

A parallel algorithm for computing the flow complex

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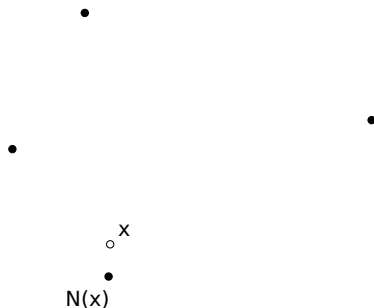
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The point set



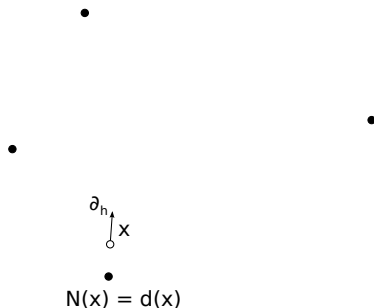
Let $P \subset \mathbb{R}^d$ be the point set in general position, with at least $d + 1$ points.

The distance function



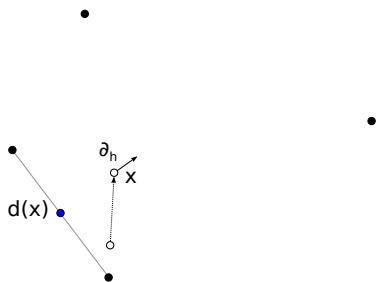
- ▶ distance function $h : \mathbb{R}^d \ni x \mapsto \min_{p \in P} \|x - p\|$
- ▶ Neighbors of x $N(x) = \{p \in P \mid h(x) = \|x - p\|\}$

Gradient of the distance function



- ▶ driver $d(x)$ is the center of the smallest enclosing ball of $N(x)$
- ▶ gradient of $h(x)$, $\partial_h(x) = \frac{x-d(x)}{h(x)}$

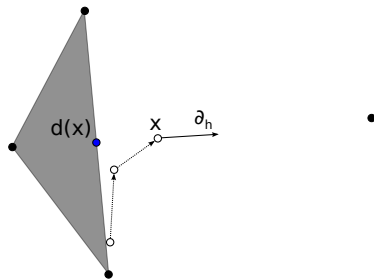
The flow



- ▶ flow $\phi : [0, \infty) \times \mathbb{R}^d \rightarrow \mathbb{R}^d$
- ▶ defined by $\phi(0, x) = x$, and

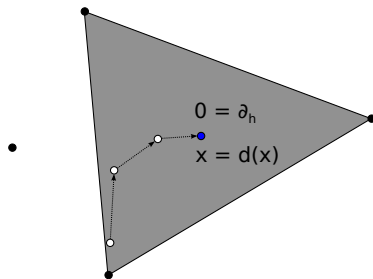
$$\lim_{t \downarrow t_0} \frac{\phi(t, x) - \phi(t_0, x)}{t - t_0} = \partial_h(\phi(t_0, x))$$

The Flow line



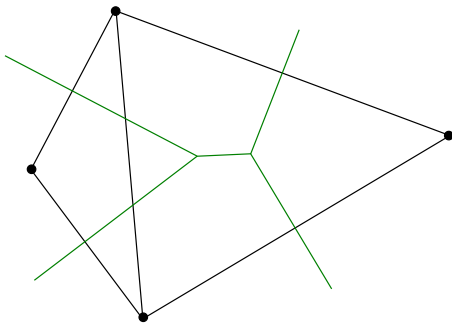
- ▶ orbit/flow-line of a point x : $\phi(x) = \{\phi(t, x) | t \geq 0\}$

