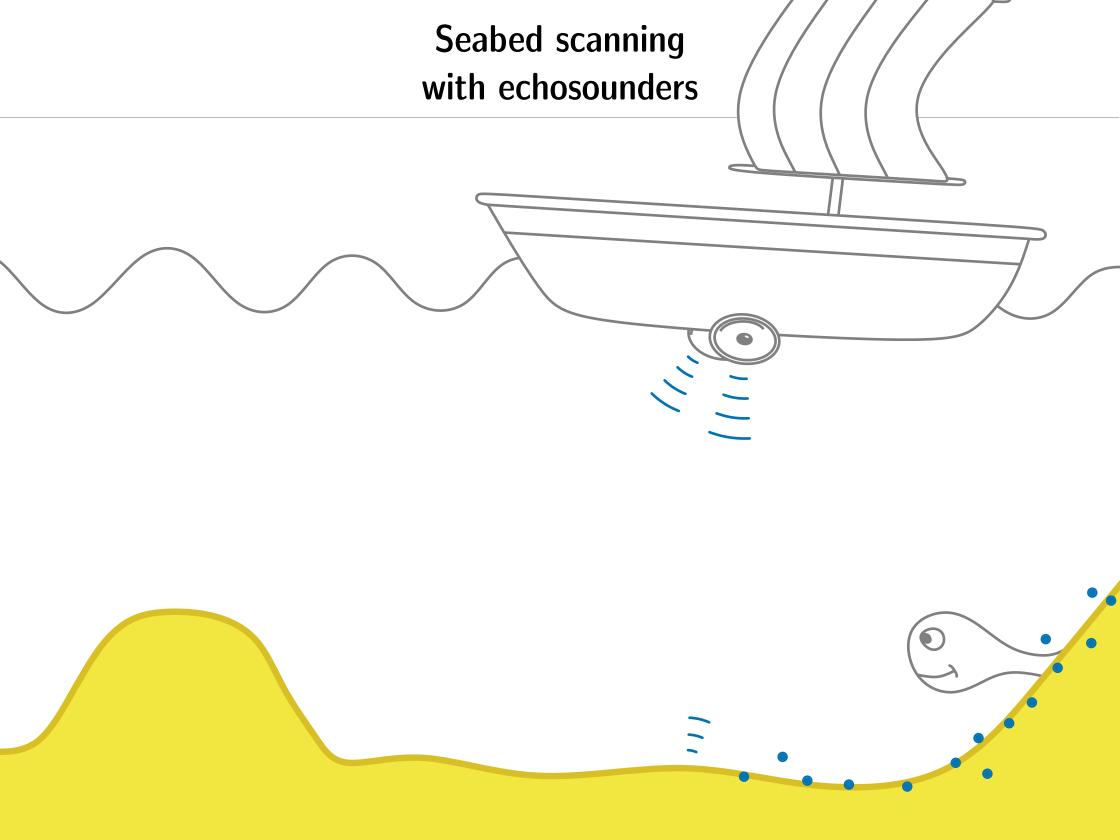


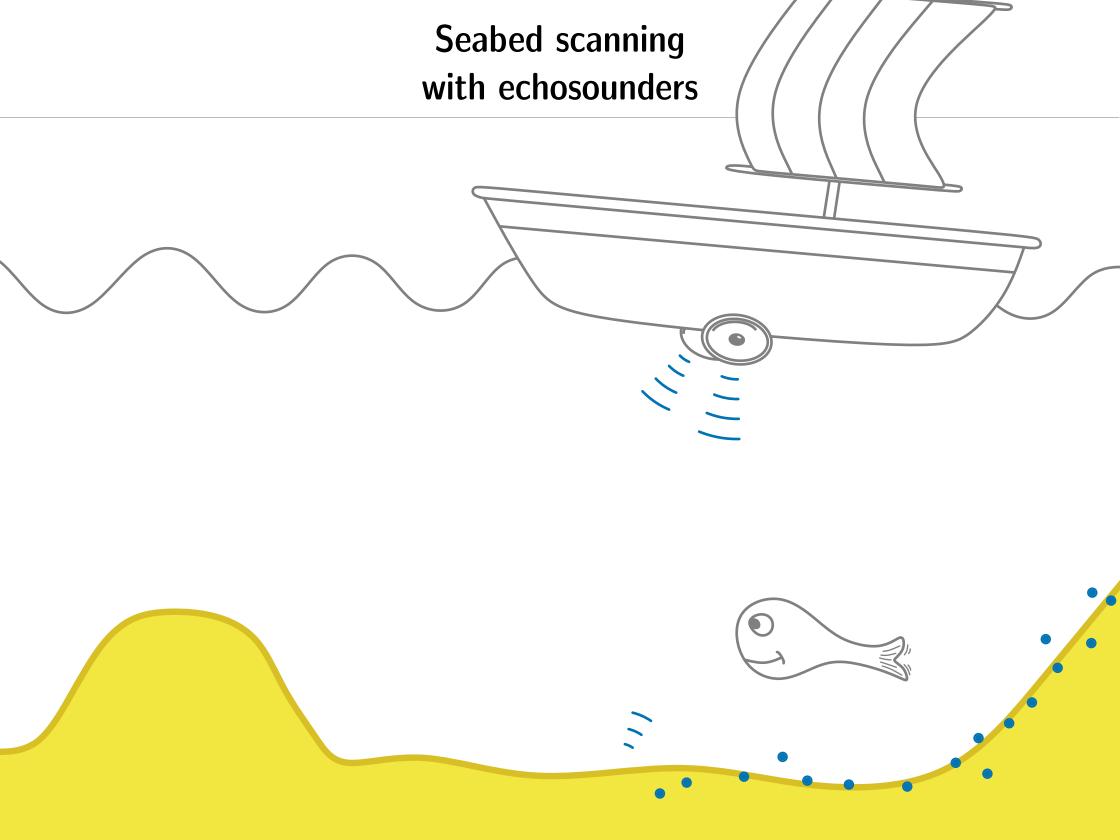


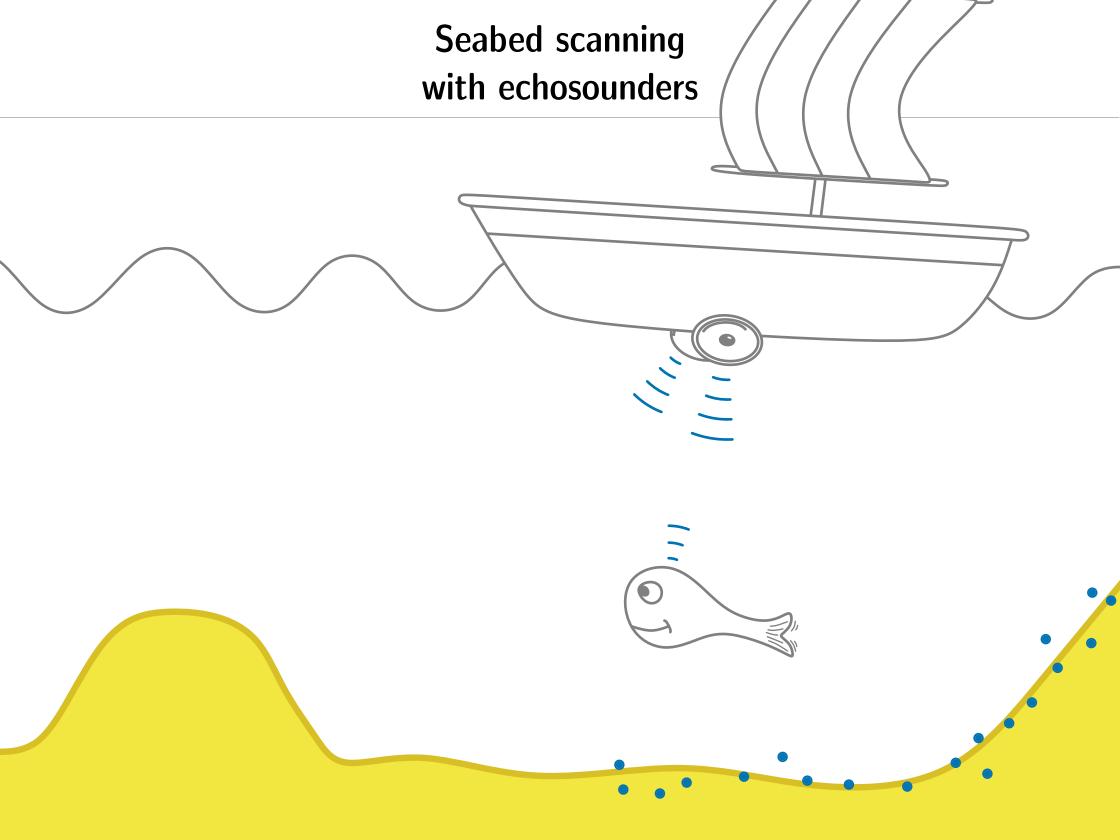


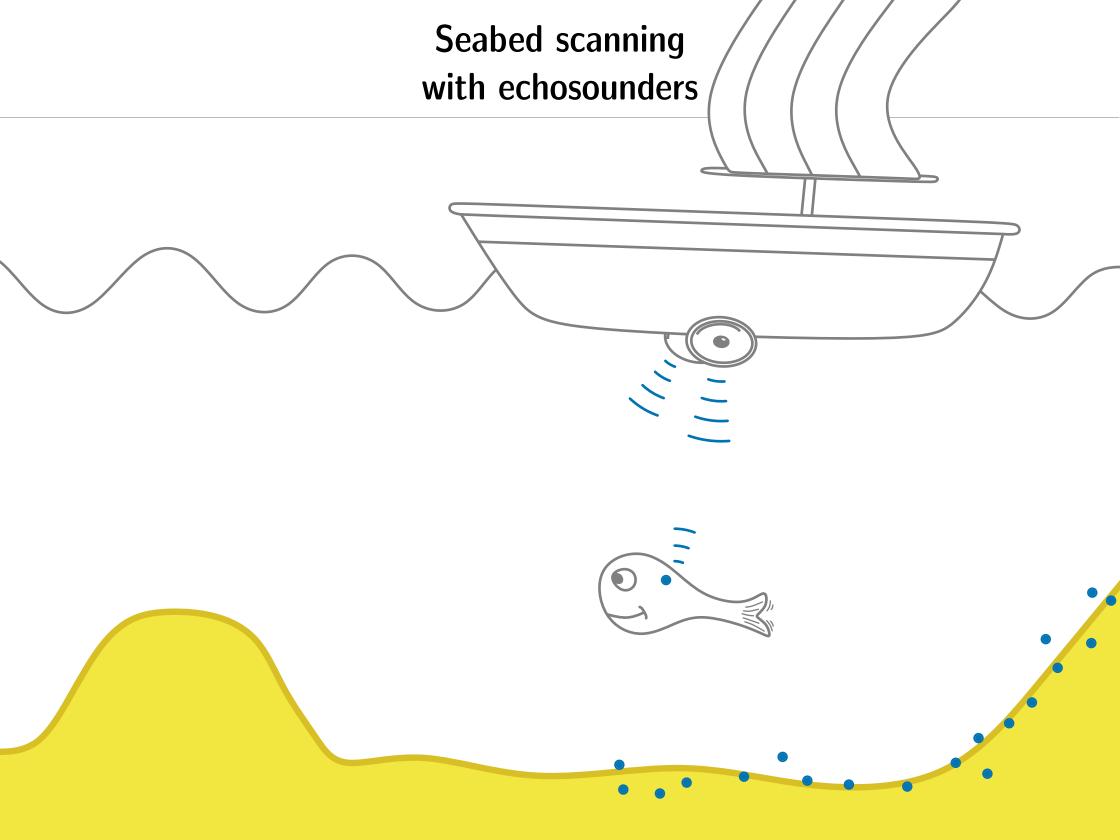


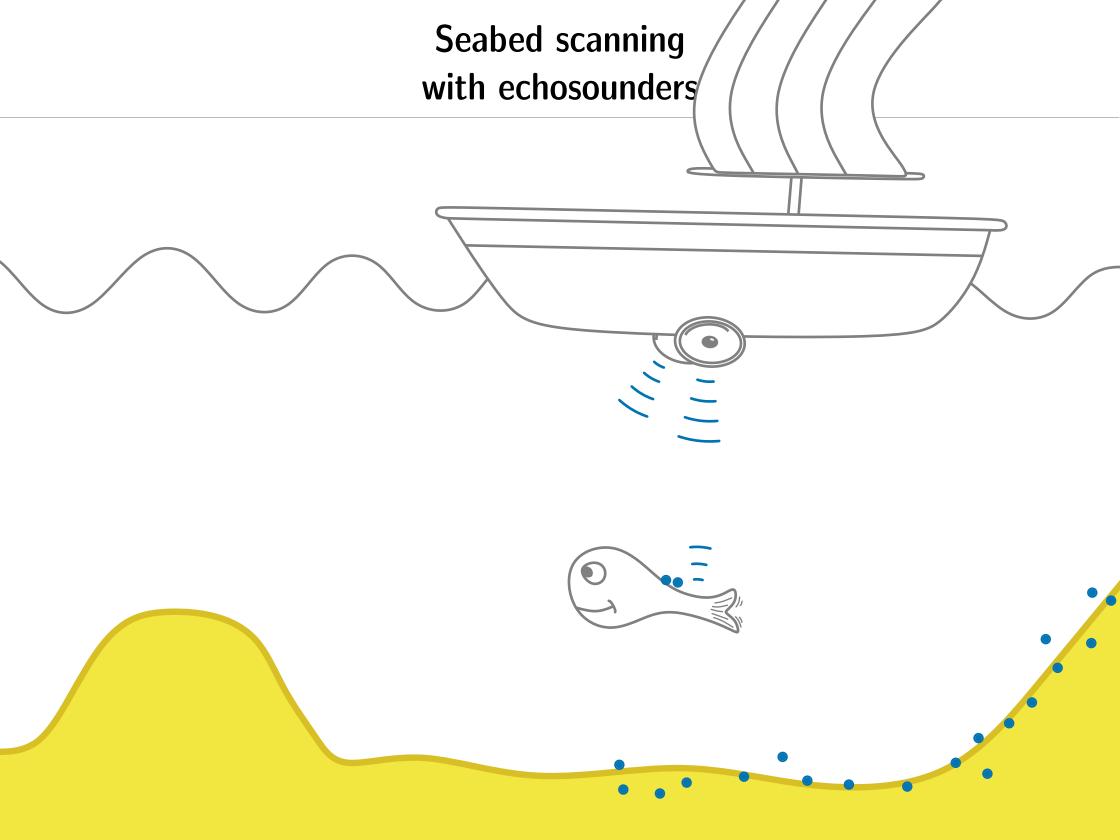


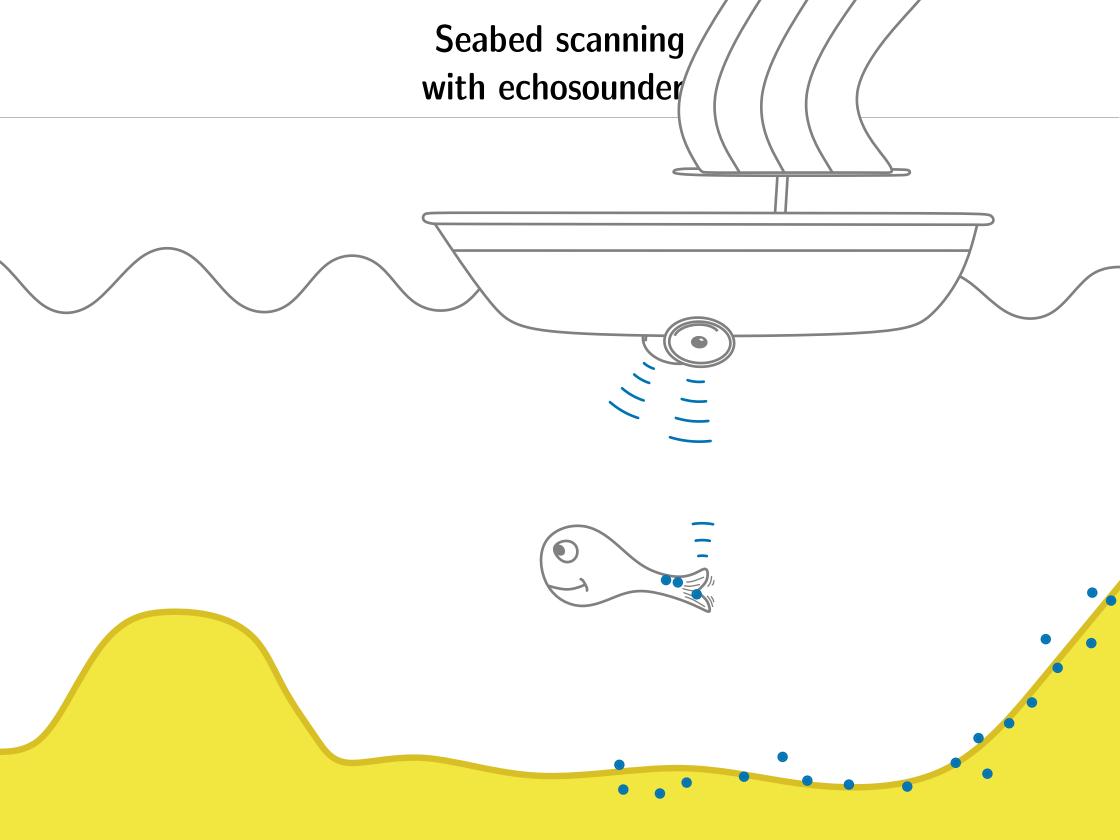


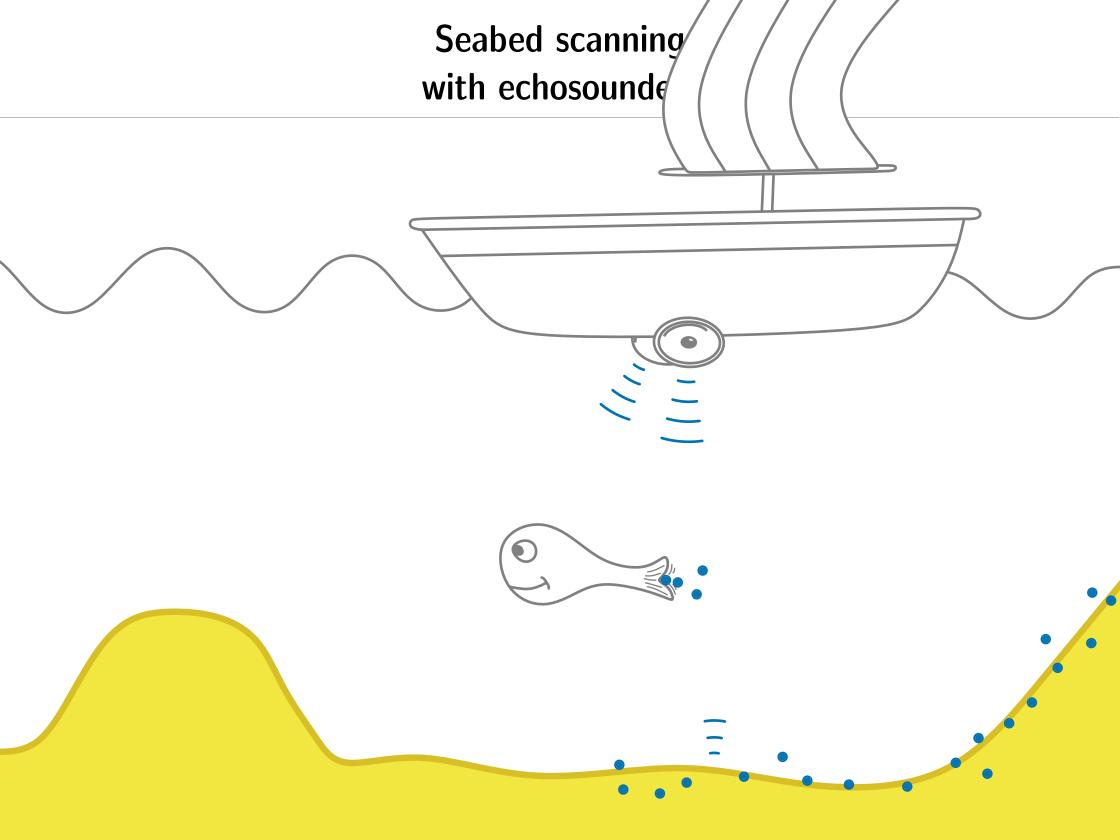


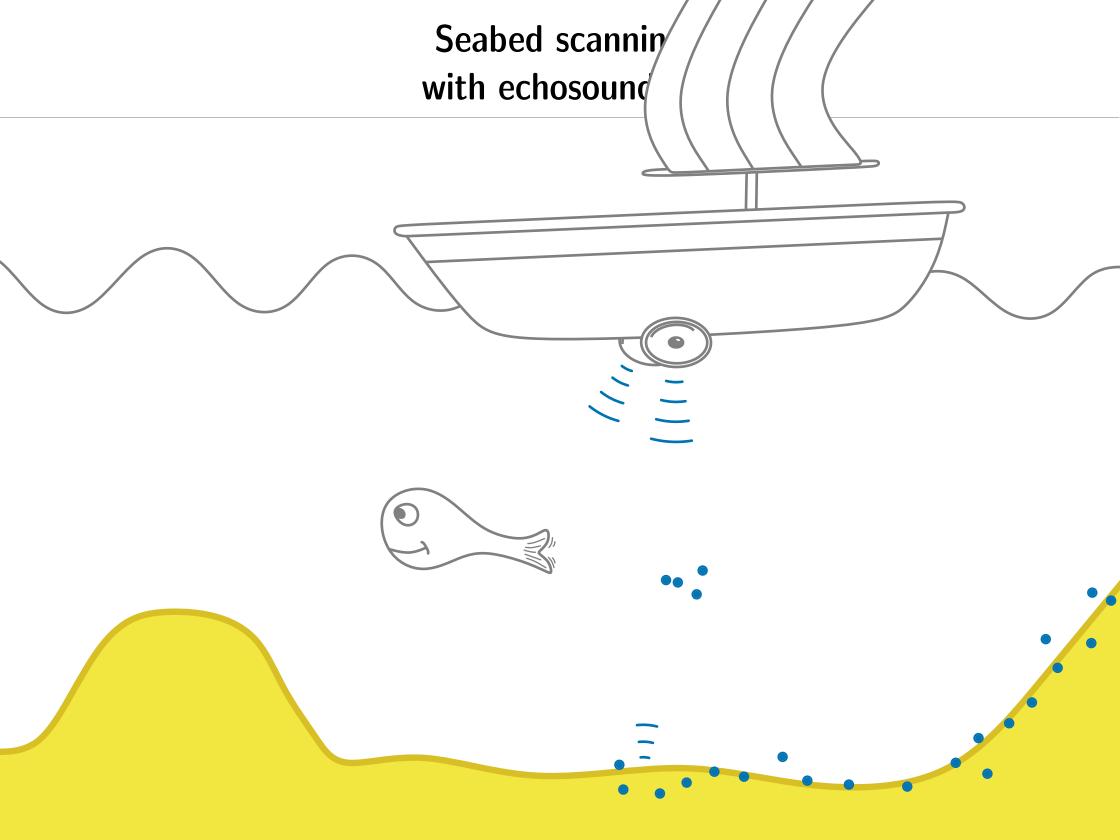


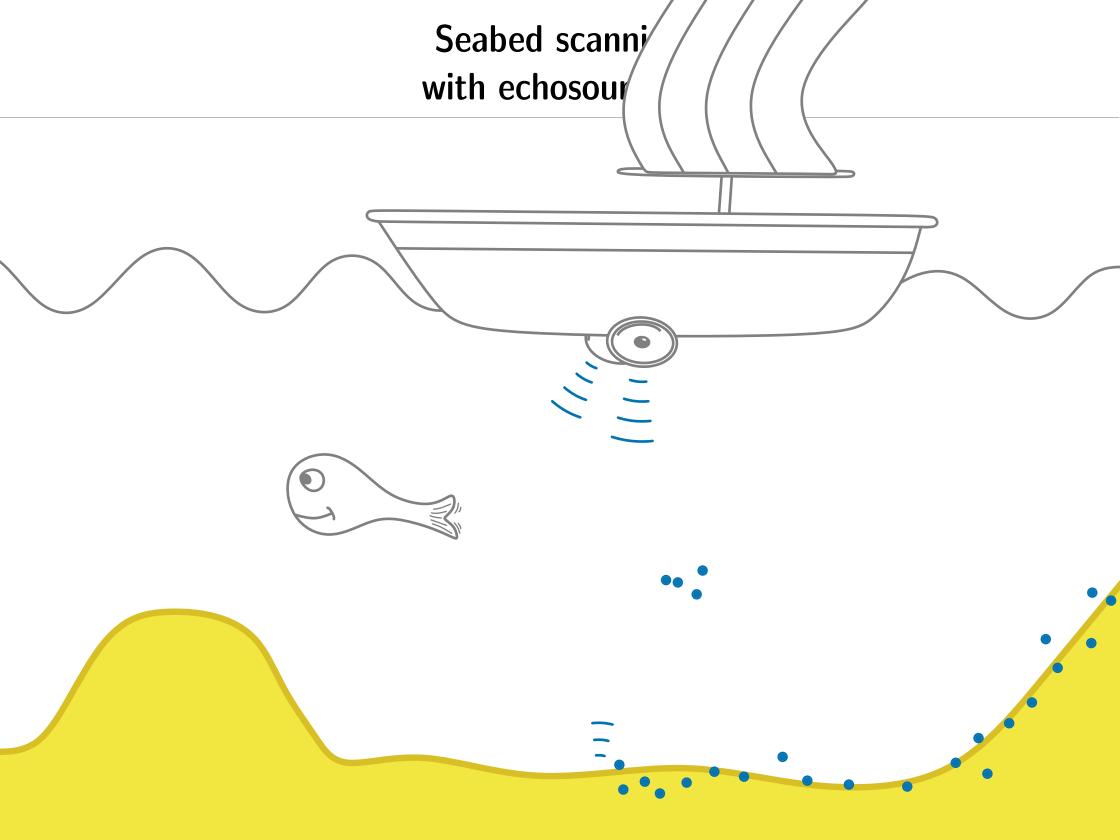


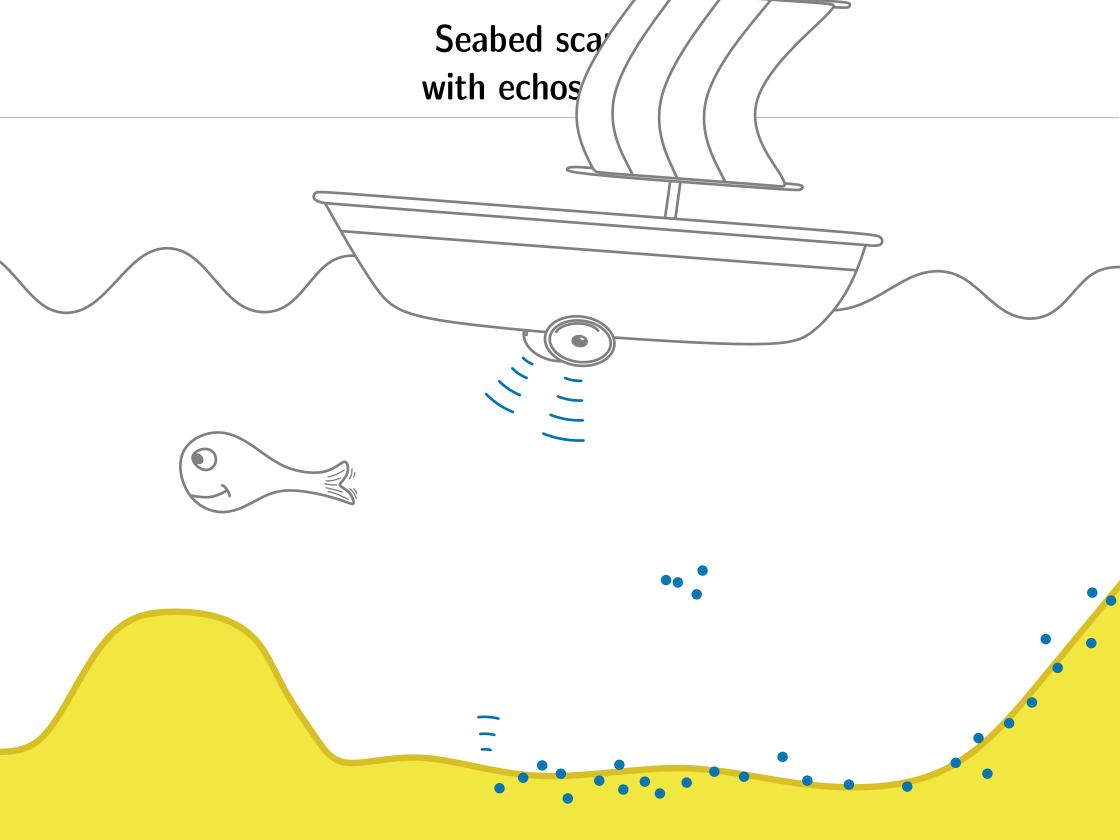


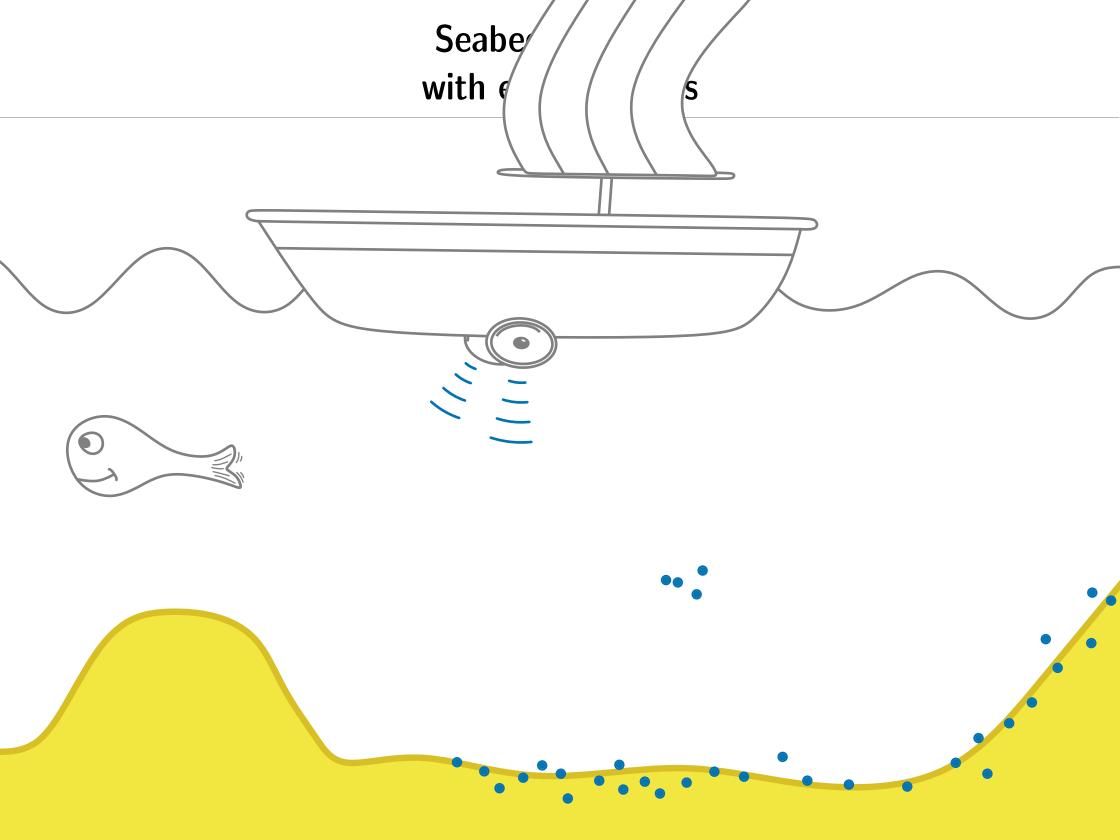


