

Computing Positively Weighted Straight Skeletons of Simple Polygons Using an Induced Line Arrangement

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Related Work

- The straight skeleton was introduced by Aichholzer et al. 1995 [1].
- Eppstein and Erickson [3] introduced in 1999 an algorithm with the current best worst-case complexity: For a simple polygon (with holes) it requires $\mathcal{O}(n^{17/11+\epsilon})$ time and space to compute the (weighted) straight skeleton.
- More recent results with lower time/space-complexity are known [2, 6]¹ but only for (unweighted) straight skeletons.

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- Our work is based on the work of Huber and Held (IJCGA 2012) on straight-skeleton computation based on motorcycle graphs [4].
- Using an *extended wavefront* they transform split events into edge events, shifting the complexity to another event.
- We revisit their work and show required changes to apply their approach to the weighted scenario.

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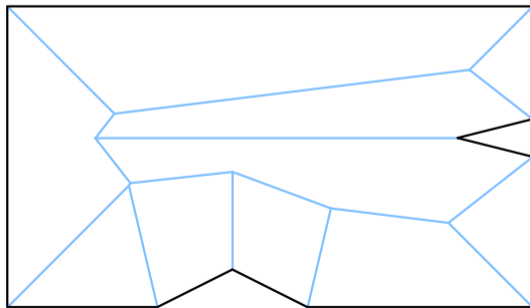
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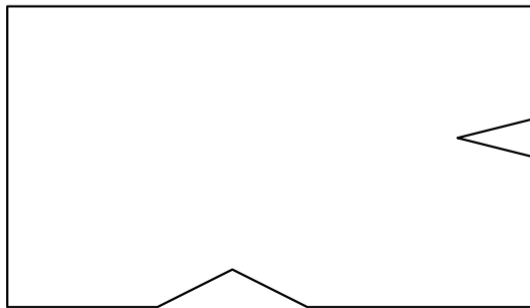
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- Consists only of straight line segments.
- Defined by a propagation process:
 - Edges move inwards in a parallel manner at unit speed.
 - The vertices of the *wavefront polygons* trace out arcs.
 - Two events: edge event and split event



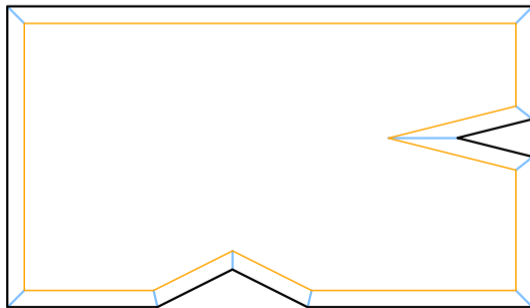
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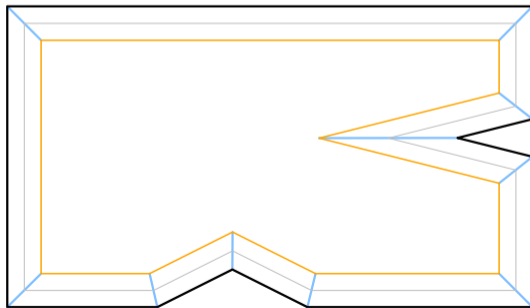
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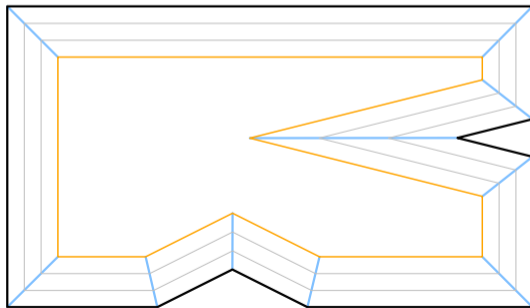
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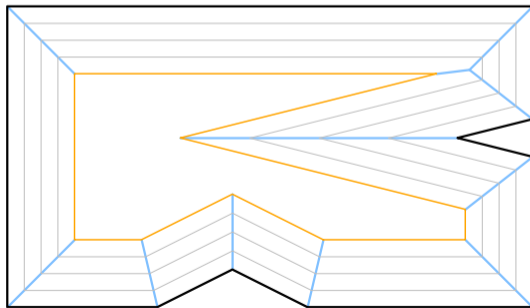
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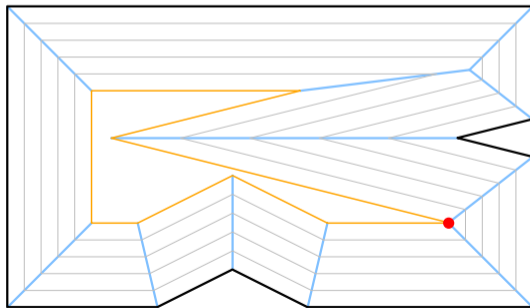
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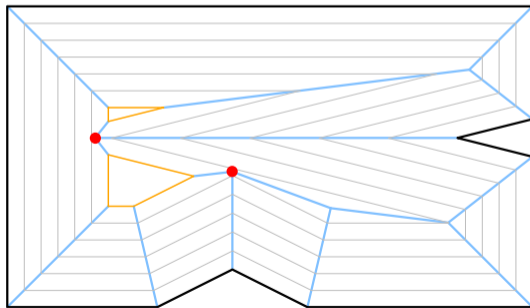
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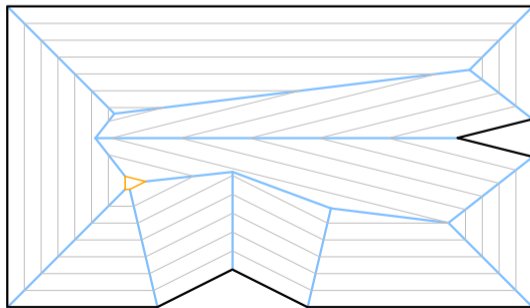
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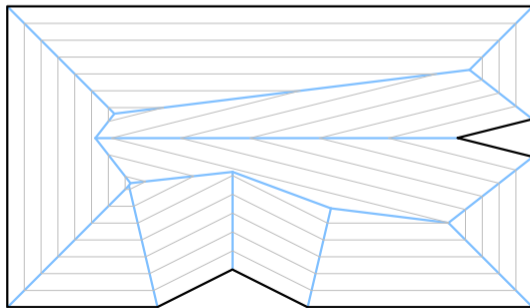
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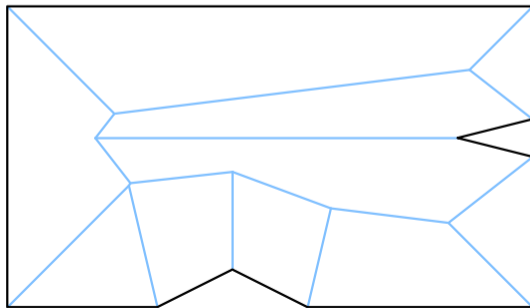
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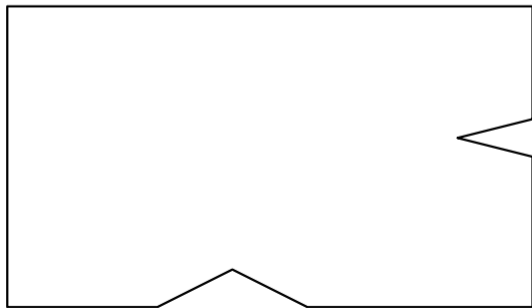
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 - Two events: edge event and split event
 - The *straight skeleton* consists of the blue arcs.



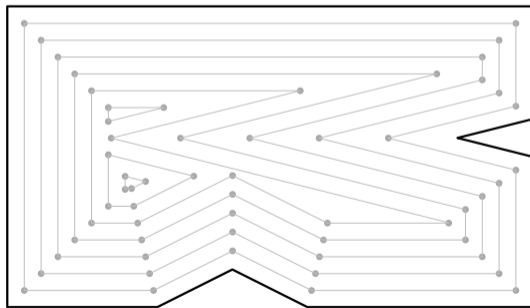
Preliminaries

- Let P a simple polygon with n vertices, r of which are reflex.
- Then $\mathcal{W}_P(t)$ is the wavefront of P at time t .
- After all components of $\mathcal{W}_P(t)$ have vanished we call the traces of the vertices of $\mathcal{W}_P(t)$ over the time-span of the propagation straight skeleton $\mathcal{S}(P)$.