Critical point

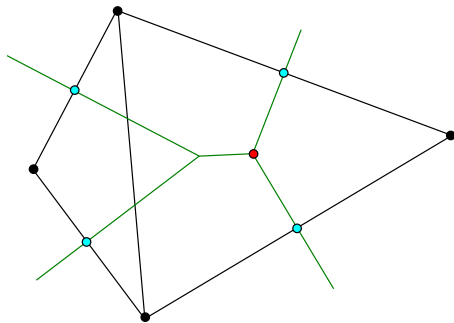


- ▶ $x \in \mathbb{R}^d$: $\partial_h(x) = 0$
- ▶ $x \in \text{conv}(N(x))$
- ▶ index of a critical point is the dimension of the affine hull of $N(x)$
- ▶ Minima with $i(x) = 0$, and Maxima with $i(x) = d$
- ▶ the remaining critical points are saddle points

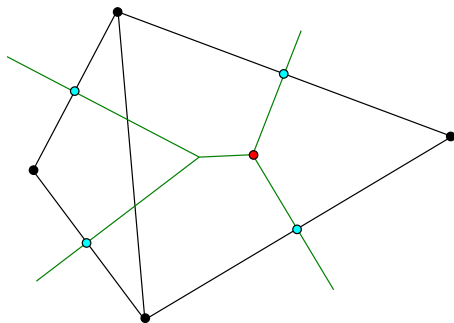
Flow complex



Flow complex



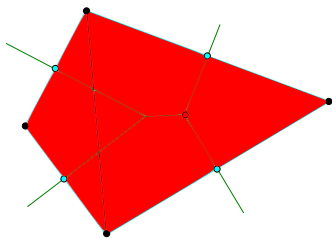
Flow complex



$$\sum_{i=0}^d (-1)^i \cdot n_i = 1$$

- ▶ n_i is the number of critical points of index i

Stable and unstable manifolds



- ▶ *stable manifold* of x contains the points, that flow into x

$$S(x) = \{y \in \mathbb{R}^d \mid \lim_{t \downarrow \infty} \phi(t, y) = x\}$$

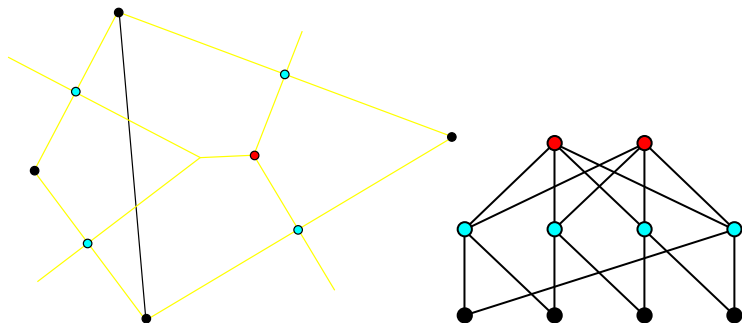
- ▶ given a neighborhood U around x

$$V(U) = \{y \in \mathbb{R}^d \mid \exists z \in U, t \geq 0 : \phi(t, z) = y\}$$

- ▶ *unstable manifold*

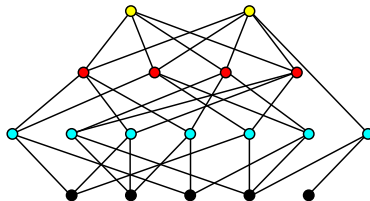
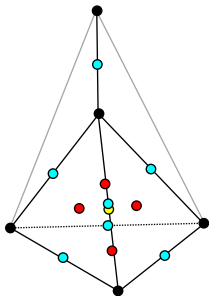
$$U(x) = \bigcap_{\text{Neighborhood } U \text{ of } x} V(U)$$

The Hasse diagram

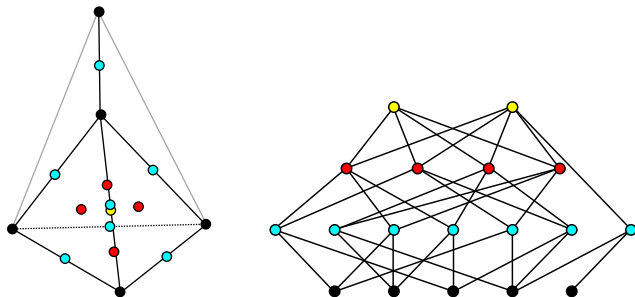


- ▶ for two critical points x and y , $S(y)$ is **incident** to $S(x)$, if $U(y) \cap S(x) \neq \emptyset$, i.e. there is a point in $U(y)$ that flows into x