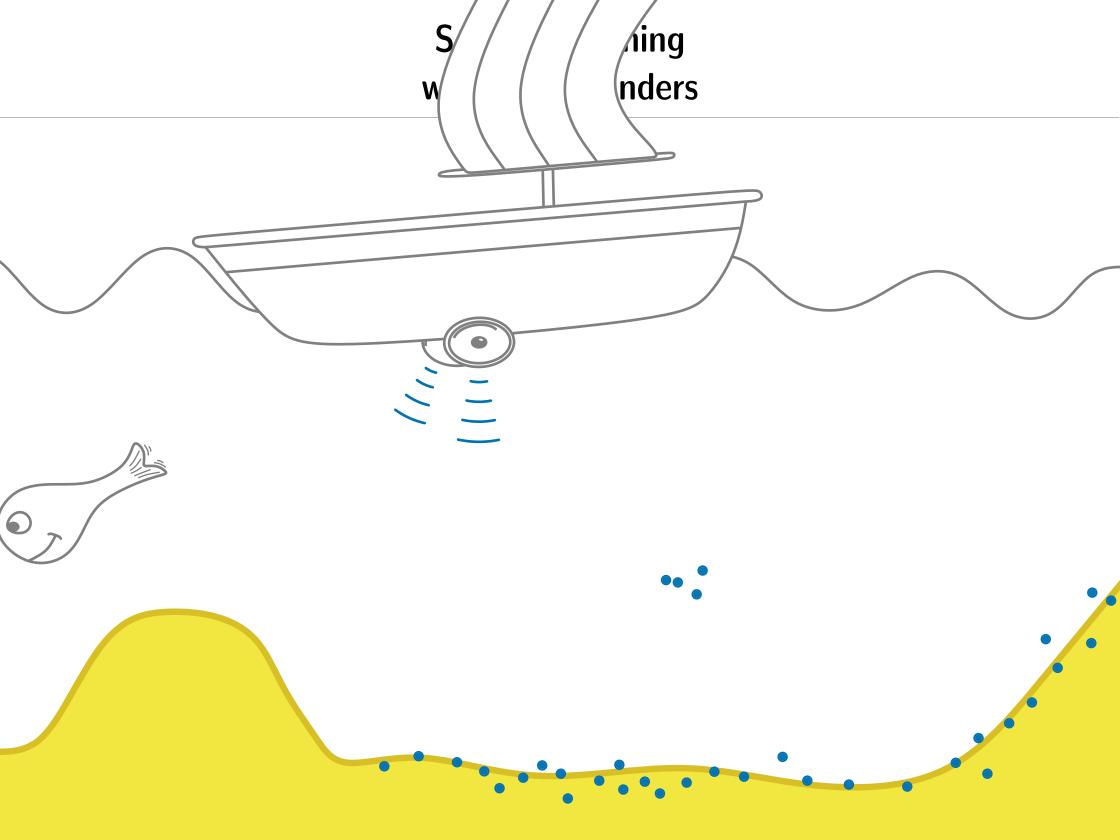


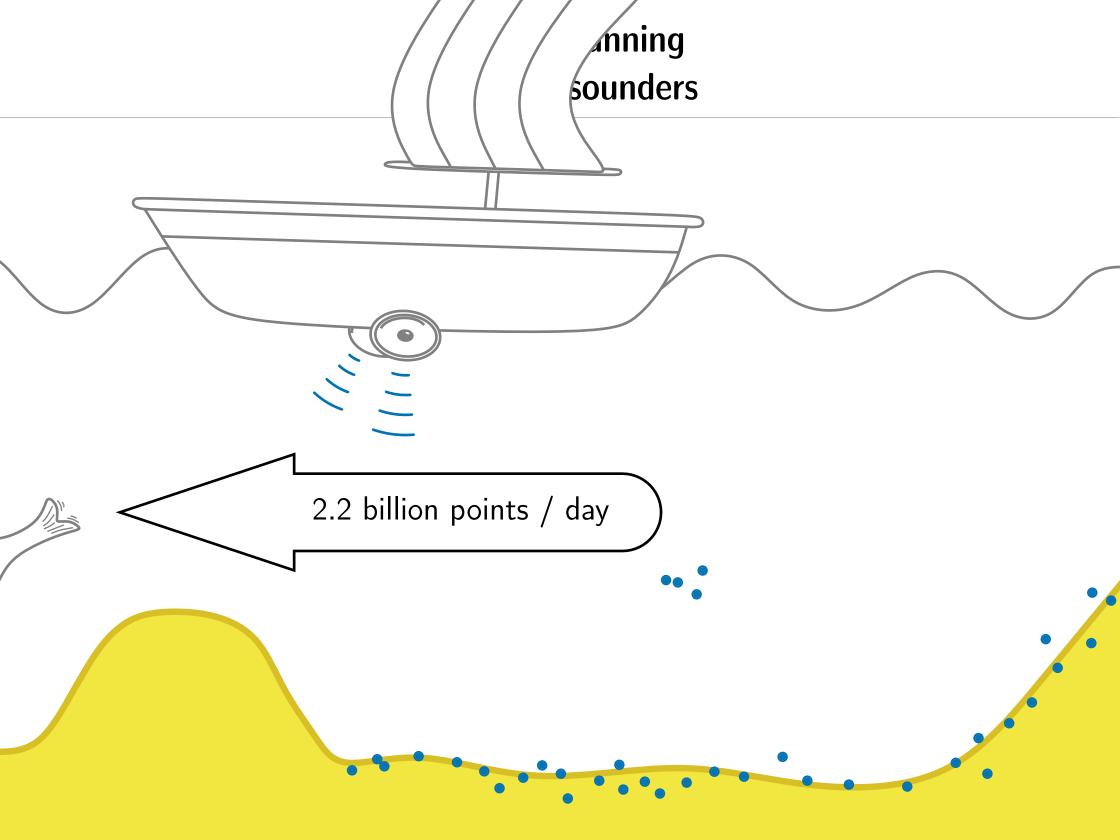


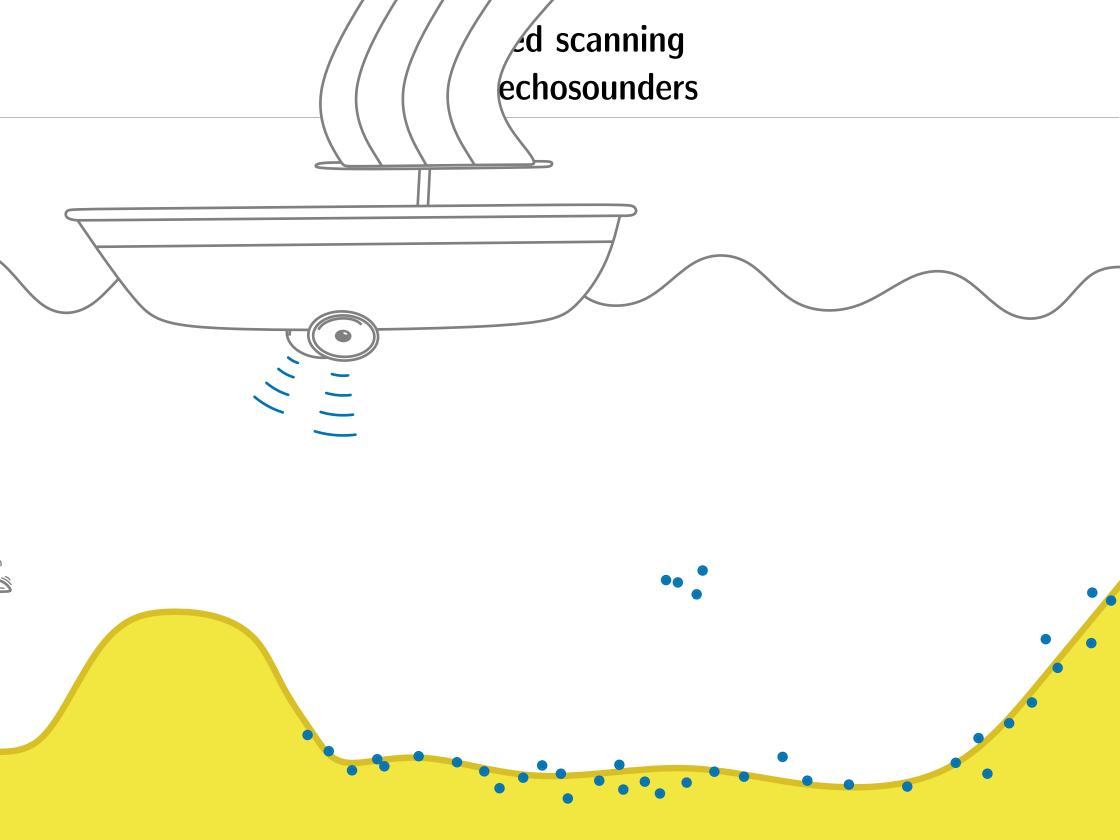


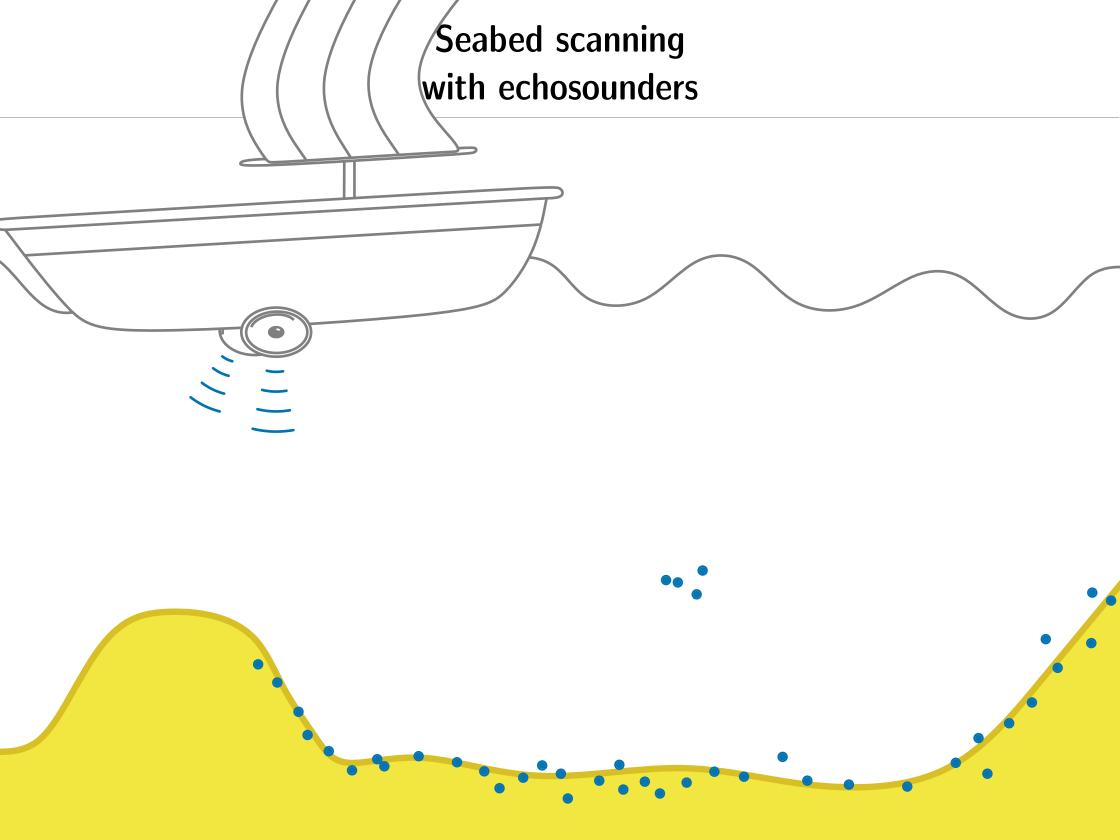


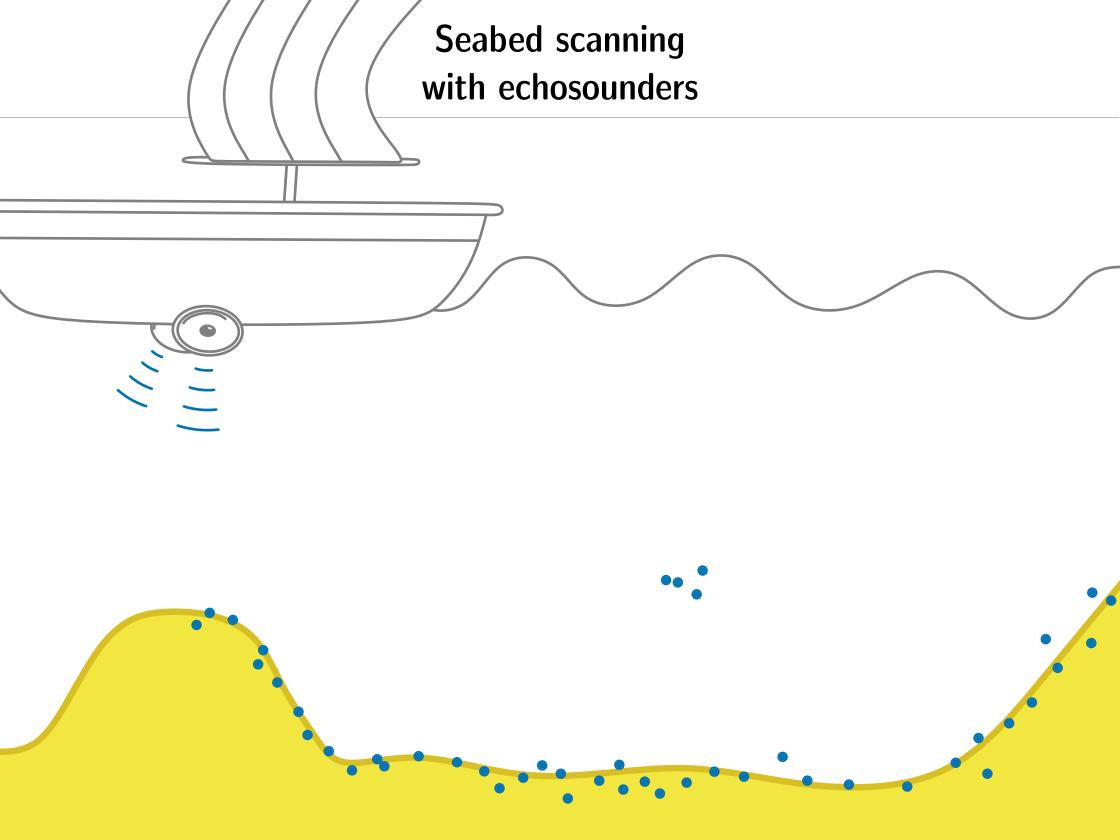


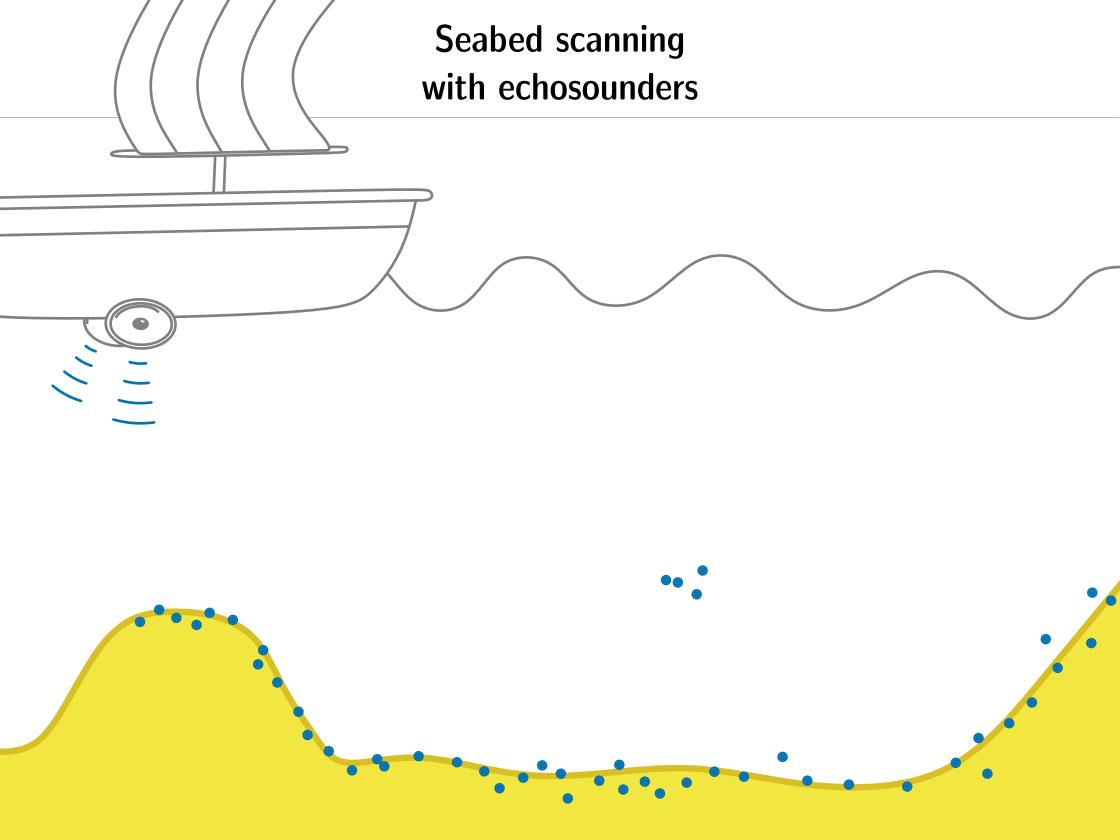


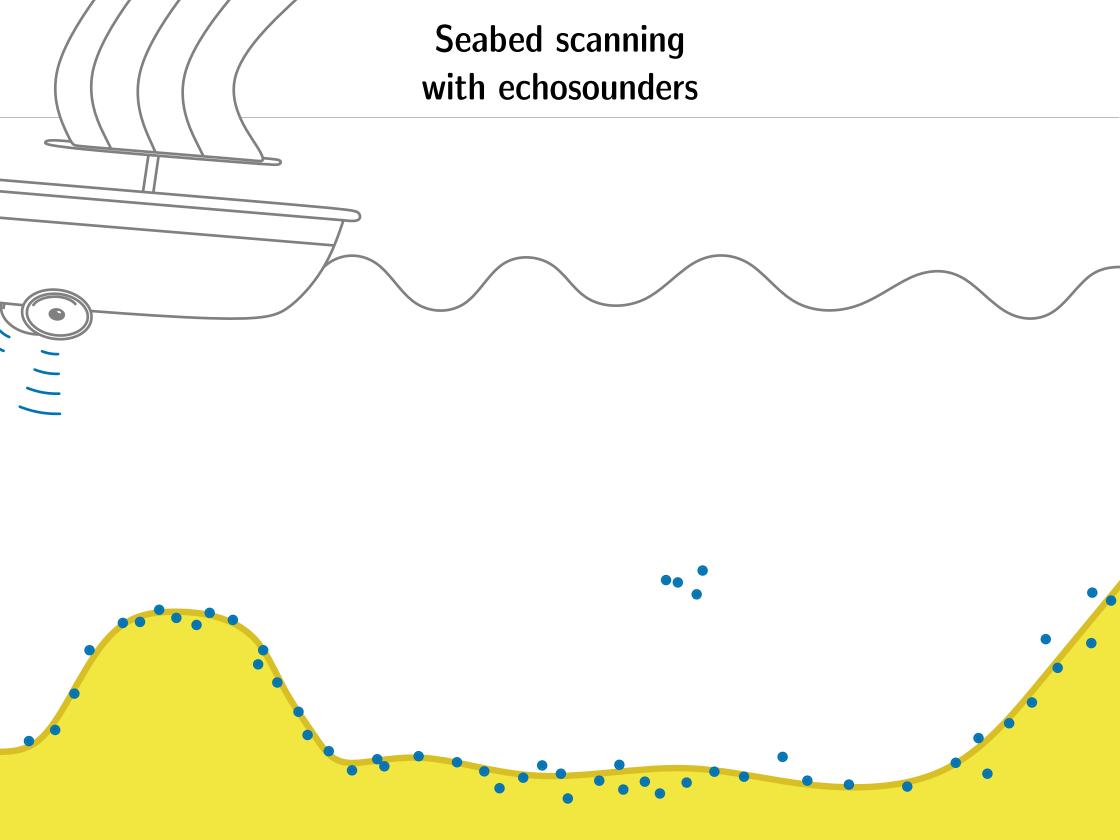


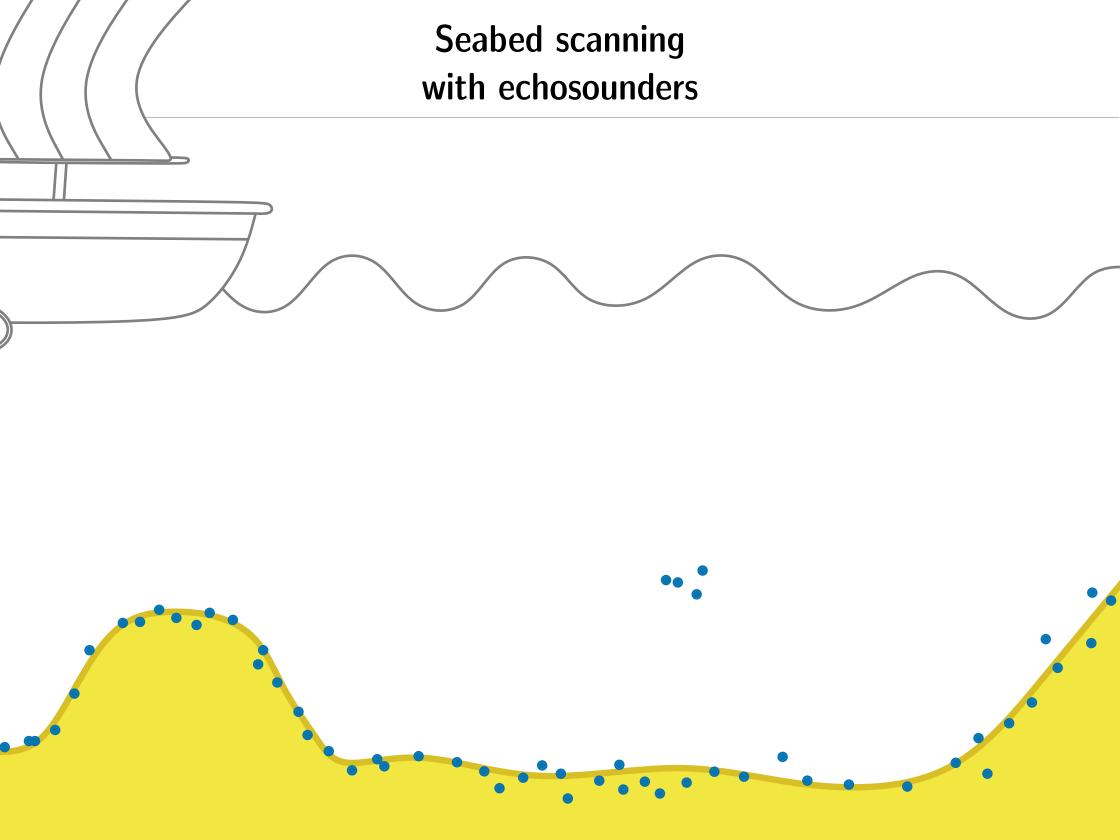


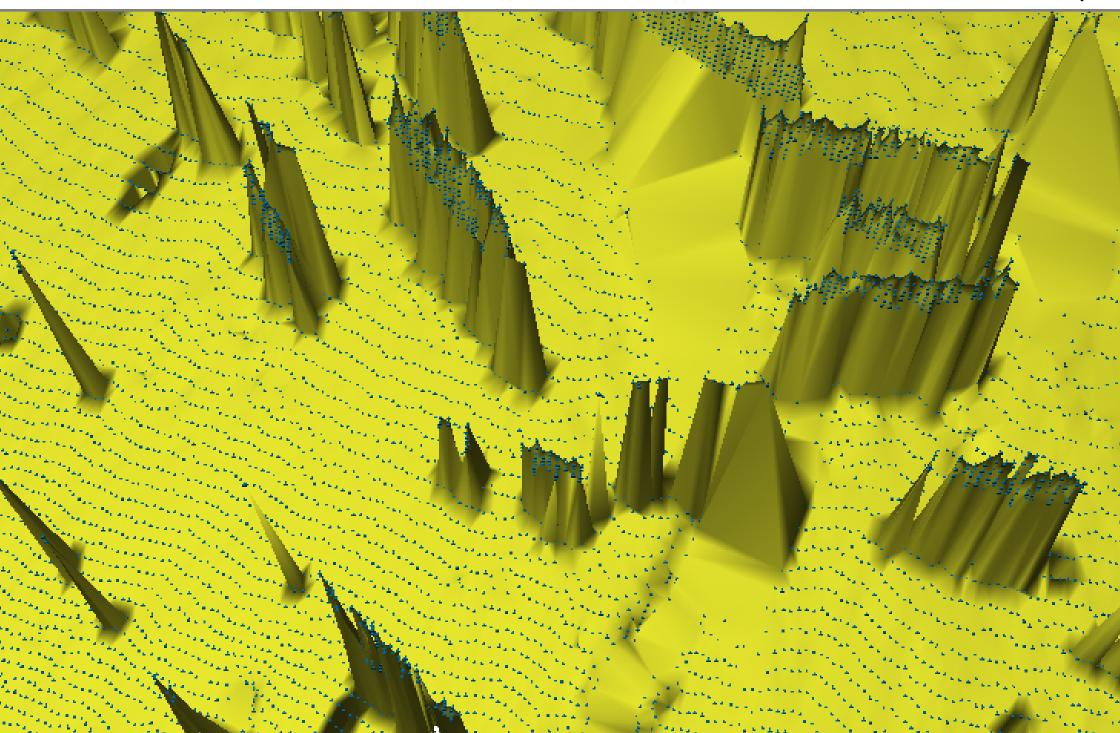


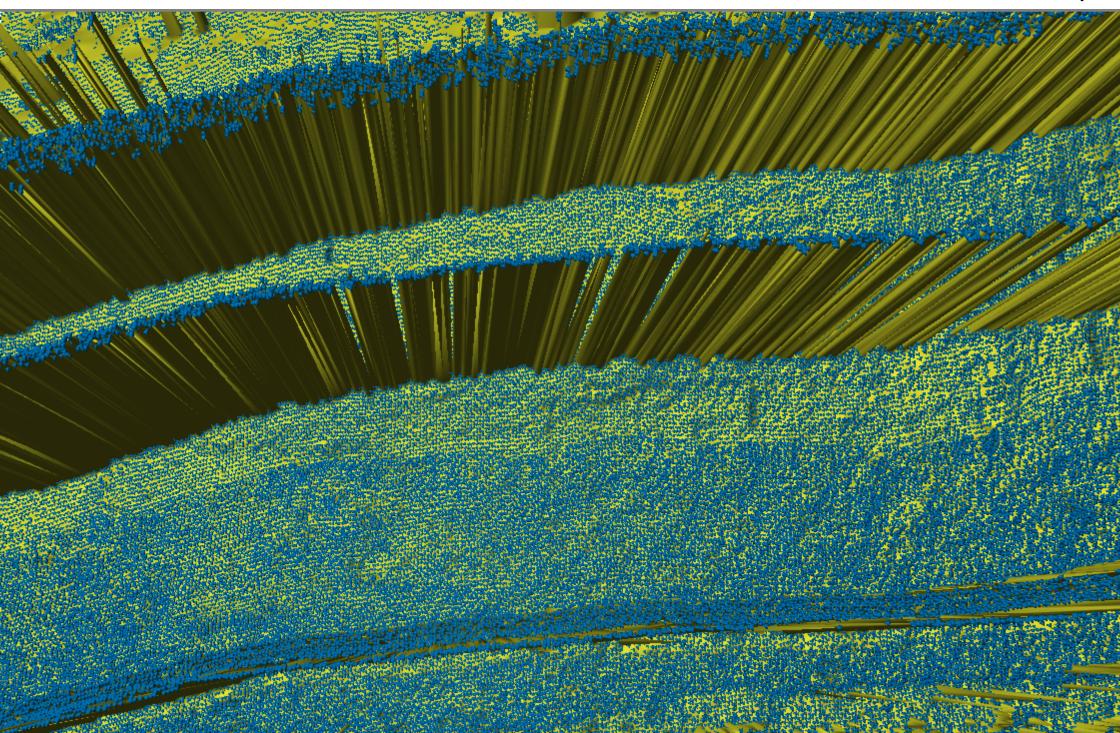


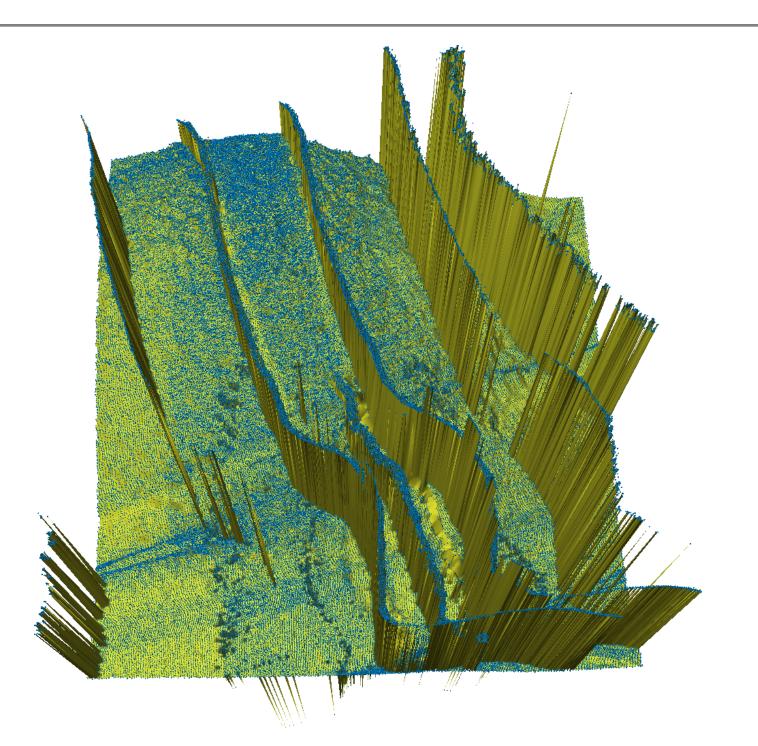