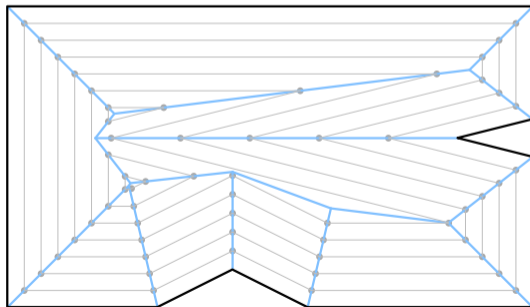
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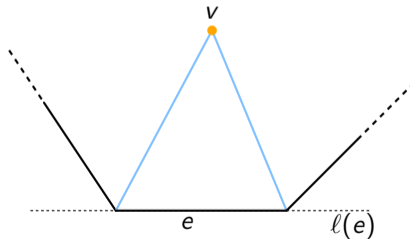
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Straight Skeletons – Finding Events

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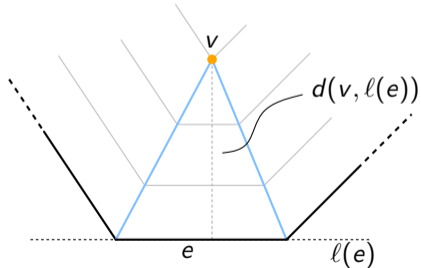
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- Enqueue all events in priority queue.
- $\mathcal{O}(n \log n)$ time and $\mathcal{O}(n)$ space.



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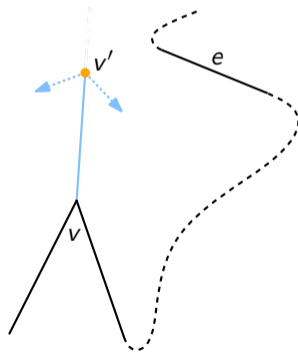
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Split Events

- Let v be a reflex vertex of P . Then there are $\mathcal{O}(n)$ possible split events for $v(t)$ in $\mathcal{W}_P(t)$.
- Enqueue event v' closest to v in $\mathcal{O}(n \log n)$ time.
- If v' is not reached again $\mathcal{O}(n \log n)$ for v'' .
- We may miss $\mathcal{O}(n)$ times.
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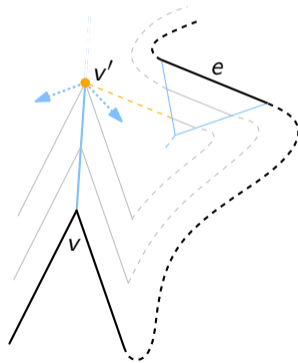
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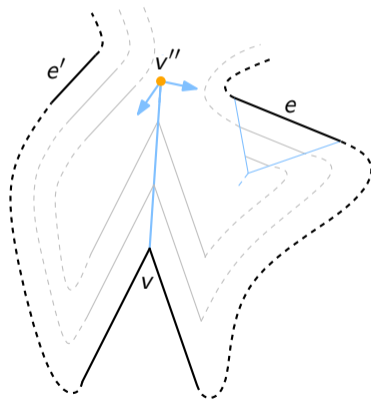
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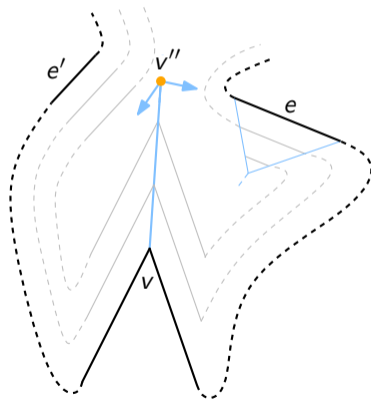
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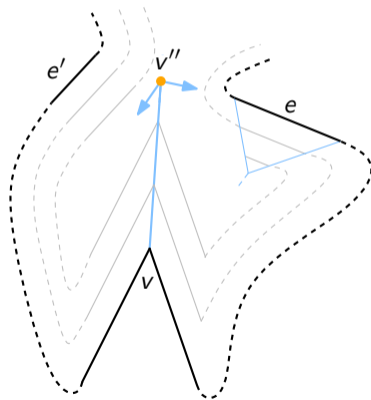
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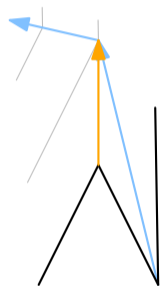
- When two reflex wavefront vertices meet at a common point².



²Related to vertex event [3].

Tracing Reflex Arcs

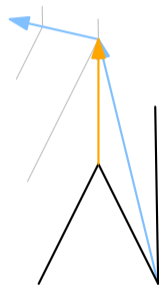
- Every event, where a reflex wavefront vertex is involved, reduces the number of reflex vertices in $\mathcal{W}_P(t)$.
- *Can we know these reflex arcs in advance?*



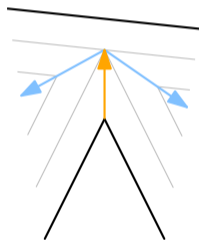
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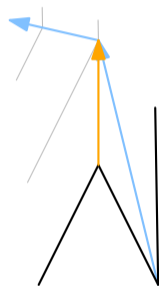
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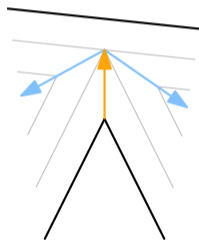
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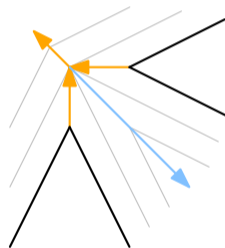
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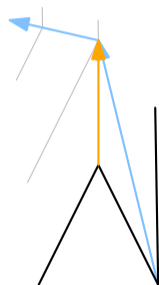


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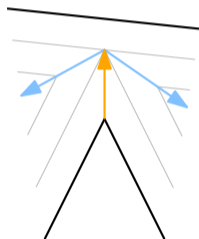
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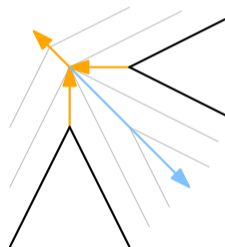
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Motorcycle Graph and $\mathcal{S}(P)$

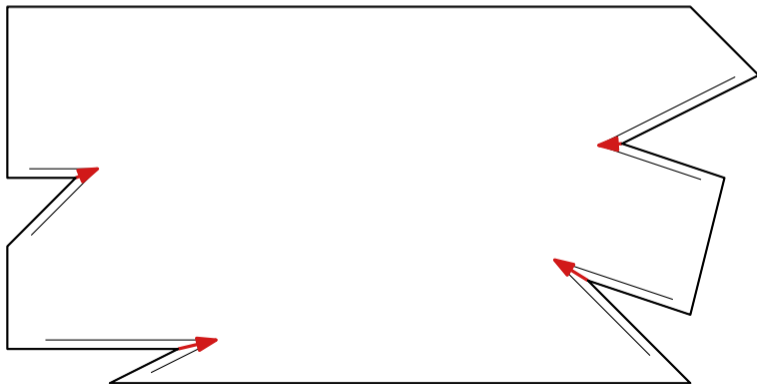
- Start a motorcycle m for every reflex vertex v of $\mathcal{W}_P(t)$ such that m inherits the velocity vector of v .
- Every motorcycle leaves a trace behind and stops if it crashes into another trace or the polygon boundary.
- The *motorcycle graph* $\mathcal{M}(P)$ ³ is formed when all motorcycles have crashed.