The algorithm



The algorithm



Two basic operations

1. Ascent: increasing $h(x)$ along the gradient
2. Descent: decreasing $h(x)$ along the gradient

The primitives

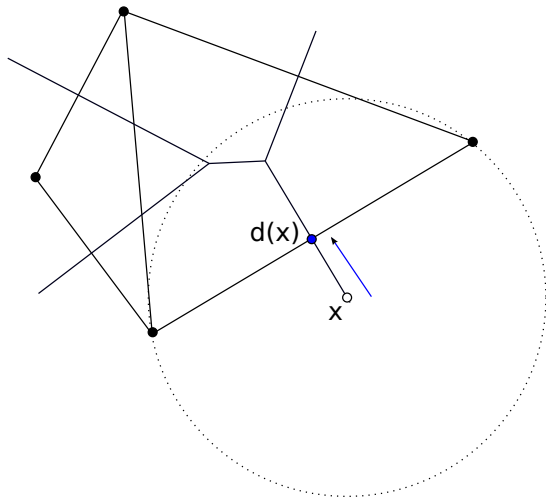
Nearest neighbor along a ray: Let v be the ray-vector, $p \in V$ and $q \in P \setminus V$. Solve $\|(x + t \cdot v) - q\|^2 = \|(x + t \cdot v) - p\|^2$, and pick the minimum $t_q > 0$, along with q

Projection onto an affine hull: In order to check for $x \in \text{CONV}(N(x))$, we compute the coefficients λ_i for which

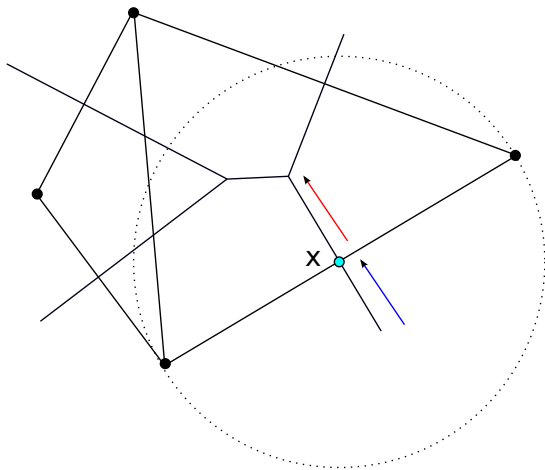
$$c = \sum_i \lambda_i \cdot p_i, \quad p_i \in V, \quad \sum \lambda_i = 1$$

Here c is the orthogonal-projection of x onto V . We use a dynamic QR-decomposition to solve the least-squares problem of computing the projection of x onto V .