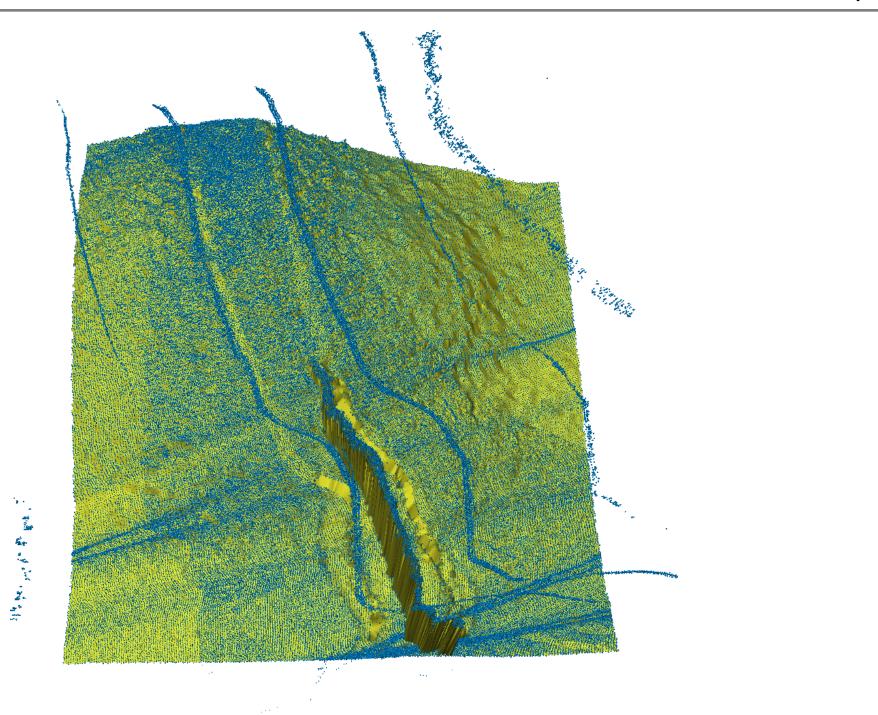






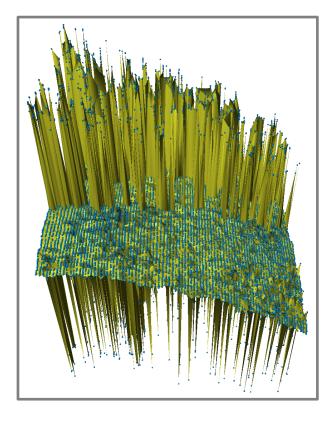






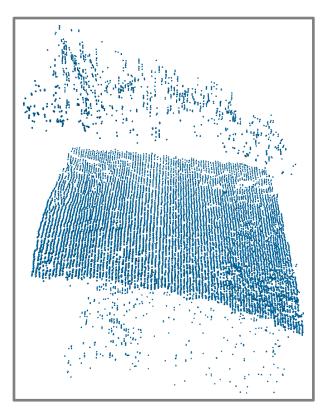
Data: StatoilHydro

1. Random spikes, possibly clustered



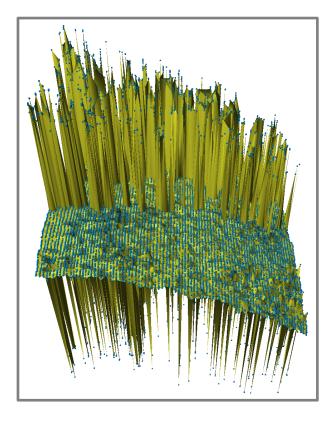
Data: StatoilHydro

1. Random spikes, possibly clustered

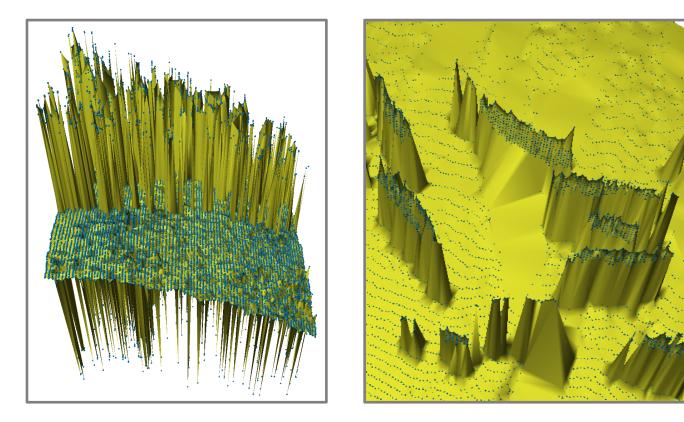


Data: StatoilHydro

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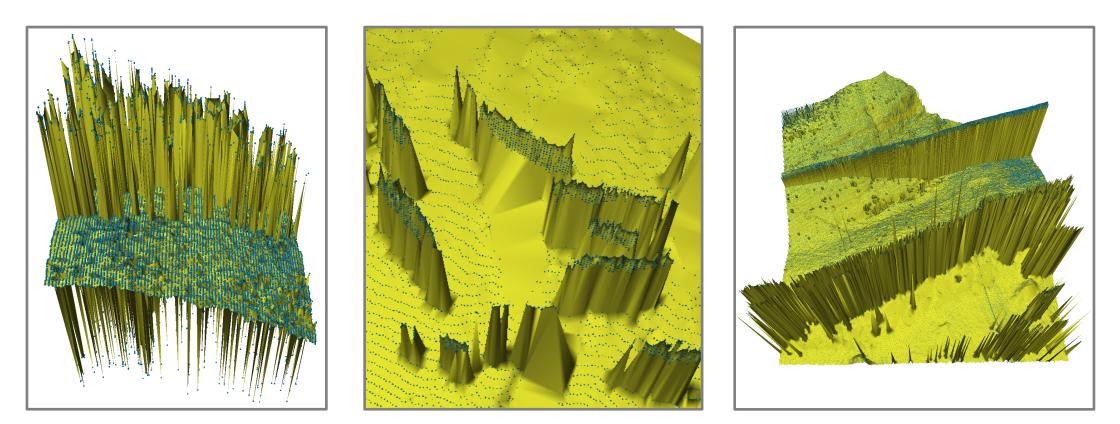