³Introduced by Eppstein and Erickson [3]

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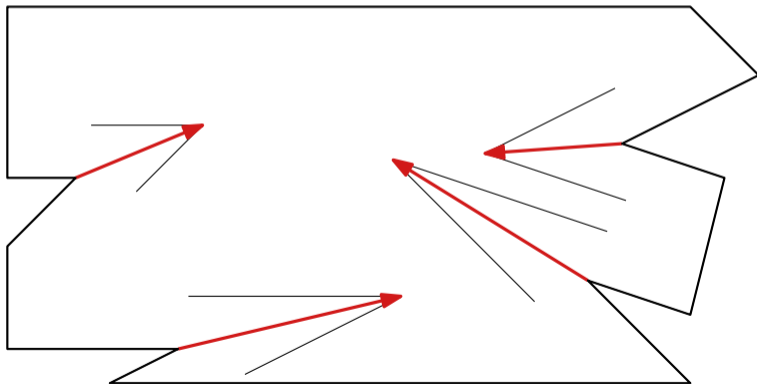
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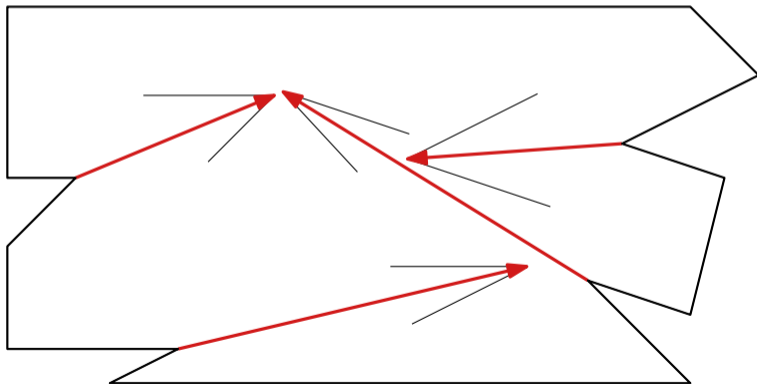
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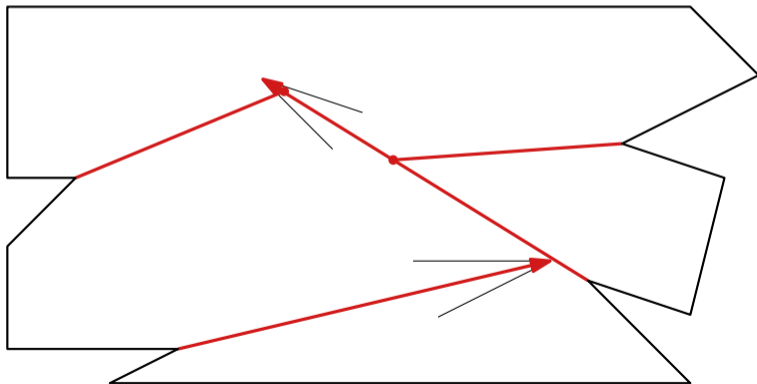
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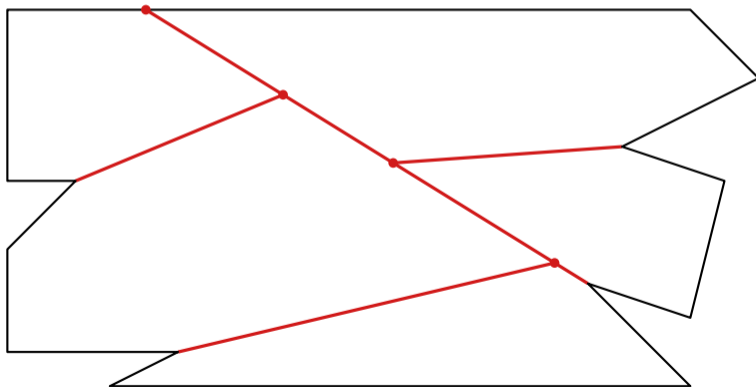
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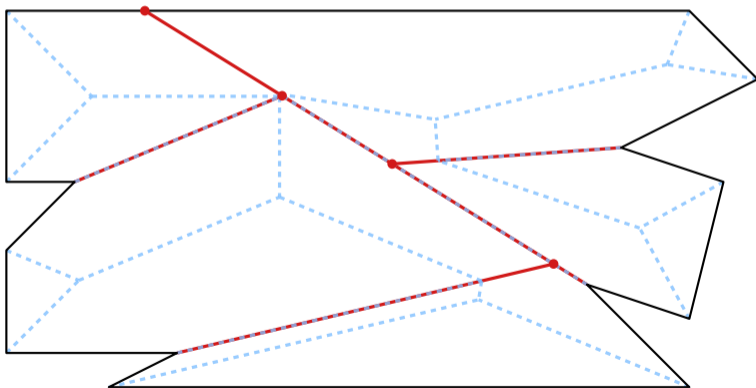
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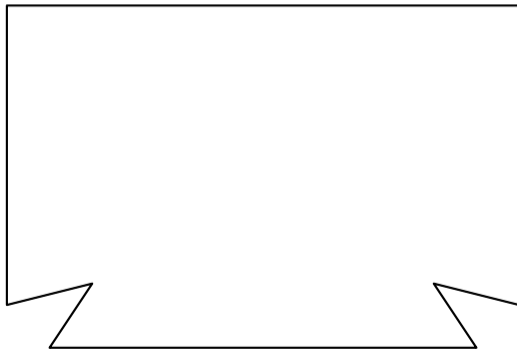
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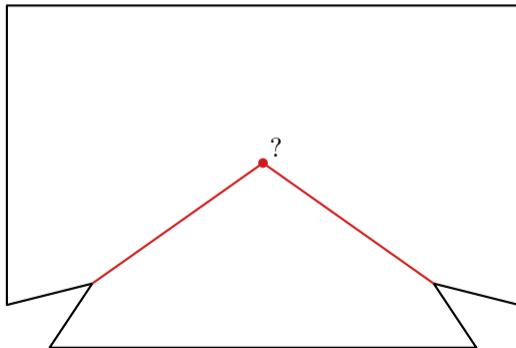
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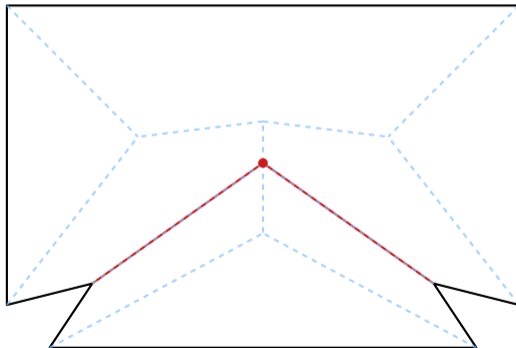
- $\mathcal{M}'(P)$ covers “all” reflex arcs of $\mathcal{S}(P)$ ⁴, such that no two motorcycles reach the same point simultaneously.
- Huber and Held [4] allow motorcycles to have different starting times and call it *generalized motorcycle graph* $\mathcal{M}(P)$.



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Motorcycle Graph and $\mathcal{S}(P)$ (cont'd)

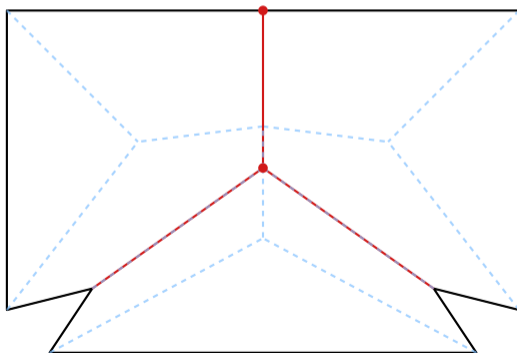
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⁴Various publications [2, 3]

Motorcycle Graph and $\mathcal{S}(P)$ (cont'd)