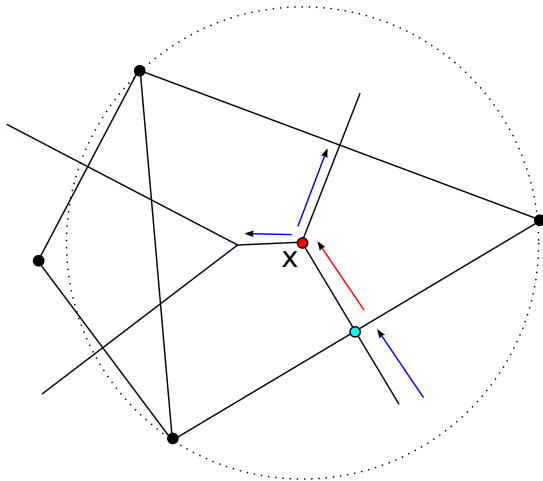
Example



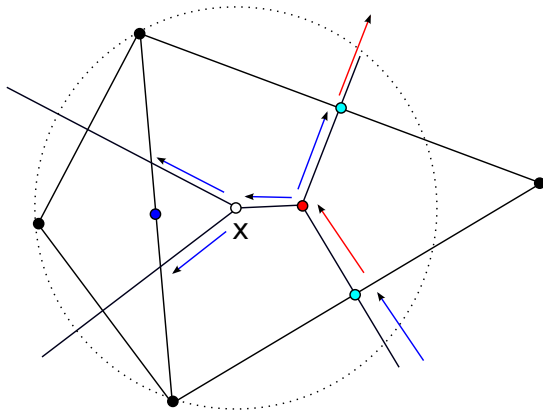
Example



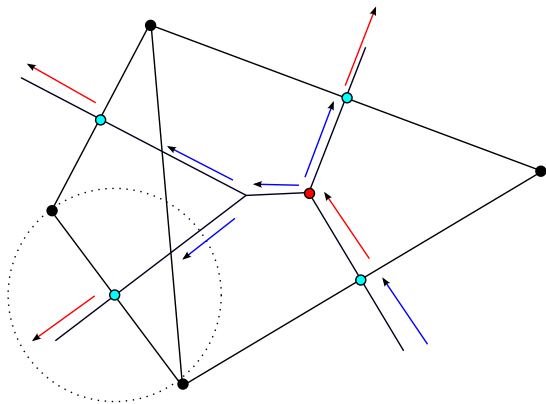
Example



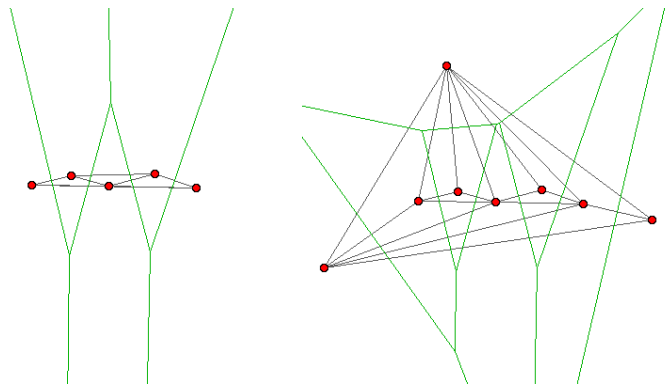
Example



Example



Bounding simplex



- ▶ low-dimensional(sub-)structures serialize the exploration
- ▶ a bounding simplex allows for a scalable parallel exploration

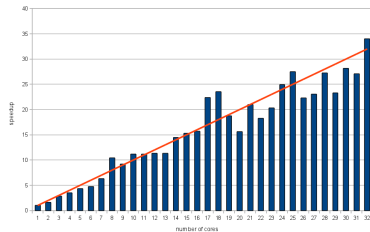
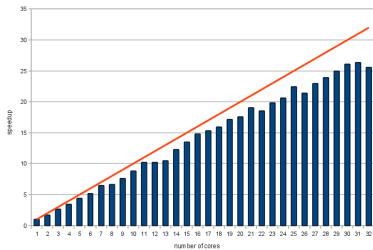
Experimental results

Critical point statistics

data set	dimension	distribution	time (s)
random within sphere	3	50-137-122-34	0.03
random within sphere	4	50-199-284-178-44	2.66
curve	3	50-73-56-32	0.7
spiral-3d	3	30-64-61-26	0.03
spiral-3d	4	30-41-12-0-0	0.97
spiral-3d	5	30-41-12-0-0-0	182.4
beethoven-3d-model	3	500-1076-649-72	11.81
cup-3d-model	3	300-766-599-132	0.92

Experimental results

Scalability



- ▶ On the left: 200 random points in 4 dimensions
- ▶ On the right: 100 random points in 5 dimensions

Conclusions

- ▶ a new algorithm to compute the flow complex, without explicit computation of the Delaunay triangulation
- ▶ numerical robustness
- ▶ it is inherently parallelizable, and scales well with the number of cores in a shared-memory architecture
- ▶ future work: high-level scalability by subdivision

Thank you!

Questions?