- 1. Random spikes, possibly clustered
- 2. Non-permanent physical objects (e.g. fish)



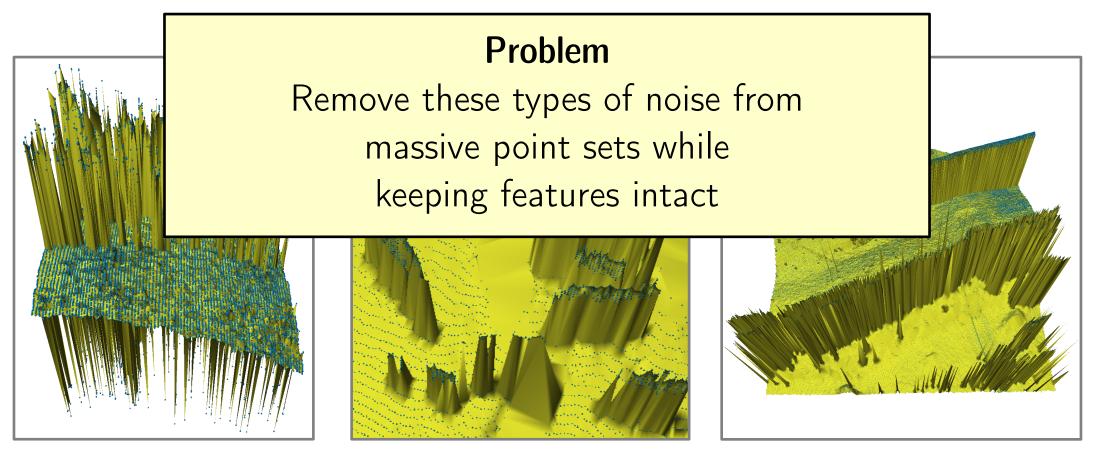
Noise types

- 1. Random spikes, possibly clustered
- 2. Non-permanent physical objects (e.g. fish)
- 3. Structural noise

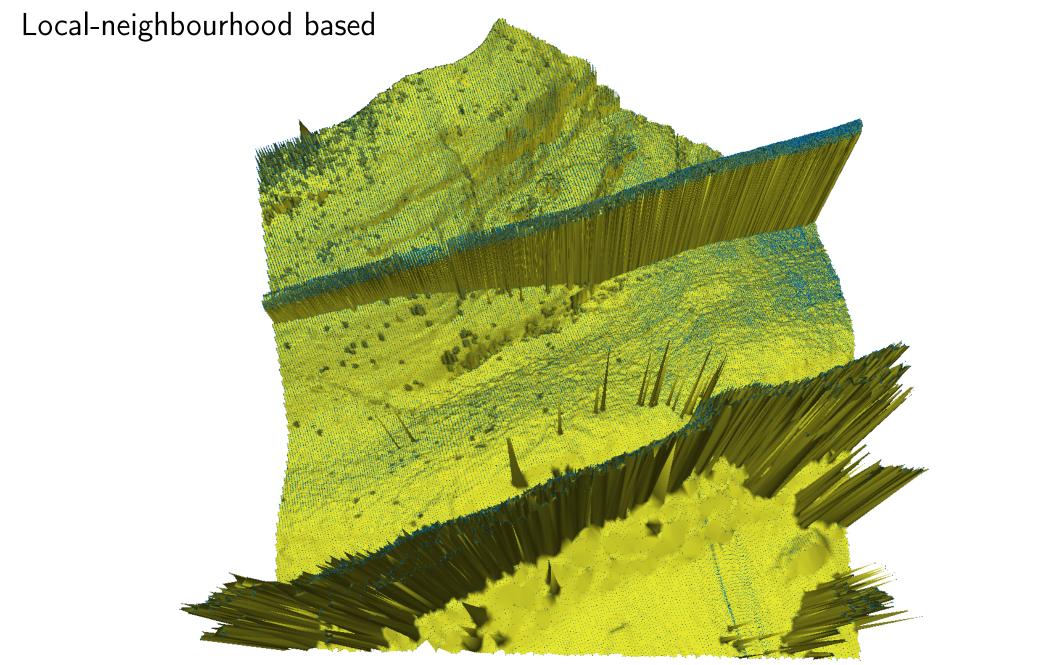


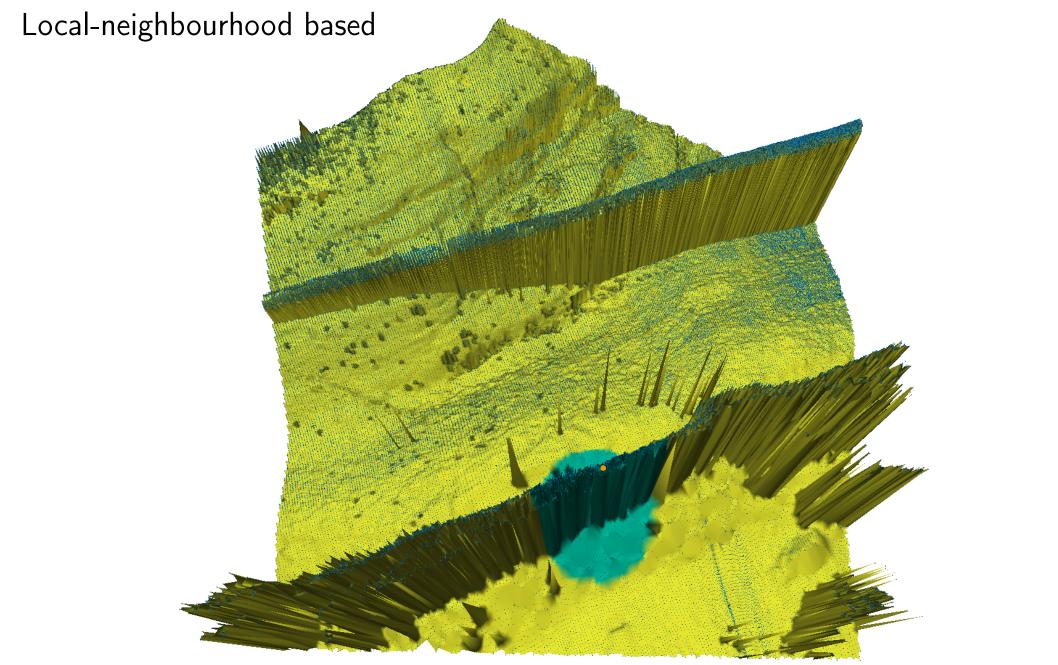
Noise types

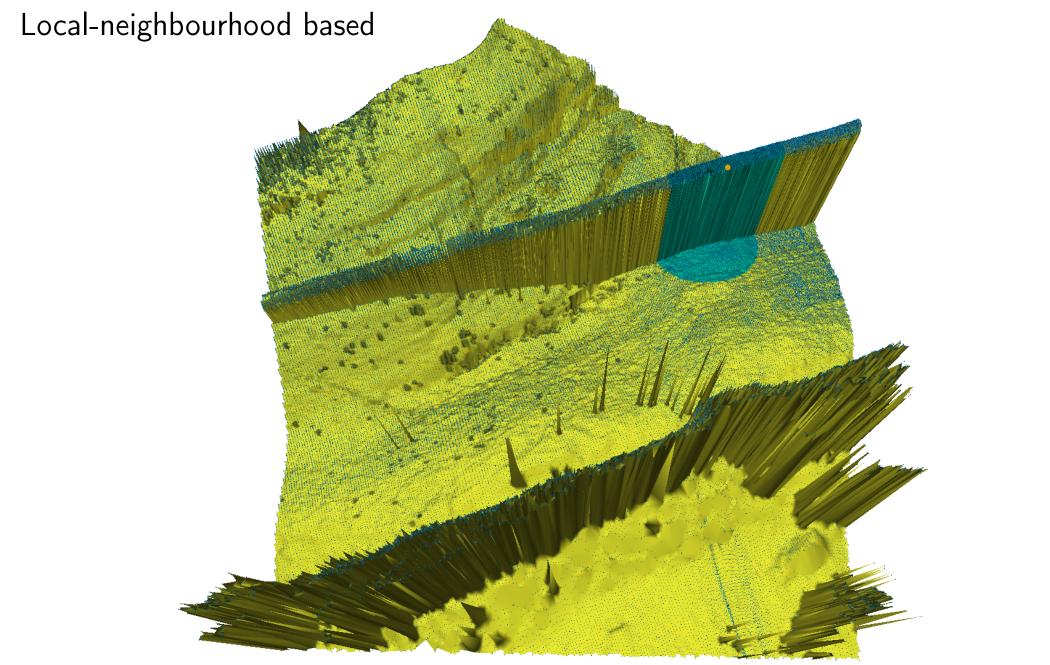
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Local-neighbourhood based