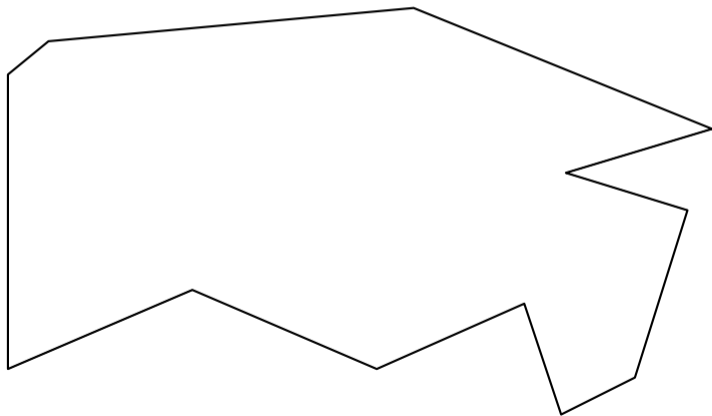
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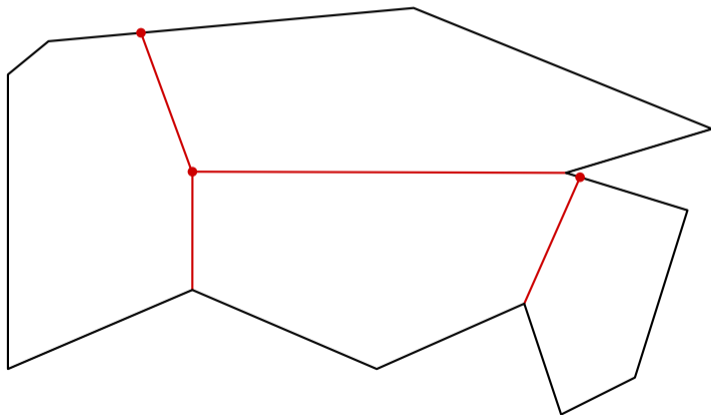
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- $\mathcal{W}_P^*(t)$ consists of the wavefront at time t and the portion of $\mathcal{M}(P)$ that lies inside that wavefront.
- All regions in $\mathcal{W}_P^*(t)$ are convex and all events are edge events.



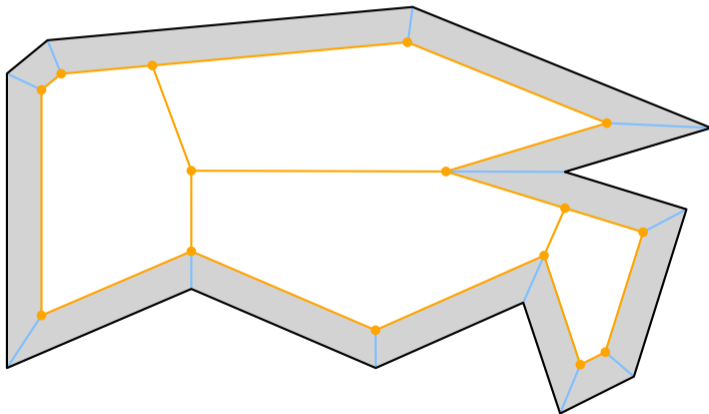
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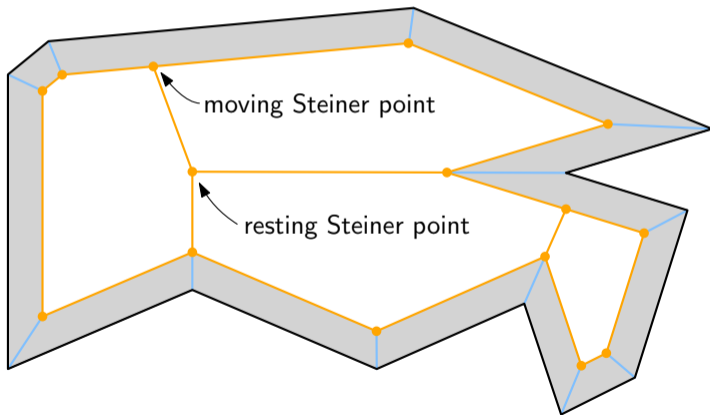
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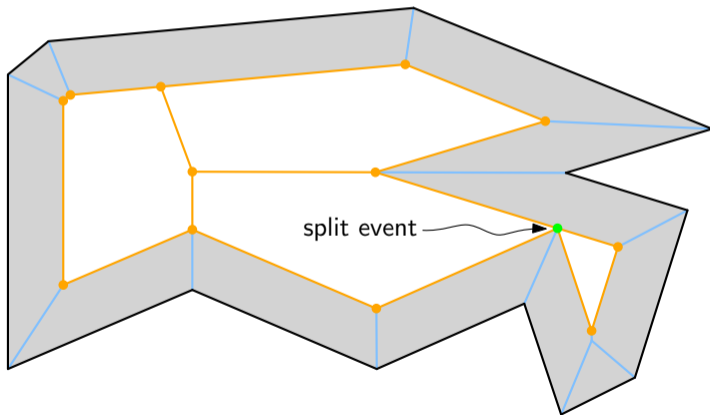
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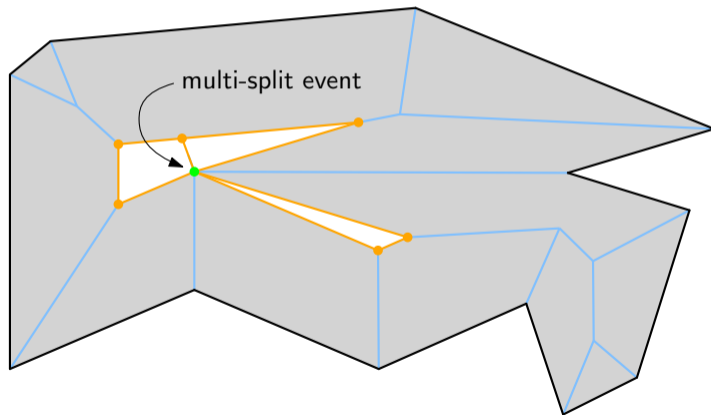
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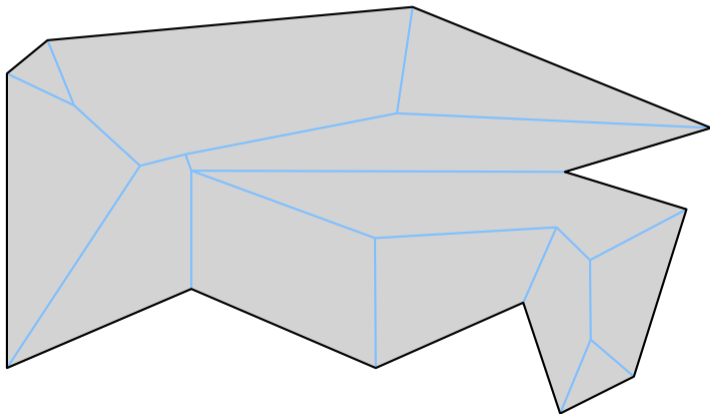
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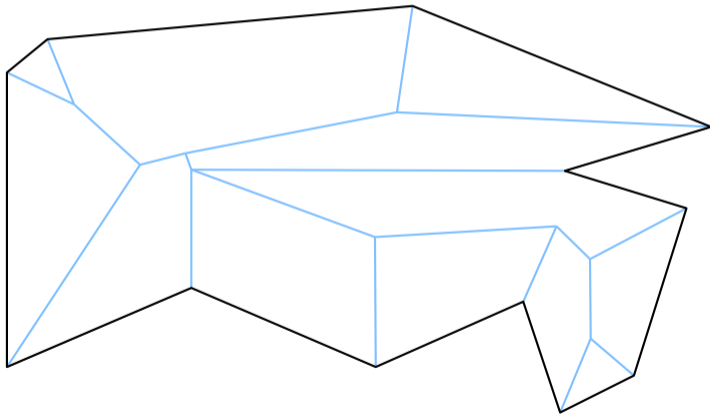
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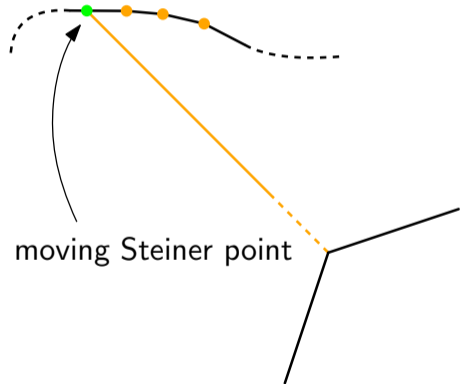
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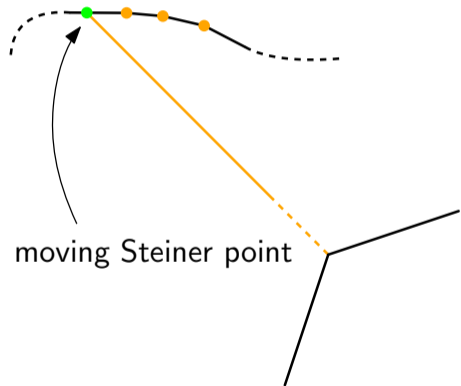
Extended Wavefront \mathcal{W}_P^* (cont'd)

- Using $\mathcal{W}_P^*(t)$ one can compute $\mathcal{S}(P)$ in $\mathcal{O}((n + nr) \log n)$ time and linear space.
- The number of *switch events*, i.e., when a wavefront vertex meets a moving Steiner point (where the arc of a motorcycle edge ends) is in $\mathcal{O}(nr)$.