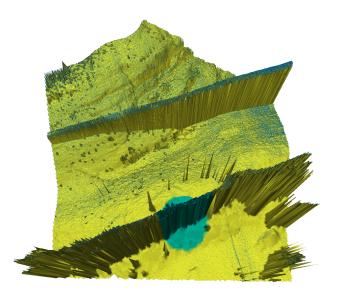






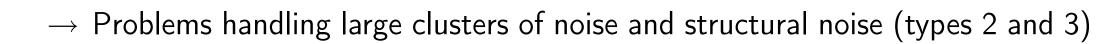
Local-neighbourhood based

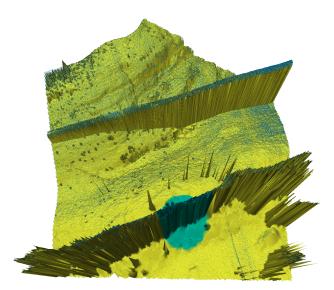
- E.g. CUBE [Calder & Mayer 2003], industry standard
 - Place grid over points
 - Estimate heights at grid nodes
 - Stastical analysis of points in neighbourhood
 - Remove points far away from estimated surface



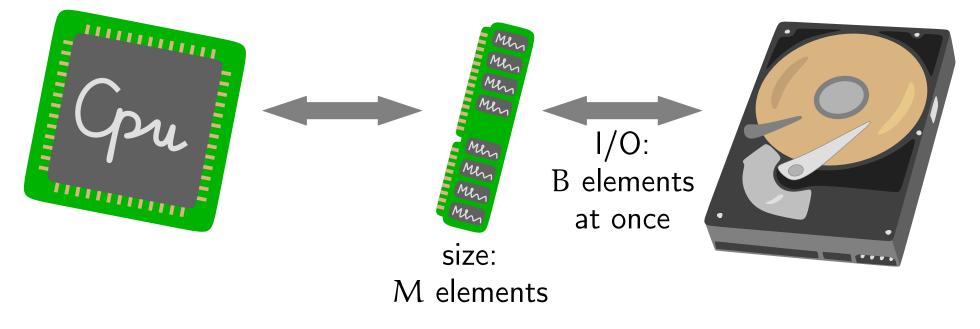
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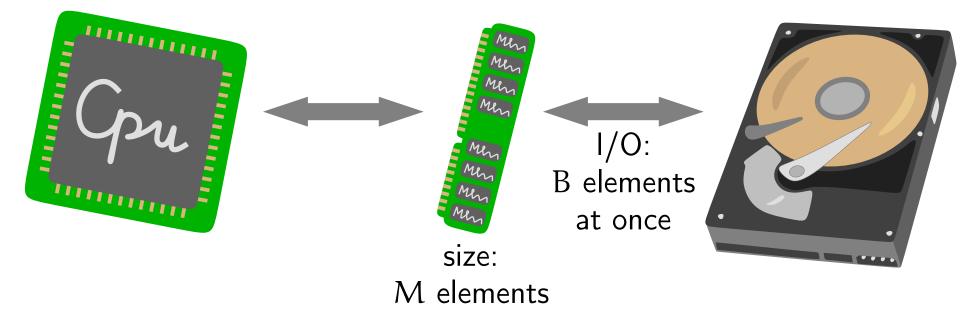




I/O model: analyze number of data transfers between internal and external memory

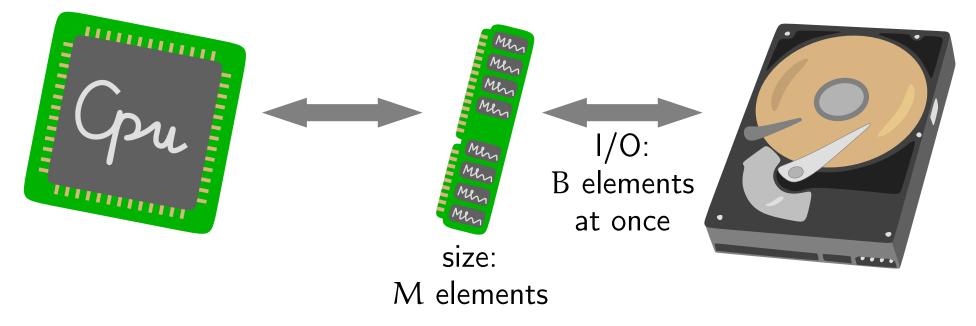


I/O model: analyze number of data transfers between internal and external memory



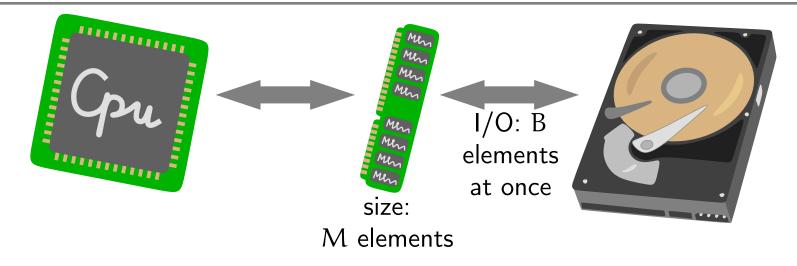
• Scanning N elements: $\Theta(\operatorname{scan}(N)) = \Theta(N/B)$ I/Os

I/O model: analyze number of data transfers between internal and external memory



- Scanning N elements: $\Theta(\mathtt{scan}(N)) = \Theta(N/B)$ I/Os
- Sorting N elements: $\Theta(\texttt{sort}(N)) = \Theta(N/B \log_{M/B} N/B)$ I/Os

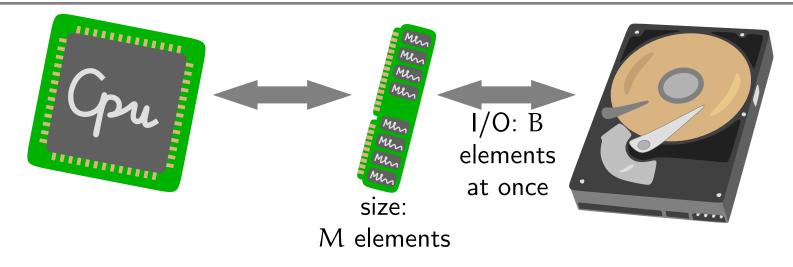
Previous work



Delaunay triangulation for computing a TIN DEM

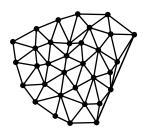


Previous work

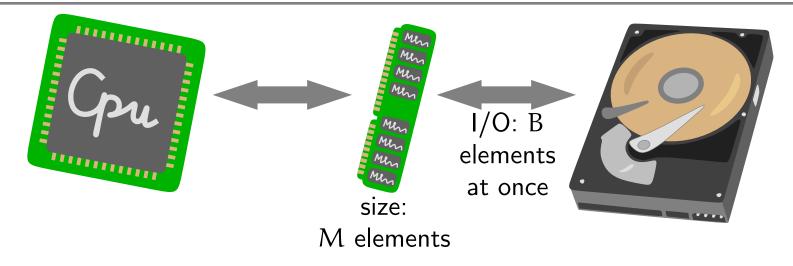


Delaunay triangulation for computing a TIN DEM

- O(sort(N)) [Goodrich et al. 1993, Kumar & Ramos 2002]
- Practical O(sort(N)) [Agarwal et al. 2005]



Previous work

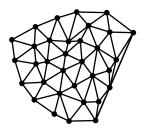


Delaunay triangulation for computing a TIN DEM

- O(sort(N)) [Goodrich et al. 1993, Kumar & Ramos 2002]
- Practical O(sort(N)) [Agarwal et al. 2005]

Connected components

- $O(\operatorname{sort}(|E|) \log_2 \log_2(B\frac{|V|}{|E|}))$ [Munagala and Ranade 1999]
- $\mathsf{Practical} \ O(\mathtt{sort}(N) \log_2(N/M)) \quad \texttt{[Agarwal et al. 2006] (batched union-find)}$





Our results

- Cleaning algorithm for MBES data
 - Identifies both random, local and structural noise
 - Theoretically I/O-efficient
 - Practically efficient and implementable



Our results

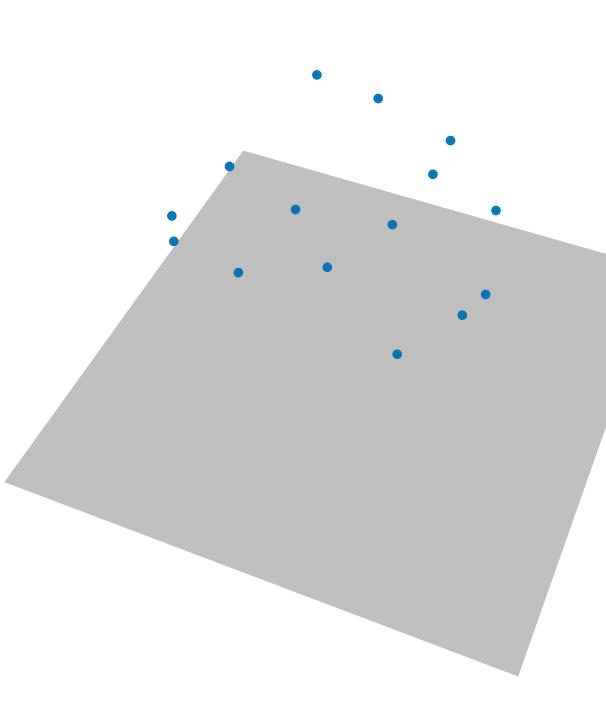
- Cleaning algorithm for MBES data
 - Identifies both random, local and structural noise
 - Theoretically I/O-efficient
 - Practically efficient and implementable

- Connected component algorithm
 - O(sort(N)) I/Os under a natural assumption
 - Practically efficient and implementable





1. Perturb "xy-duplicate" points

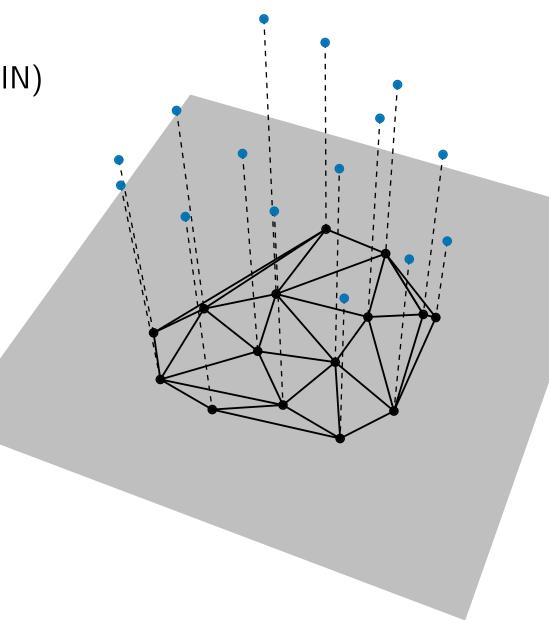


1. Perturb "xy-duplicate" points ۱۱ 11 11 11 11 11

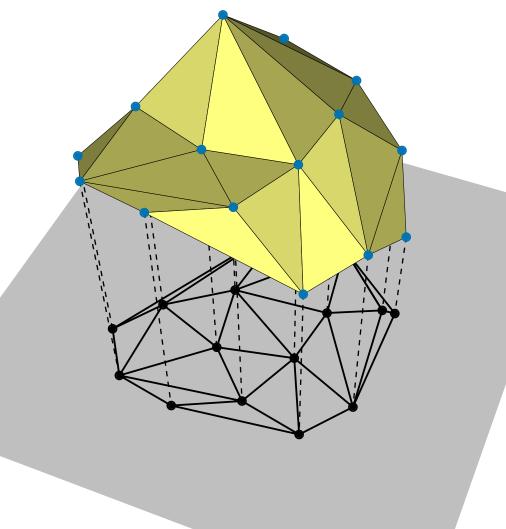
1. Perturb "xy-duplicate" points 11 11 11 11 11 11

1. Perturb "xy-duplicate" points 2. Create Delaunay triangulation (\rightarrow TIN) 11 11 11

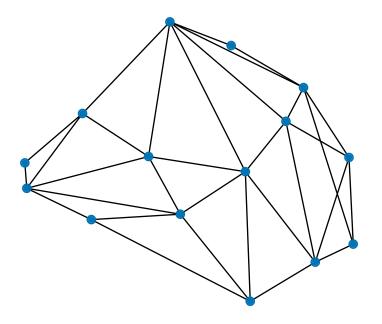
- 1. Perturb "xy-duplicate" points
- 2. Create Delaunay triangulation (\rightarrow TIN)



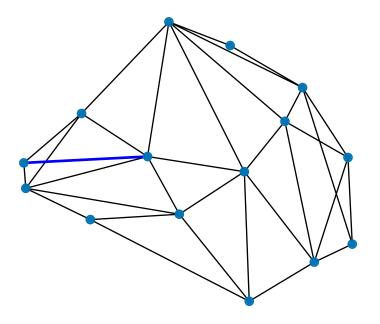
- 1. Perturb "xy-duplicate" points
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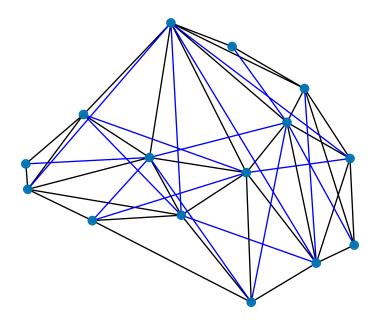
- 1. Perturb "xy-duplicate" points
- 2. Create Delaunay triangulation (\rightarrow TIN)
- 3. Add *diagonals*



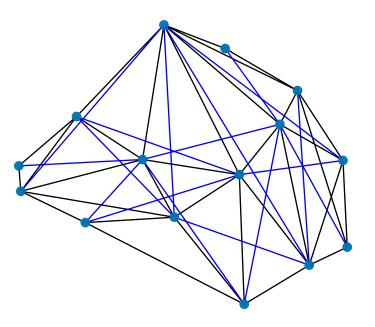
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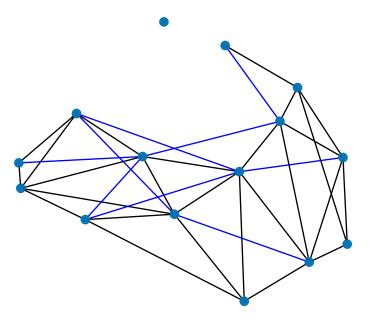
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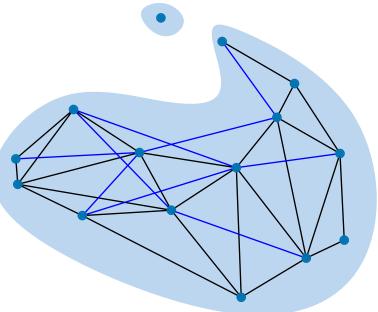
- 1. Perturb "xy-duplicate" points
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- 4. Remove edges with *z*-difference > threshold



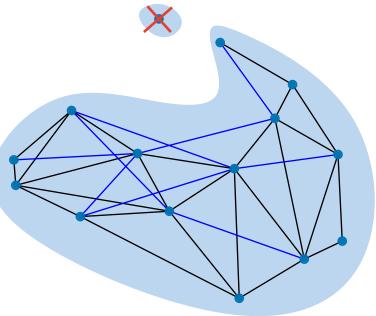
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- 1. Perturb "xy-duplicate" points
- 2. Create Delaunay triangulation (\rightarrow TIN)
- 3. Add diagonals
- 4. Remove edges with z-difference > threshold
- 5. Find largest connected component