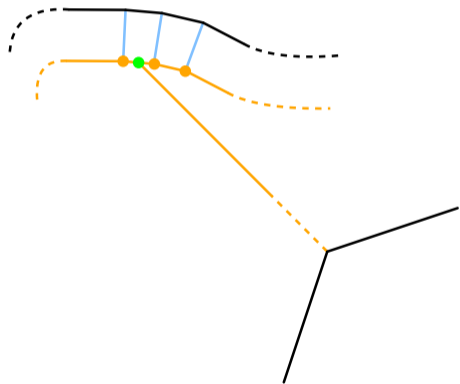
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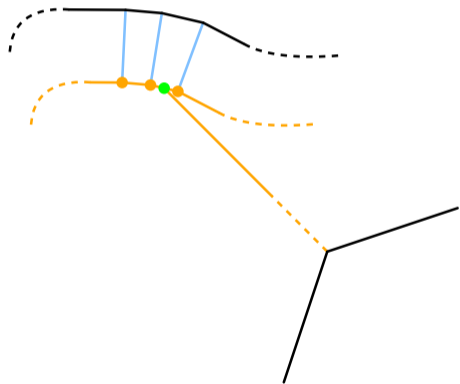
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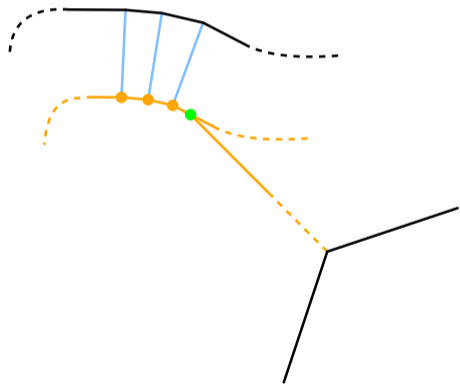
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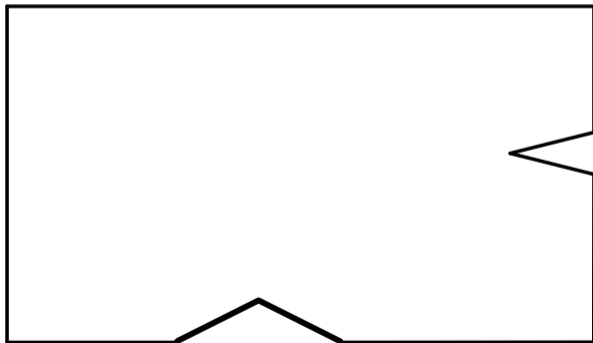
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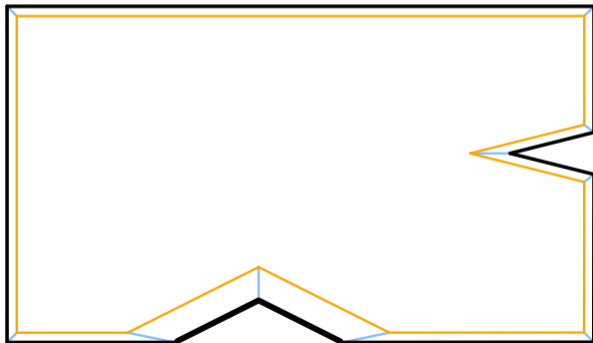
Weighted Straight Skeleton $\mathcal{S}(P, \sigma)$

- Wavefront $\mathcal{W}_P(t, \sigma)$ traces out the weighted straight skeleton $\mathcal{S}(P, \sigma)$, s.t. the weight function $\sigma \in \mathbb{R}^+$ provides the strictly positive edge weights.
- Thick edges have σ of about 3, unit weight otherwise.
- $\mathcal{S}(P)$ (left) and $\mathcal{S}(P, \sigma)$ (right).



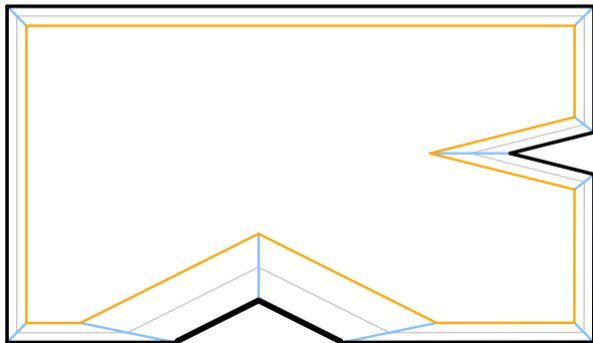
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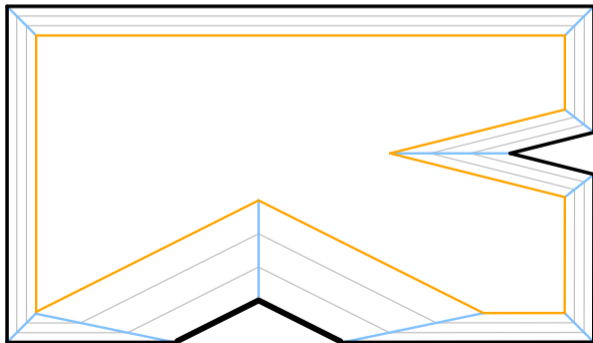
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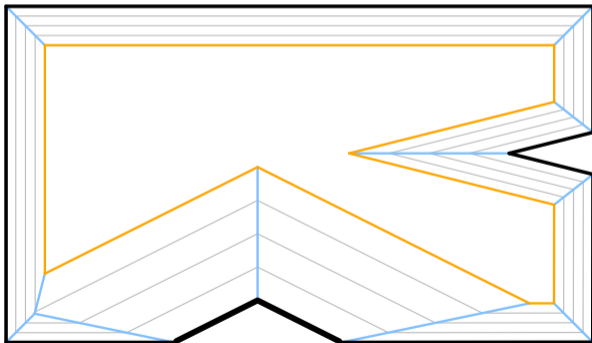
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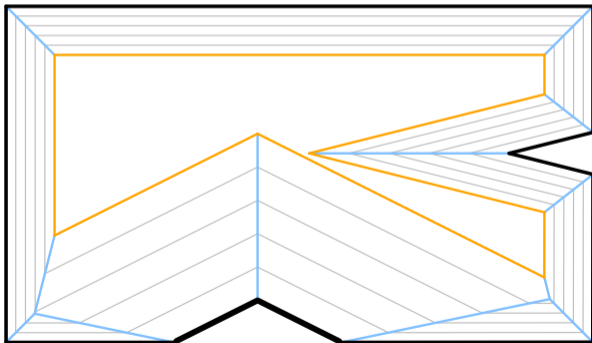
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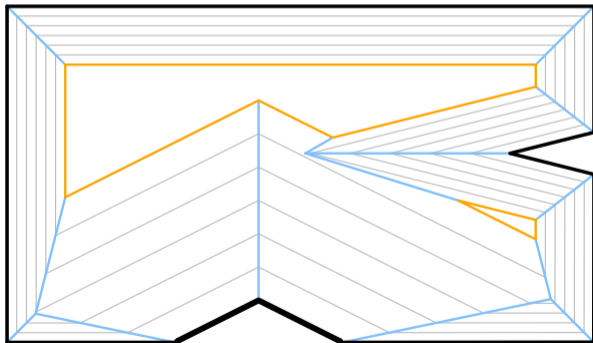
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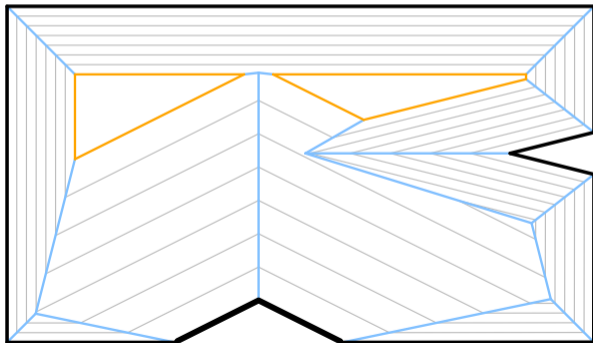
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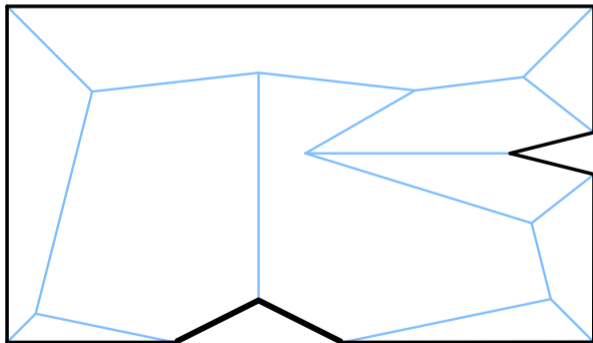
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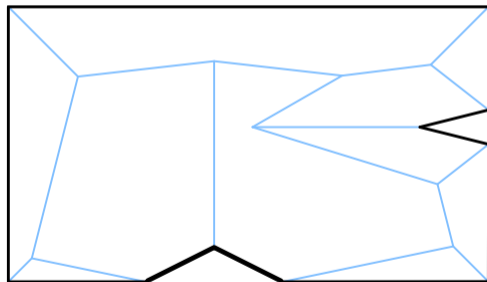
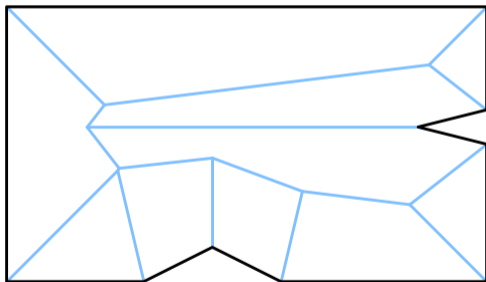
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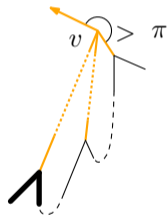
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Weighted Straight Skeleton (cont'd)

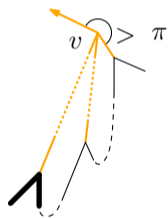
- Reflex vertices of $\mathcal{W}(P, \sigma)$ are not limited by first event.
- $\mathcal{M}(P)$ covers only initial *reflex* arcs of $\mathcal{S}(P, \sigma)$.
- $\mathcal{M}(P)$ may have to be updated n times.
- Updating $\mathcal{M}(P)$ results in re-computation.



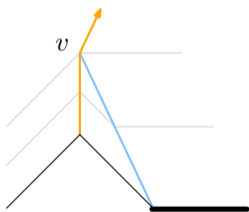
mutli-split event

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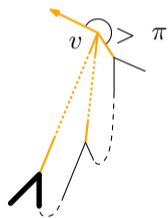
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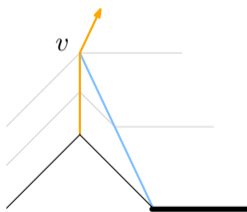
edge event

Weighted Straight Skeleton (cont'd)

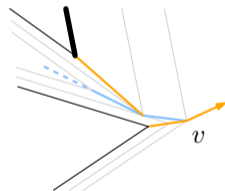
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multi-split event



edge event

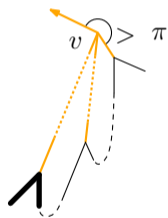


edge event¹

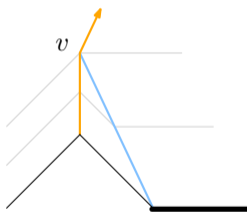
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Weighted Straight Skeleton (cont'd)

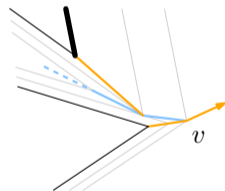
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multi-split event



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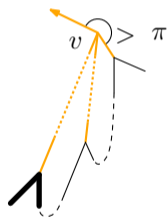


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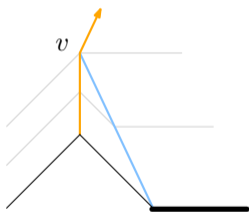
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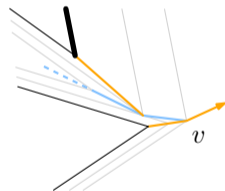
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mutli-split event



edge event



edge event¹

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Induced Line Arrangement $\mathcal{A}(P)$

- For every reflex vertex v of P we construct a line segment s .
- Let s start at v , lie on $v(t)$ of $\mathcal{W}_P(t, \sigma)$, and end at the first intersection with the boundary of P .
- The set of all such line segments forms $\mathcal{A}(P)$.

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Linear or Quadratic Space

- Store all $\mathcal{O}(r^2)$ intersections in a sorted manner: $\mathcal{O}(n^2 \log n)$ time and $\mathcal{O}(r^2)$ space. Obtain the next intersection in $\mathcal{O}(1)$ time.
- Store only the closest intersection to a point for all segments in $\mathcal{A}(P)$, i.e., overall $\mathcal{O}(r)$ intersections. Obtain the next intersection in $\mathcal{O}(r)$ time.

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Space Time Trade-off

- For a fixed k in $1 \leq k \leq r$. Let s in $\mathcal{A}(P)$ and let p the next intersection point on s .
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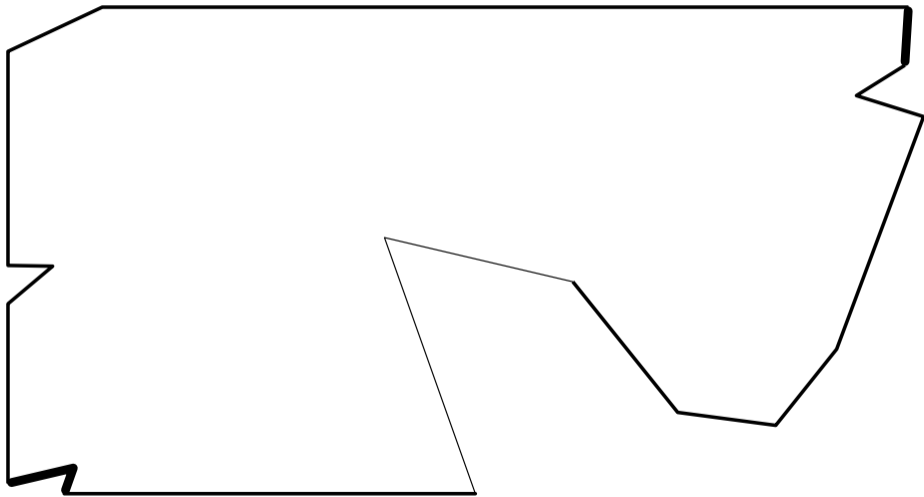
Linear or Quadratic Space

- Store all $\mathcal{O}(r^2)$ intersections in a sorted manner: $\mathcal{O}(n^2 \log n)$ time and $\mathcal{O}(r^2)$ space. Obtain the next intersection in $\mathcal{O}(1)$ time.
- Store only the closest intersection to a point for all segments in $\mathcal{A}(P)$, i.e., overall $\mathcal{O}(r)$ intersections. Obtain the next intersection in $\mathcal{O}(r)$ time.