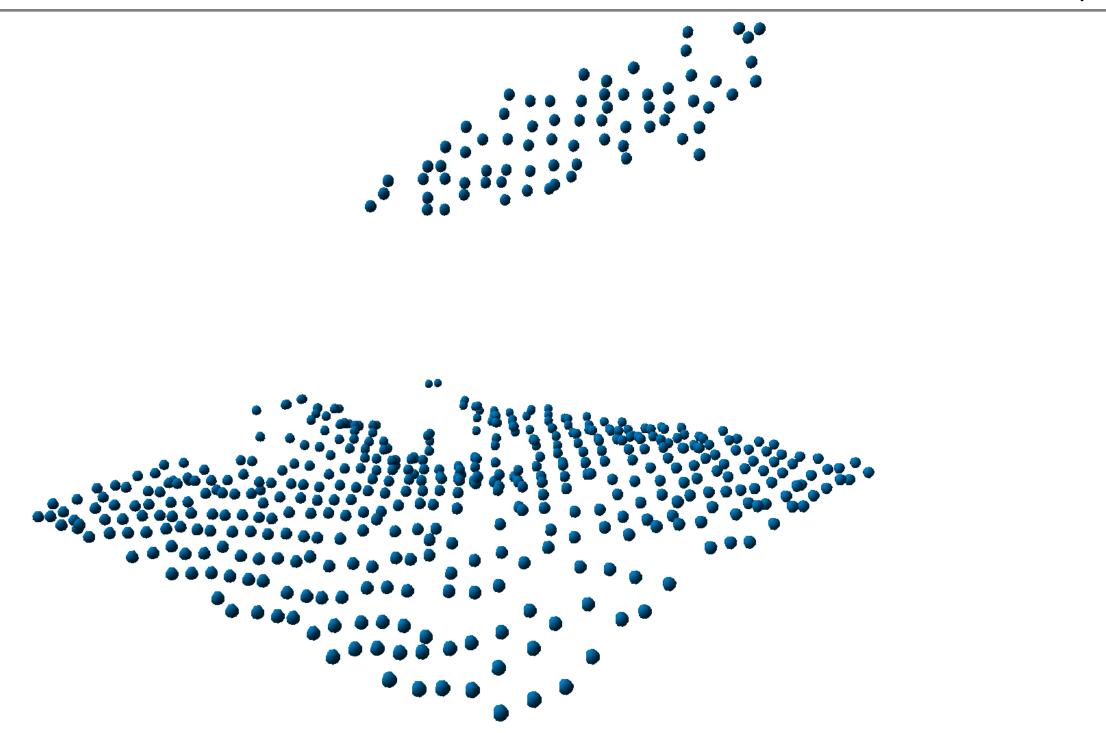


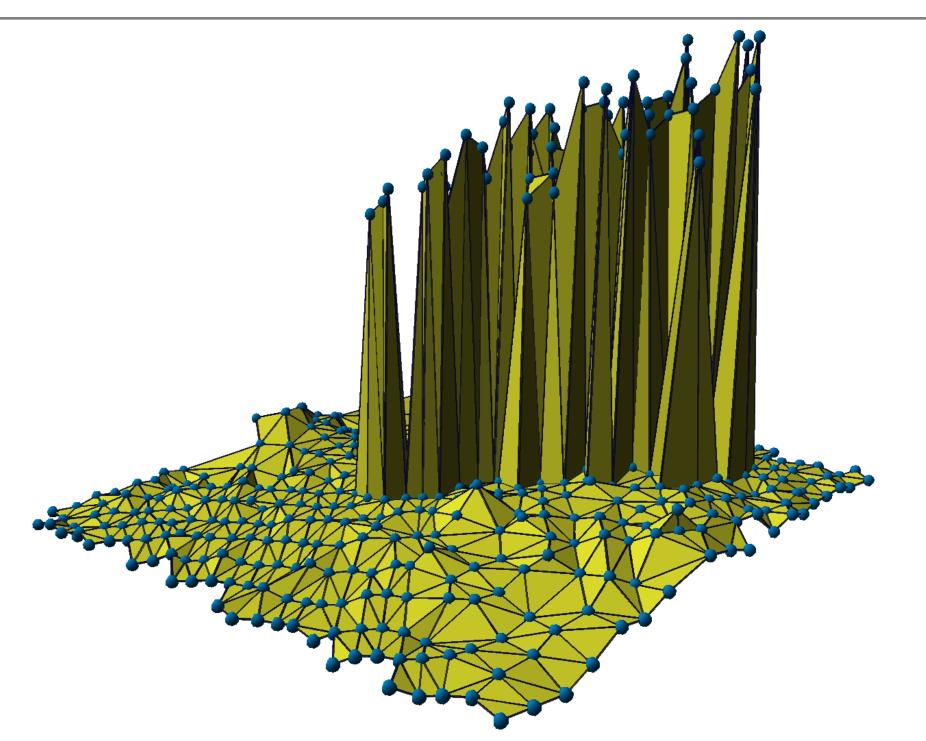
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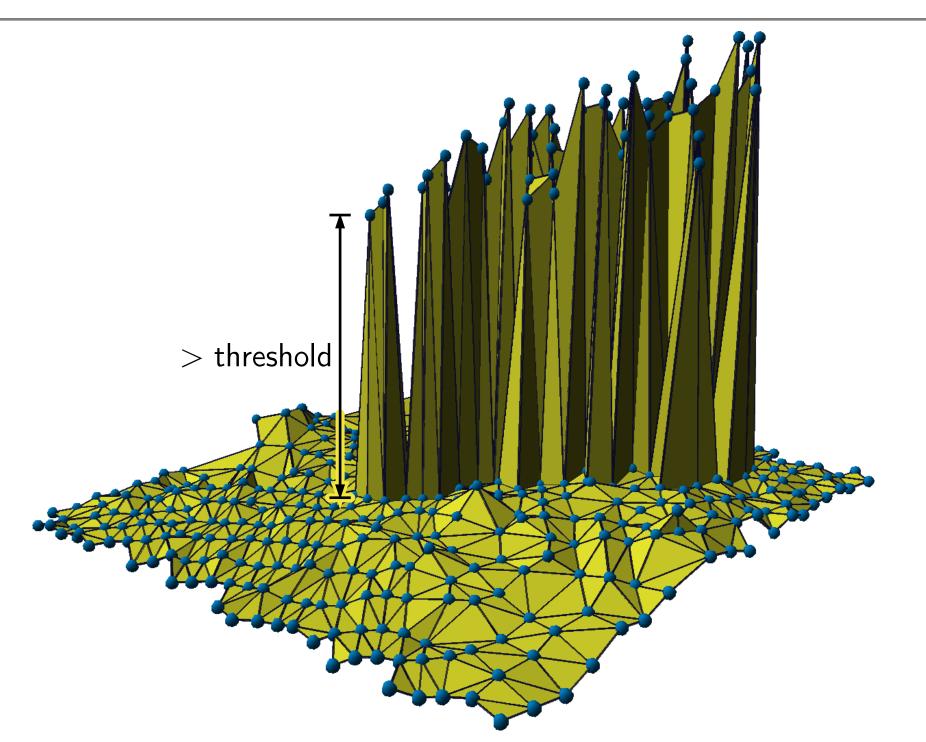


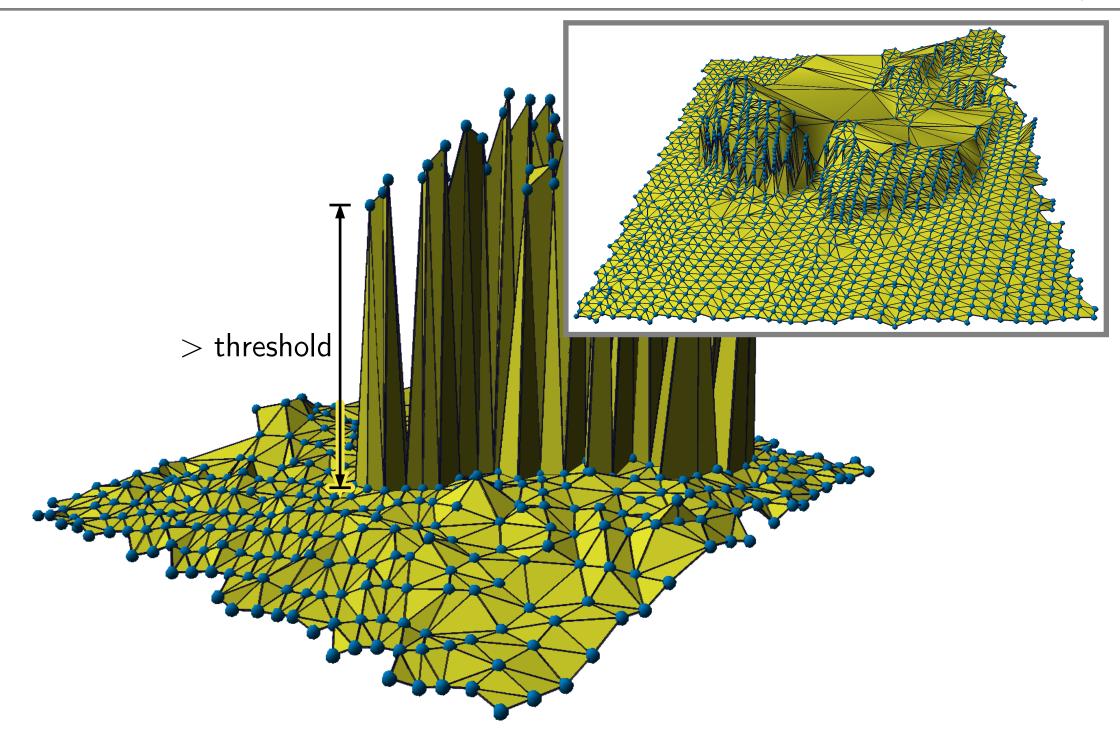
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 $O(\texttt{sort}(N)) \; \mathsf{I/Os} + \texttt{connected components} = O(\texttt{sort}(N) \log \log B) \; \mathsf{I/Os}, \, \texttt{or} \; O(\texttt{sort}(N)) \; \mathsf{I/Os} \; \texttt{under a practical assumption}$

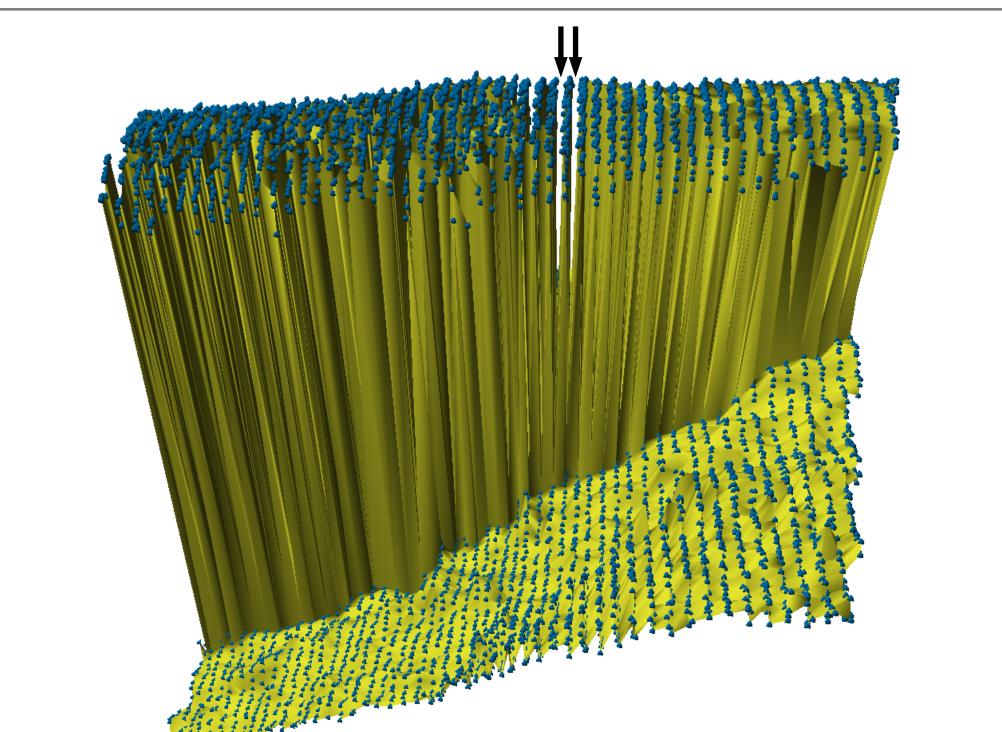




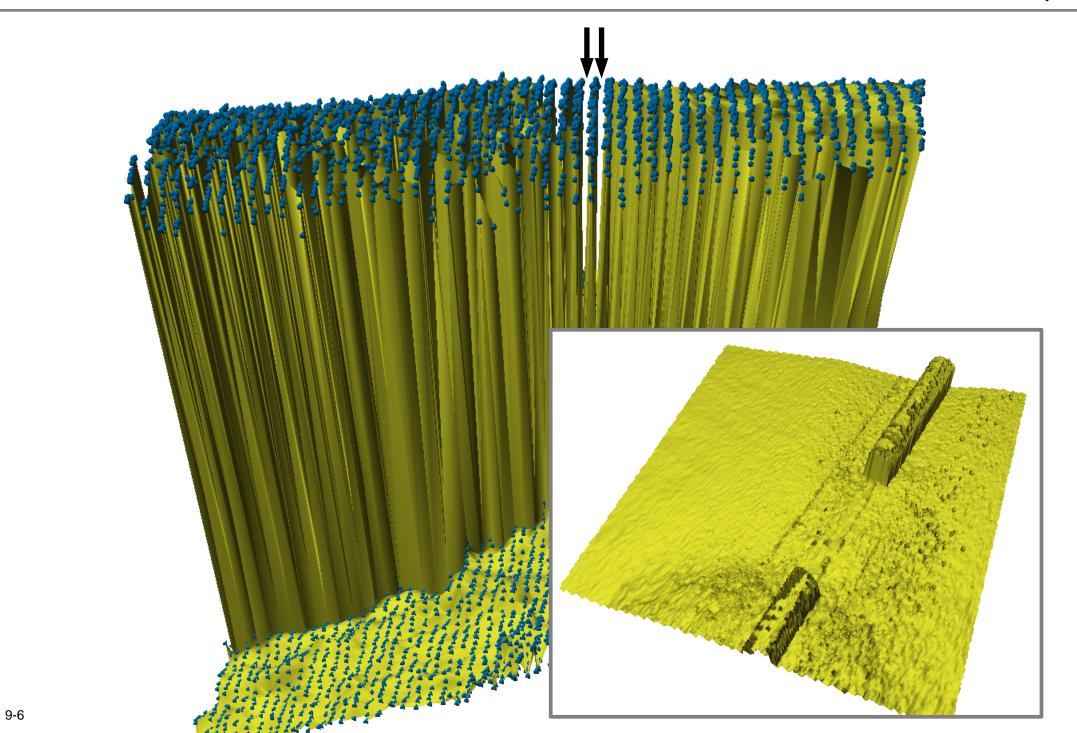




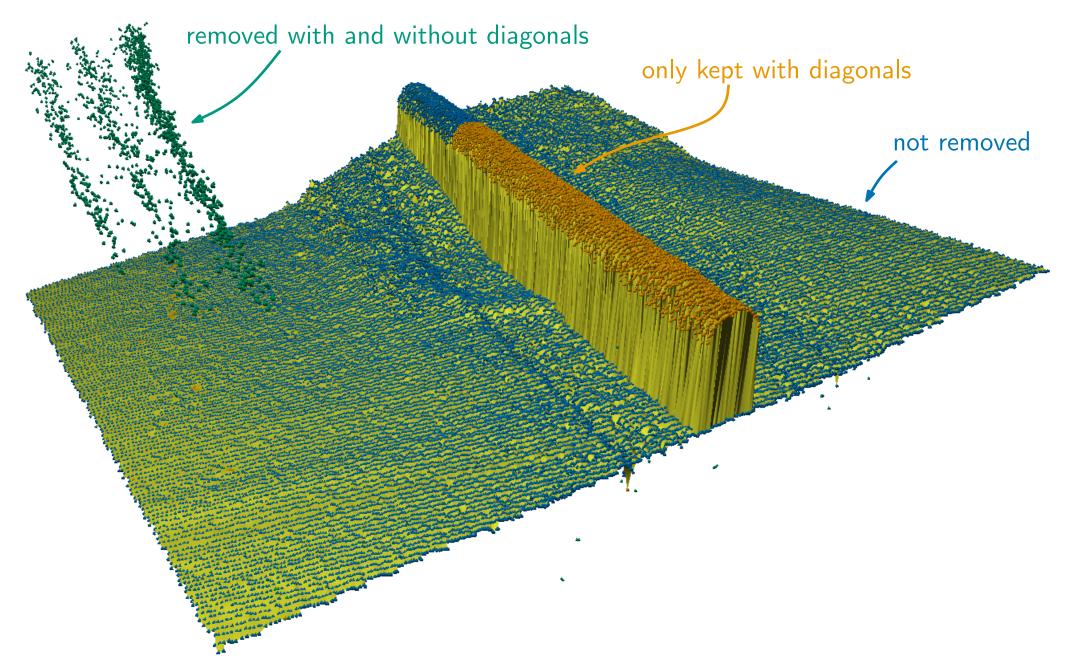
$\begin{array}{l} Why \ it \ works \\ \text{no diagonals} \Rightarrow \text{pipeline disconnected} \end{array}$



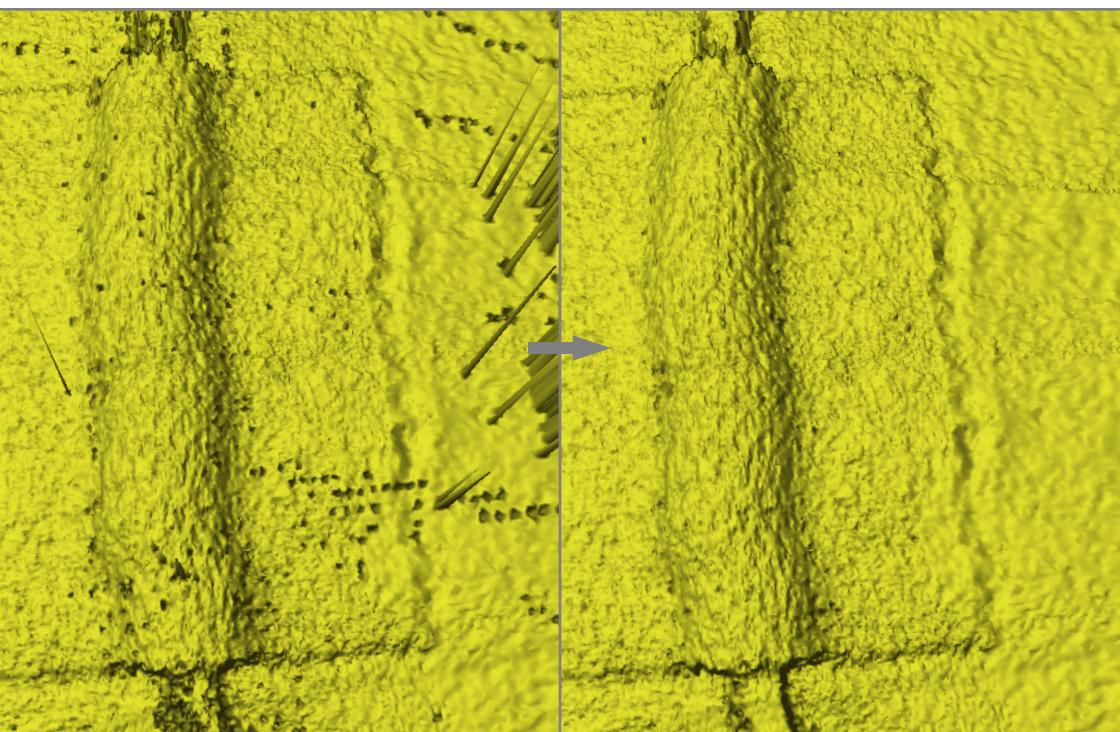
$\begin{array}{l} Why \ it \ works \\ \text{no diagonals} \Rightarrow \text{pipeline disconnected} \end{array}$



Why it works with vs. without diagonals

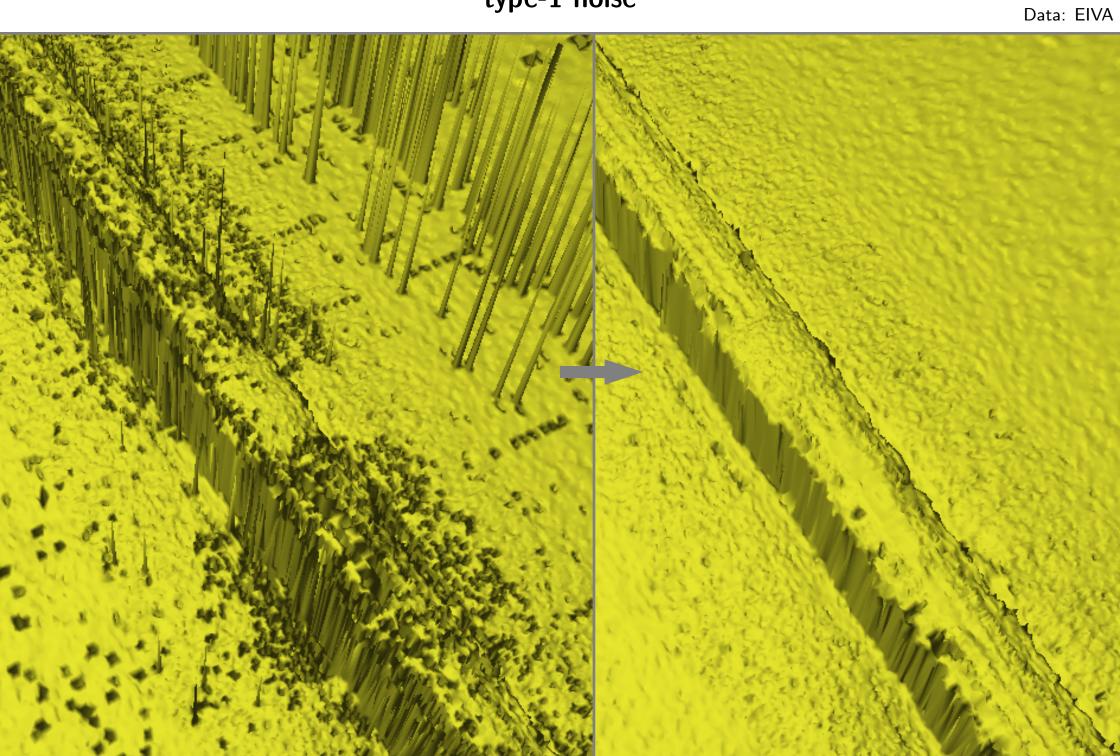


Results type-1 noise



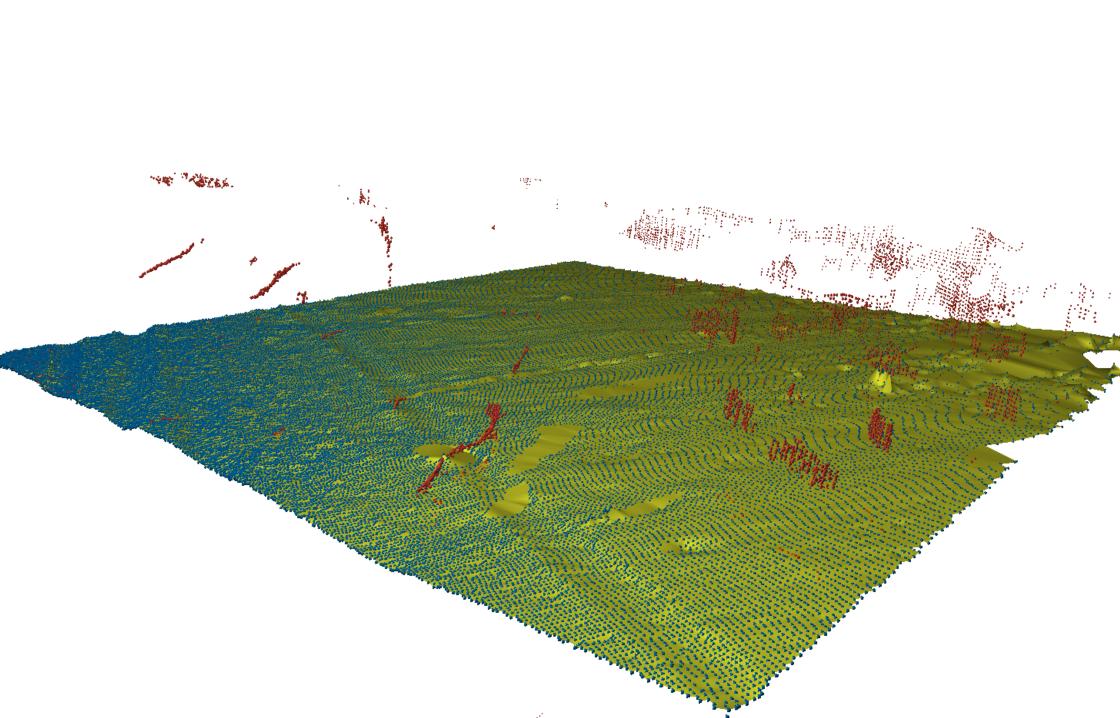
Data: EIVA

Results type-1 noise

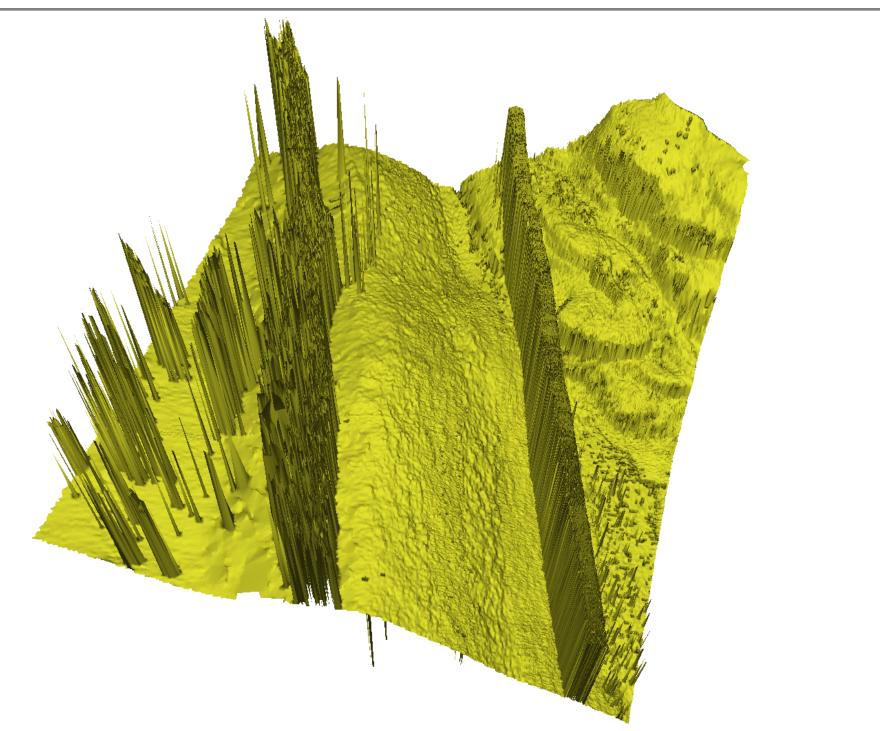


Results type-2 noise

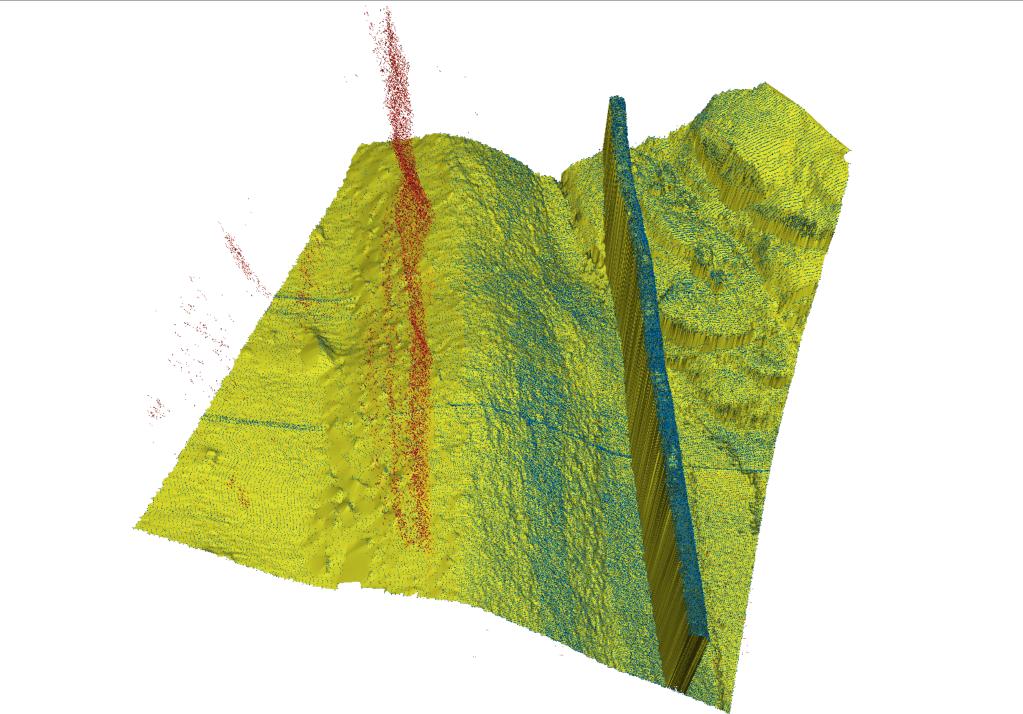
Results type-2 noise



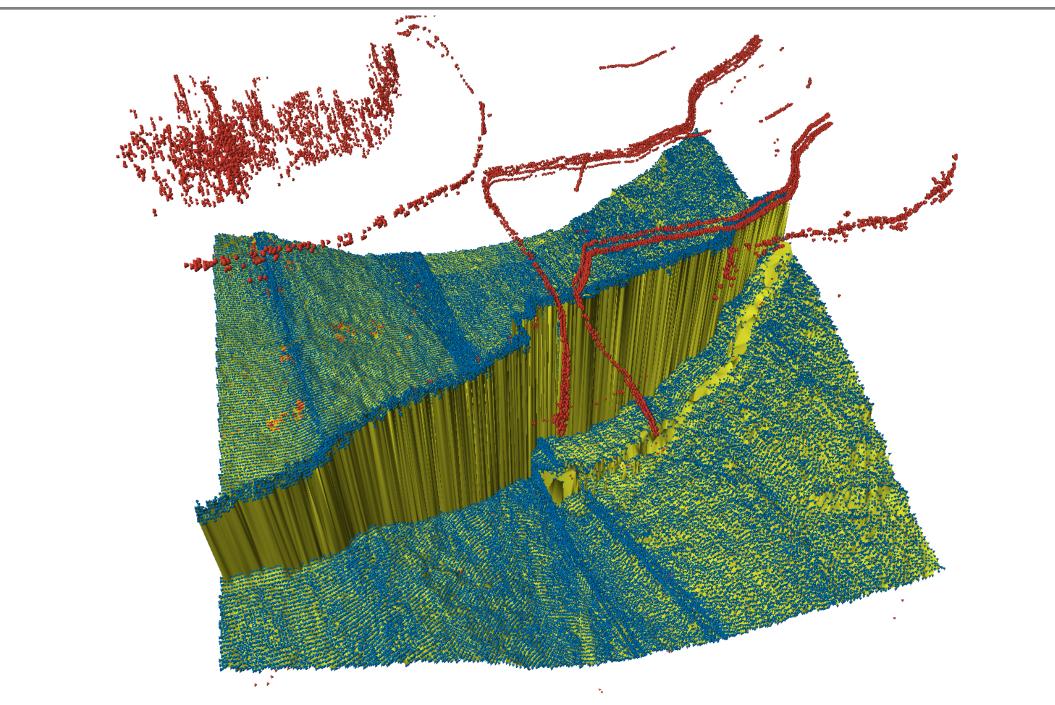
Results type-3 noise



Results type-3 noise



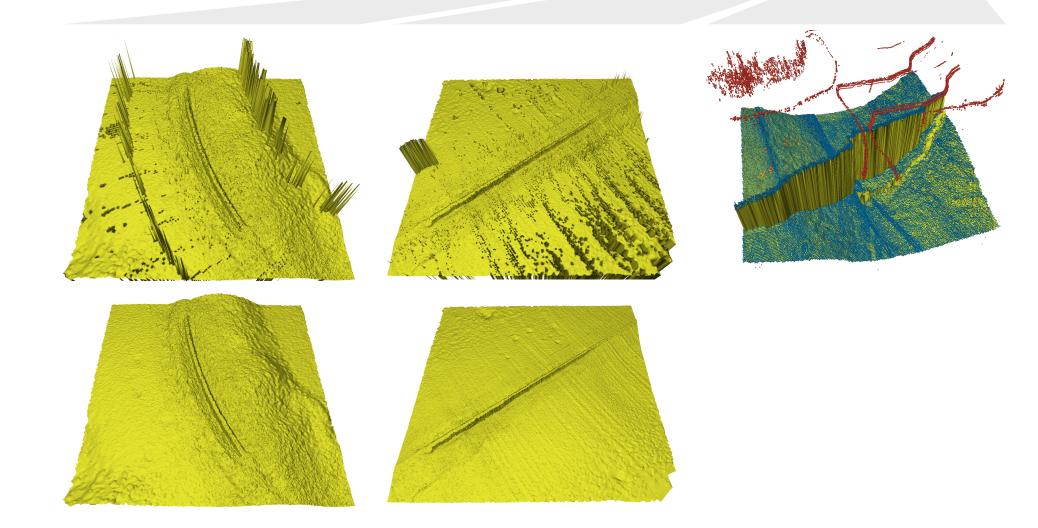
Results type-3 noise



Some numbers

Data: StatoilHydro, EIVA

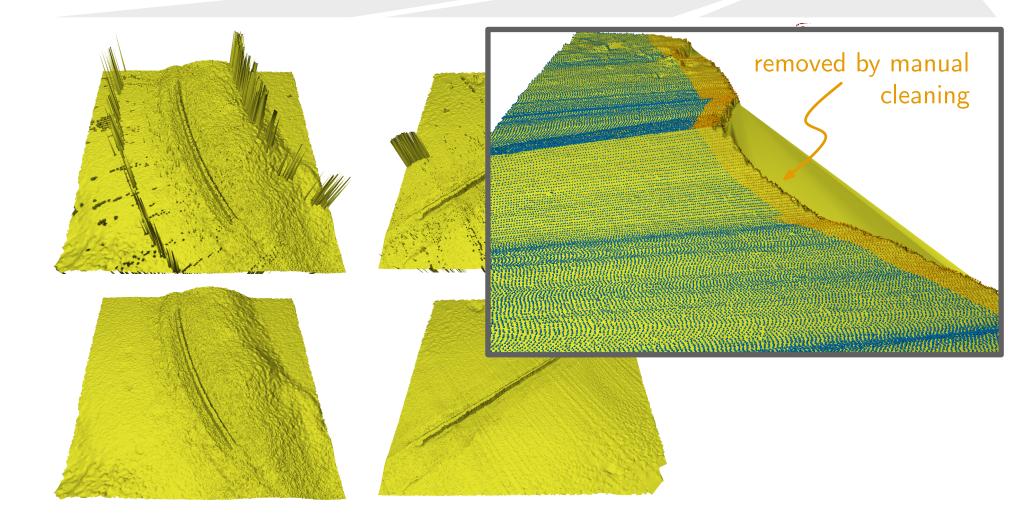
Noise	little	some	much
Threshold	5 cm	5 cm	35 cm
Manually removed: not auto.	0.4%	13%	18%
Not manual. removed: only auto.	0.4%	0.3%	0.8%



Some numbers

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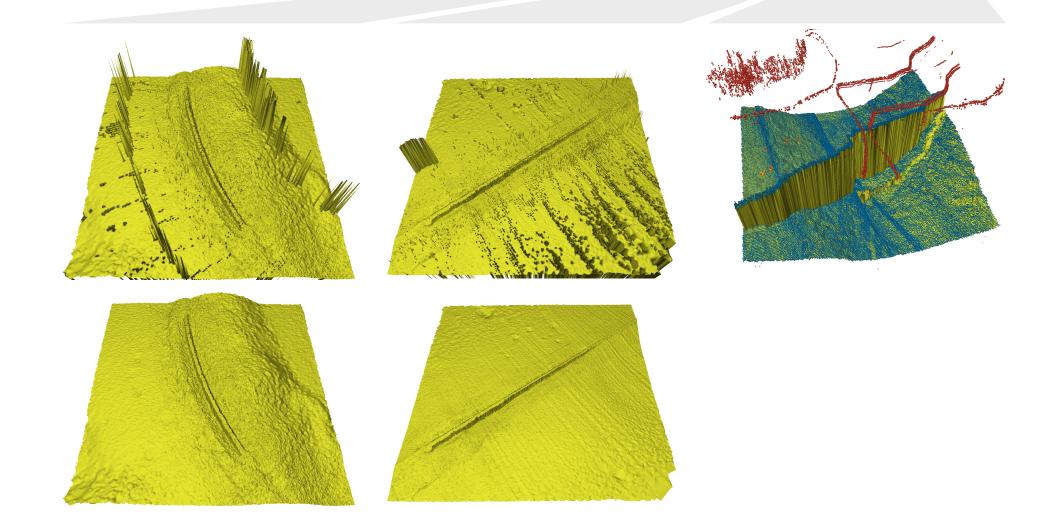
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Conclusion, future work

Implemented in commercial product

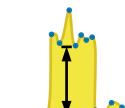


Conclusion, future work

Implemented in commercial product

- \circ Open problem: defining theoretical model of outlier noise
 - Objective theoretical performance analysis
 - Compare Delaunay triangulation with other neighbourhood graphs





Implemented in commercial product

- Open problem: defining theoretical model of outlier noise
 - Objective theoretical performance analysis
 - Compare Delaunay triangulation with other neighbourhood graphs
- Open problem: find easier alternative to Delaunay triangulation Requirements:
 - Good connectivity
 - Fast to compute

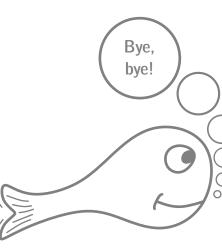




Implemented in commercial product

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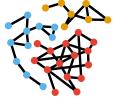
12-4







Connected component algorithm



Compute connected component labelling:
 vertices have equal labels ⇔ they are in the same connected component

Connected component algorithm

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- Algorithm: two phases, sweeping over edge & vertex lists
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 - Up phase: compute final component labels

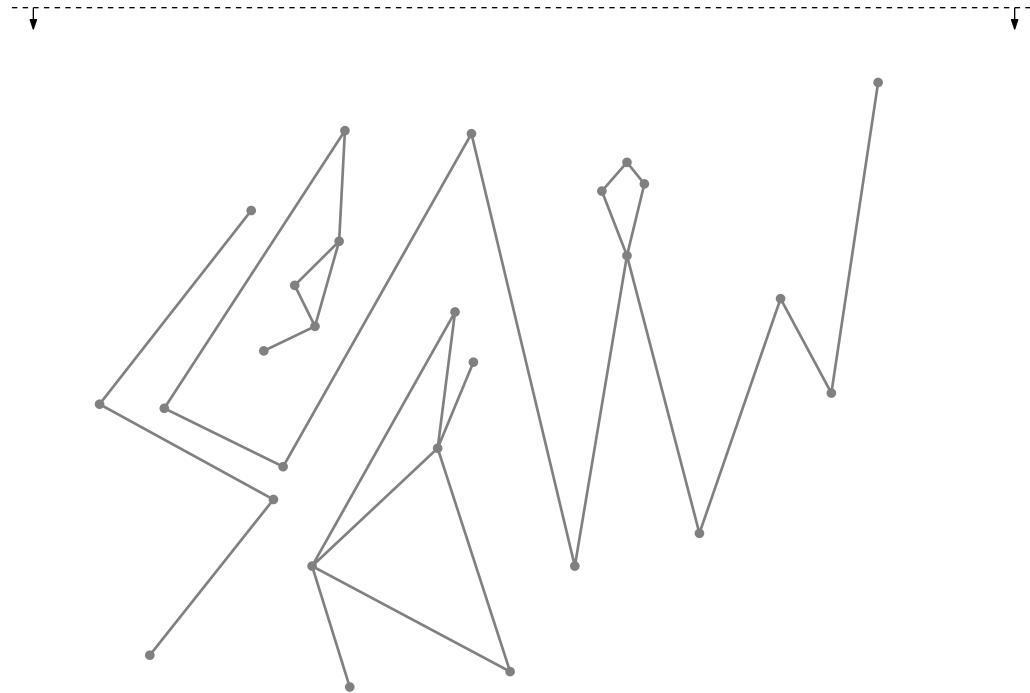
Assumption: edges intersecting sweep line always fit in main memory

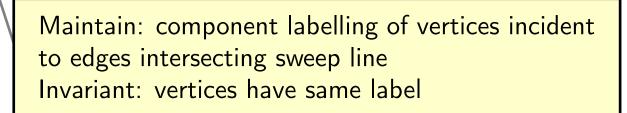
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 - Down phase: augment some vertices with additional connectivity info.
 - Up phase: compute final component labels