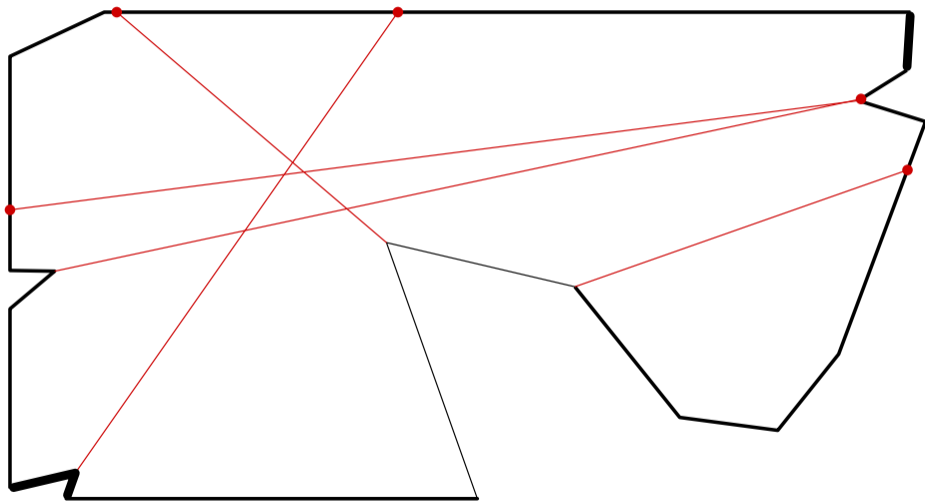
Space Time Trade-off

- For a fixed k in $1 \leq k \leq r$. Let s in $\mathcal{A}(P)$ and let p the next intersection point on s .
- Instead of p we compute and store the next k intersections in $\mathcal{O}(r \log r)$ time.
- On s are at most r intersections points, thus we have to compute the next k intersections at most r/k times.
- We require $\mathcal{O}(kr)$ space for $\mathcal{A}(P)$.

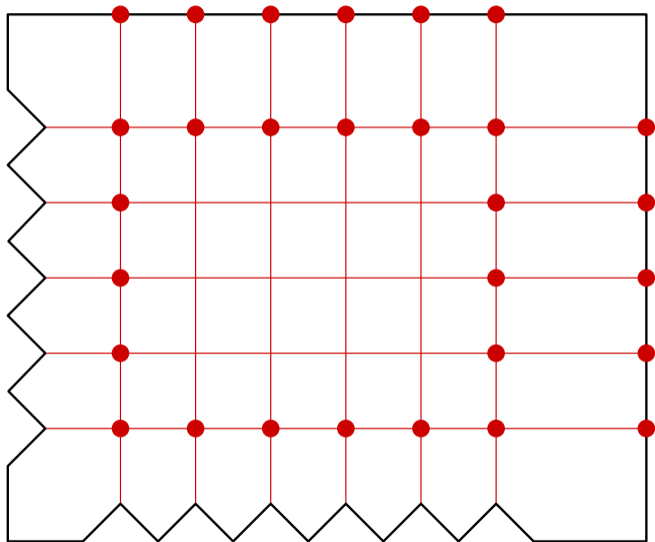
Extended Wavefront $\mathcal{W}_p^*(t, \sigma)$