Assumption: edges intersecting sweep line always fit in main memory

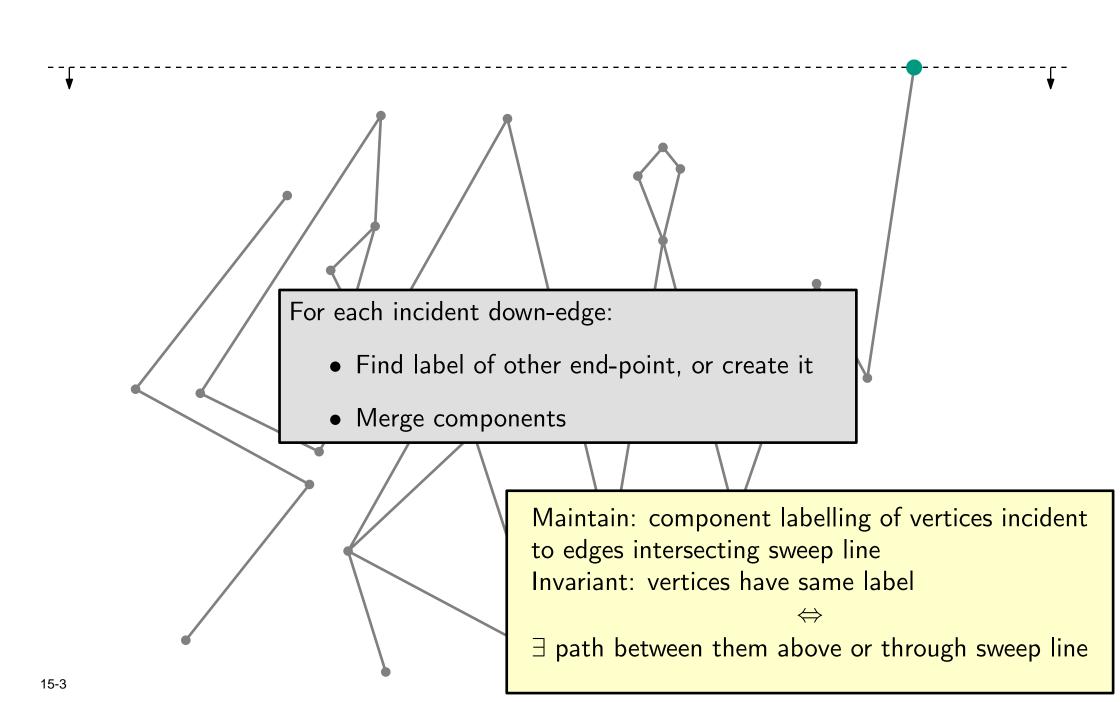
• Total number of I/Os necessary: O(sort(N))

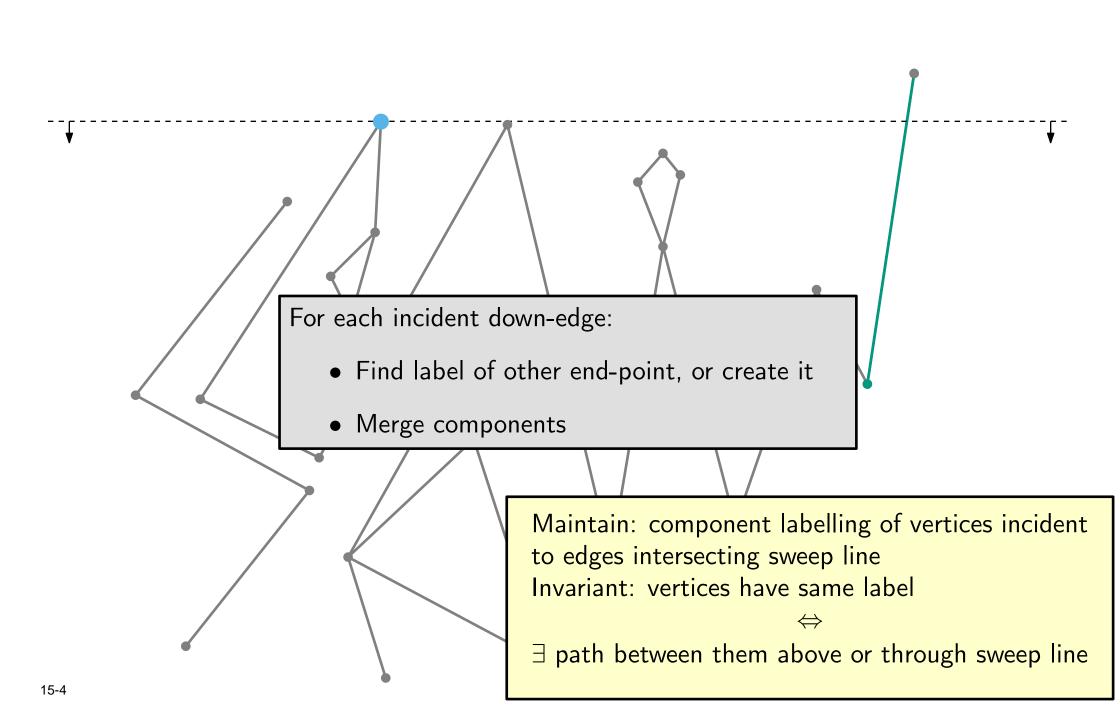


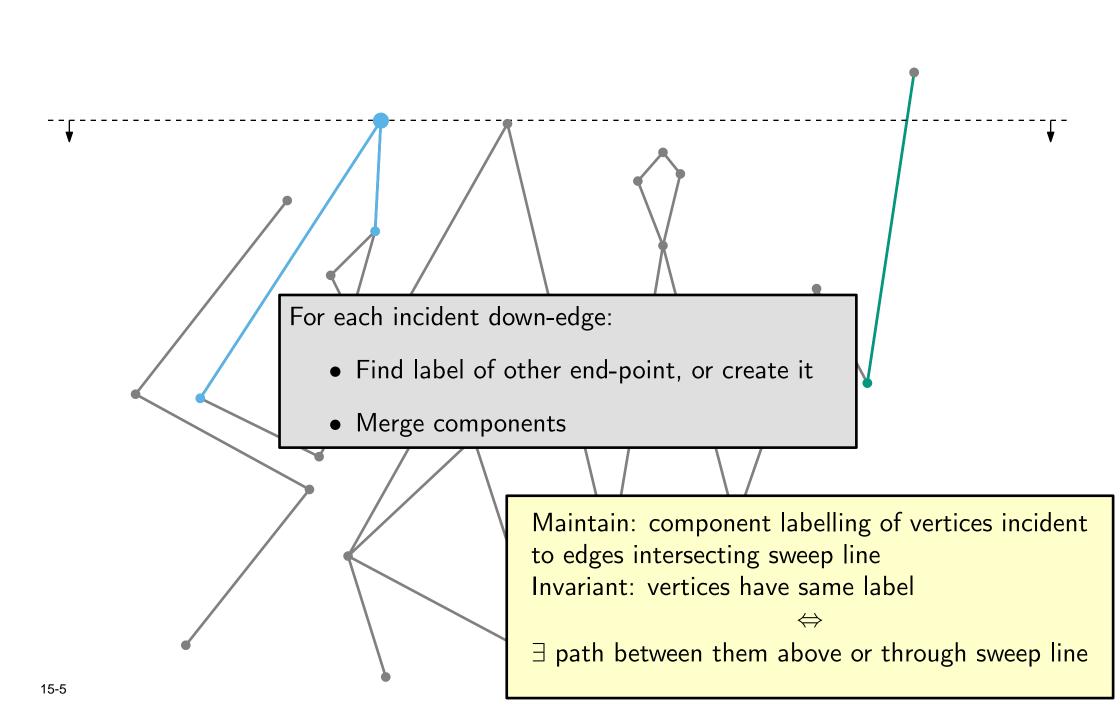


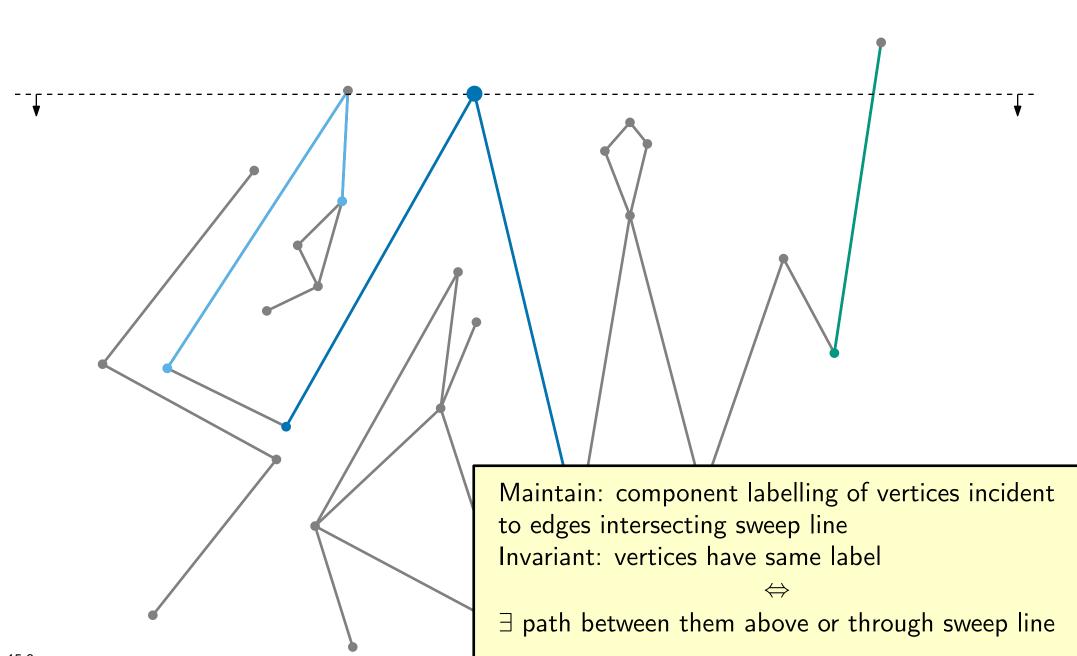
 \exists path between them above or through sweep line

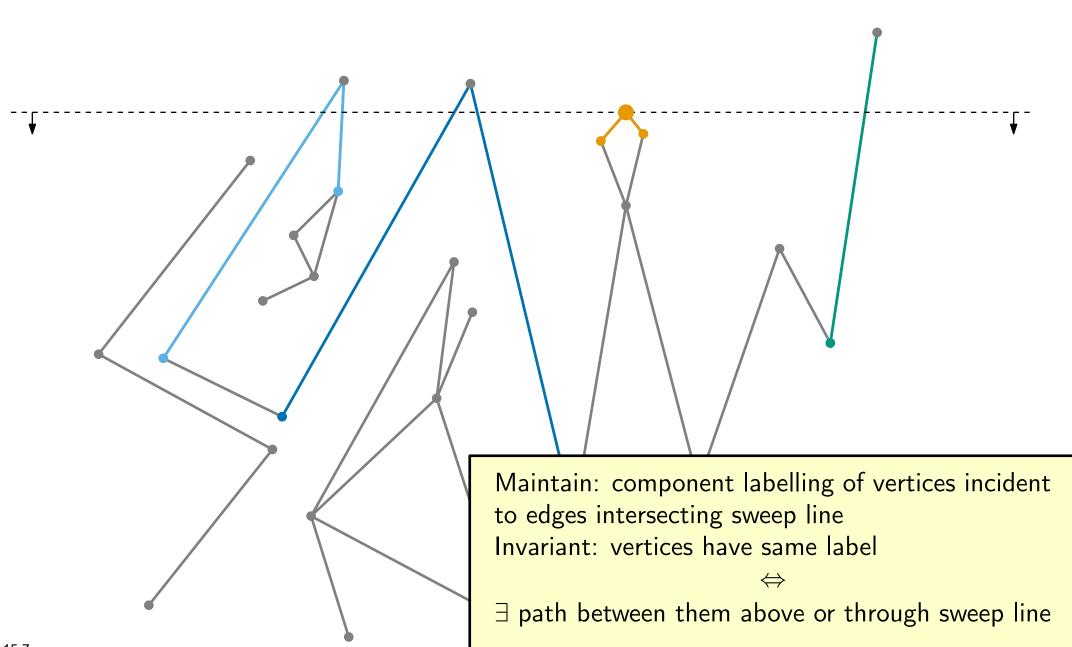
 \Leftrightarrow

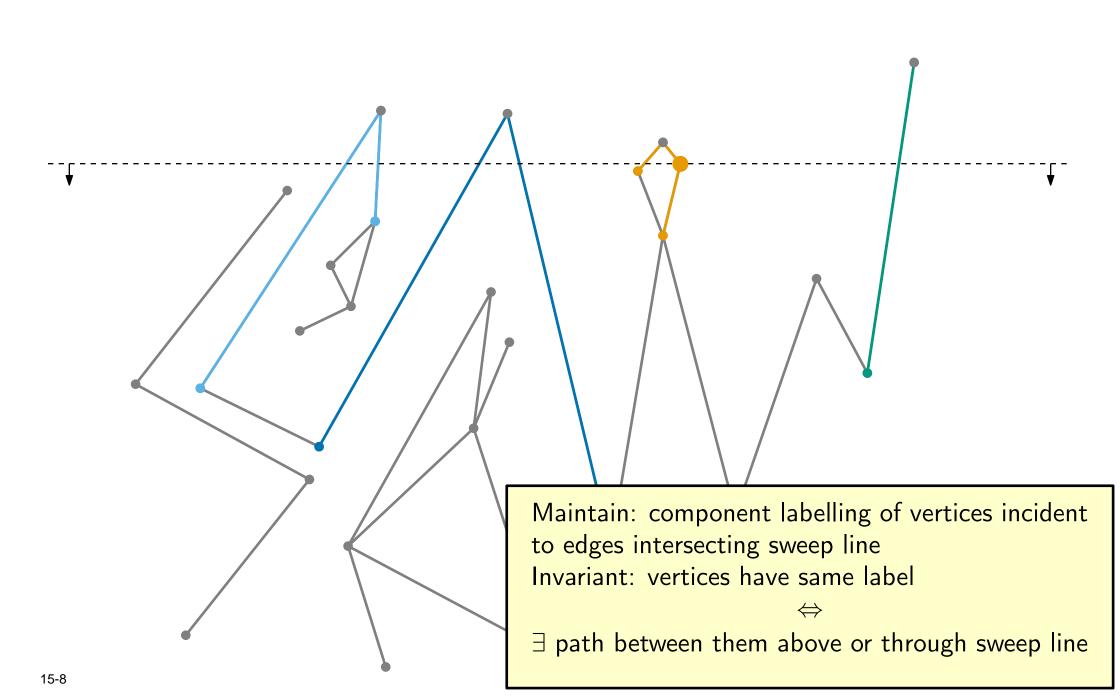


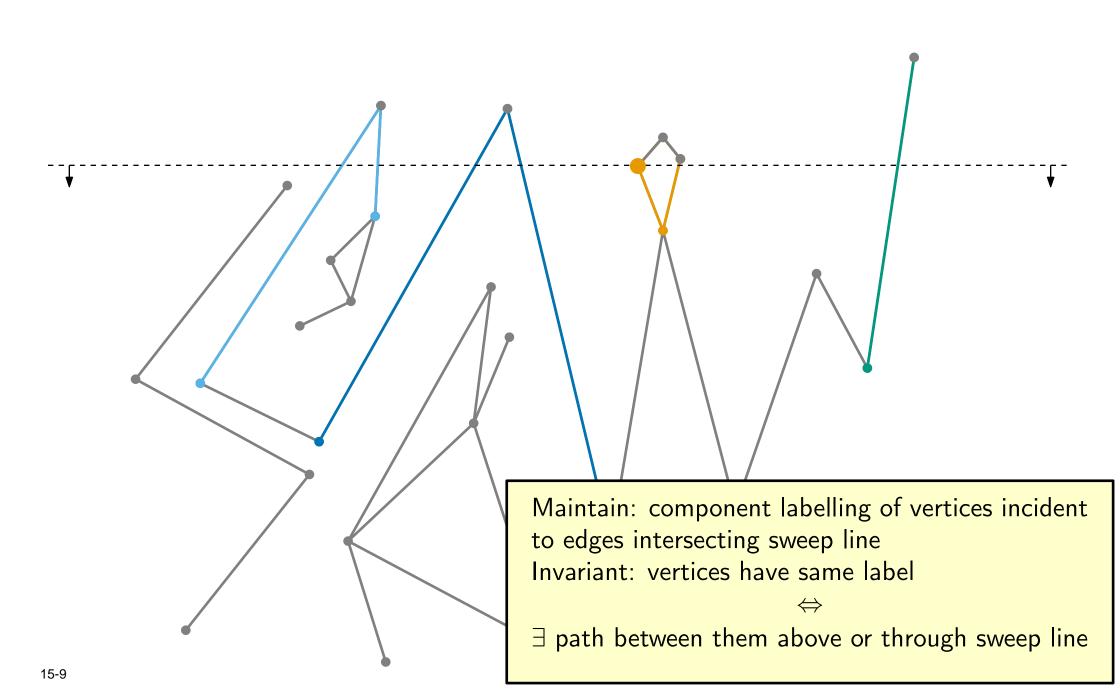


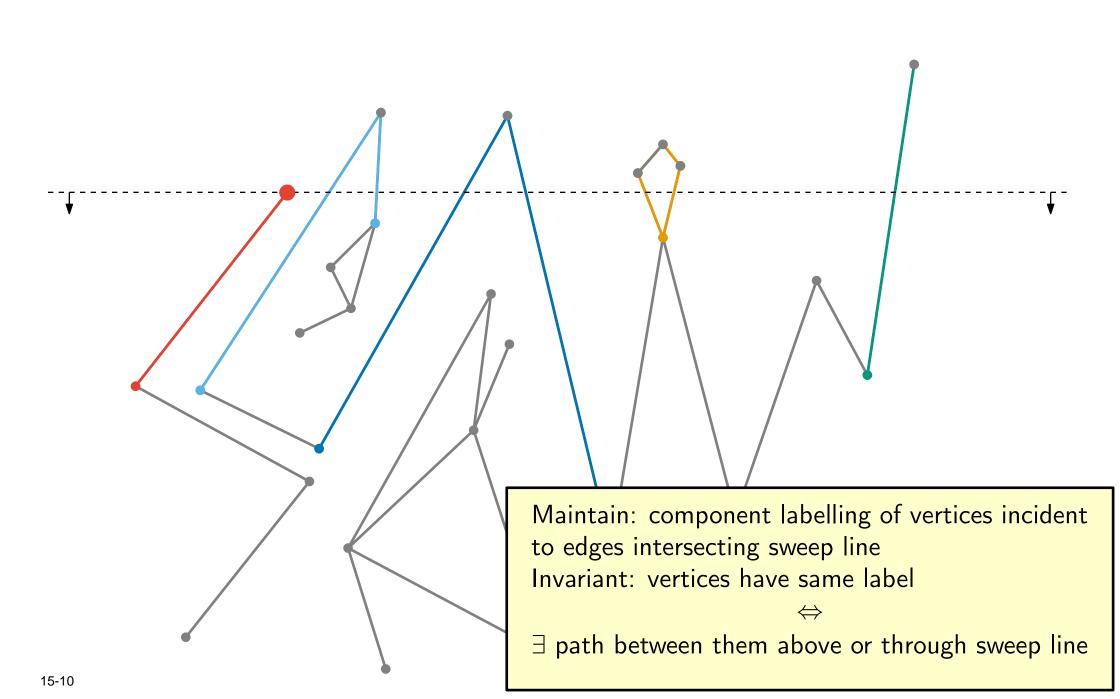


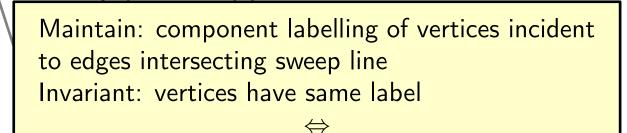












 \exists path between them above or through sweep line