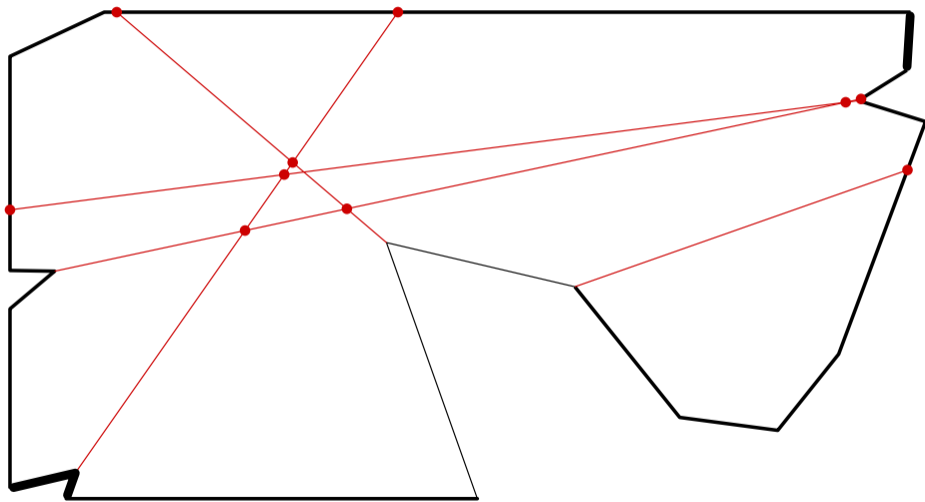
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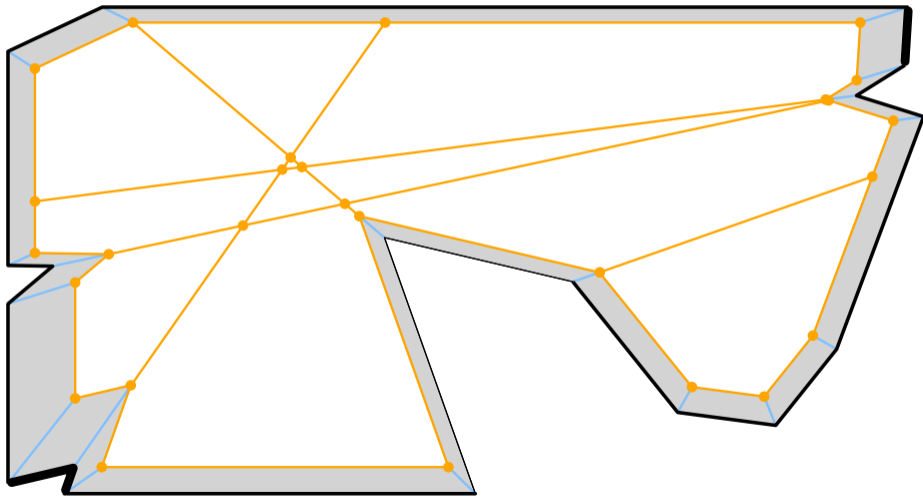
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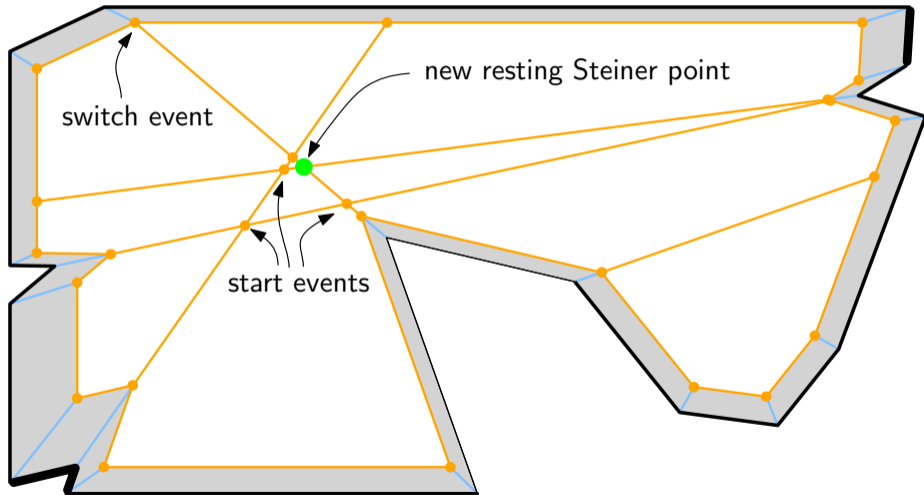
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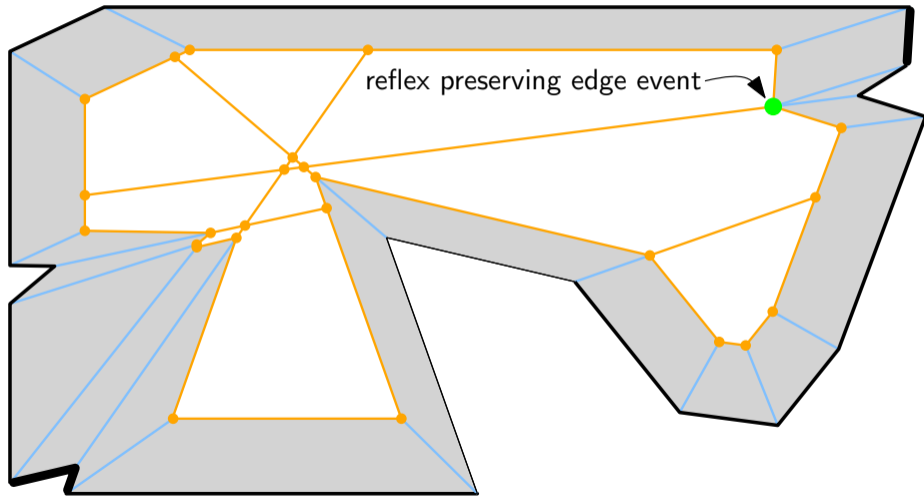
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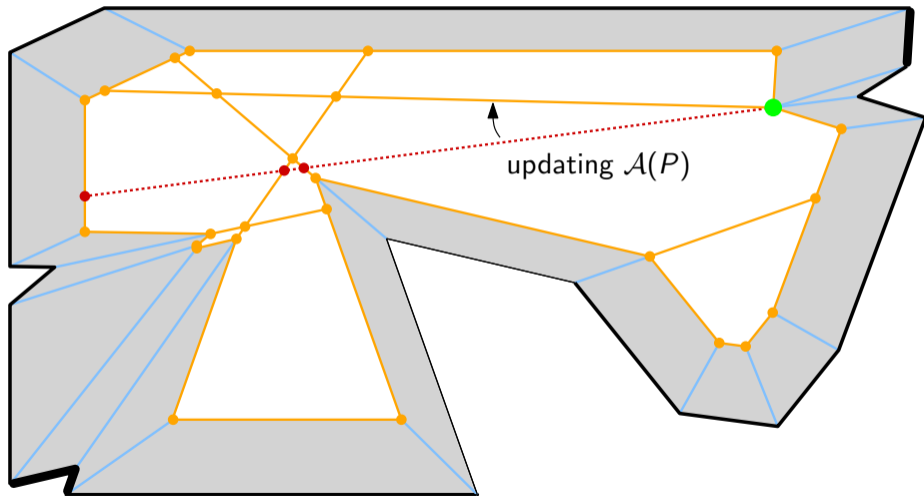
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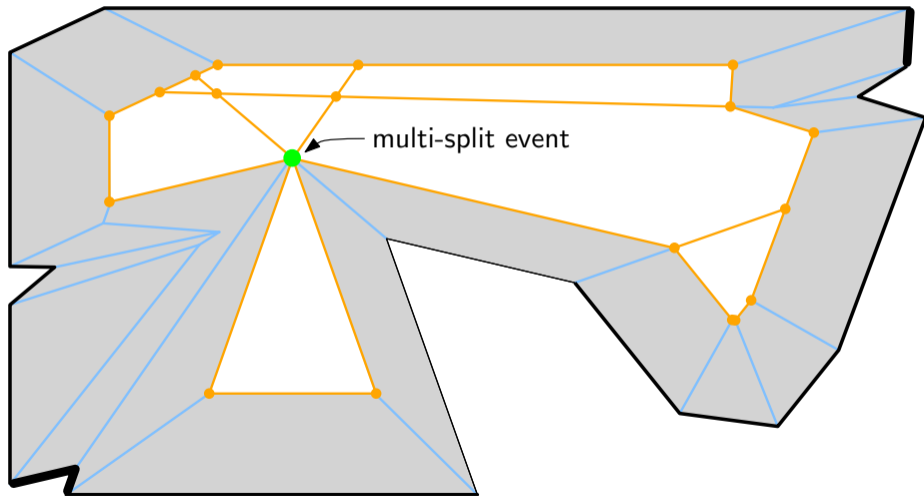
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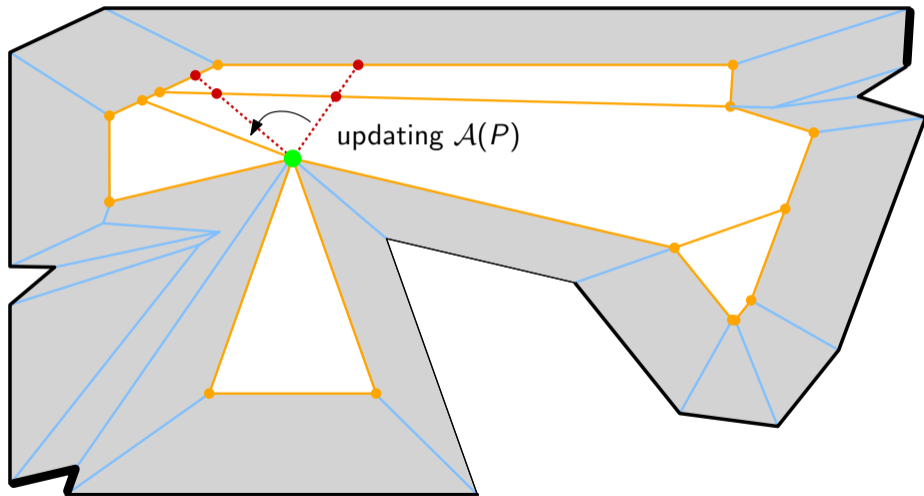
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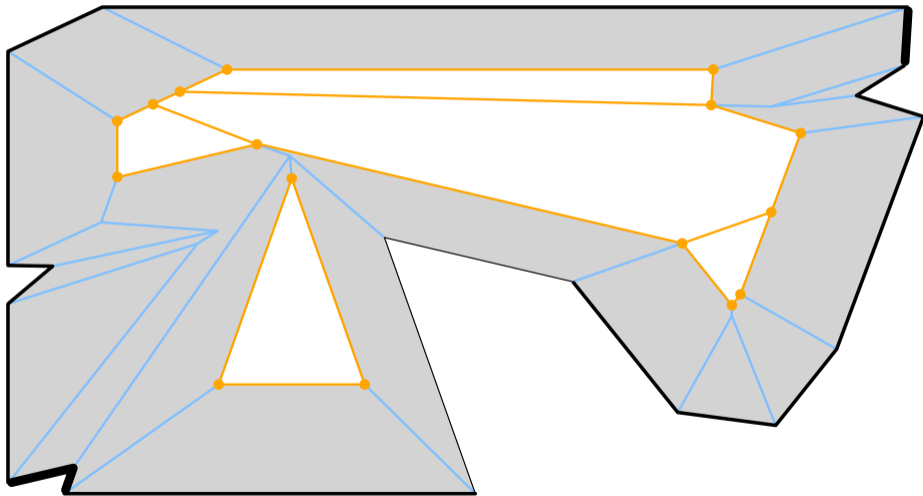
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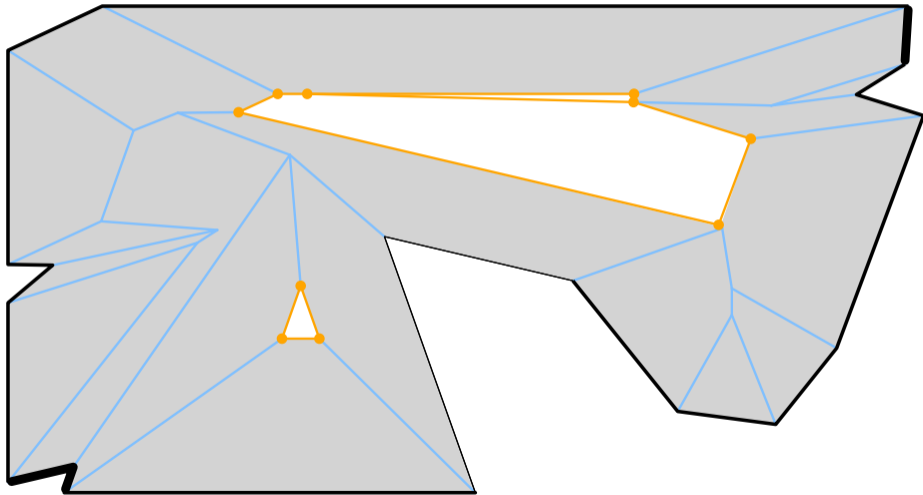
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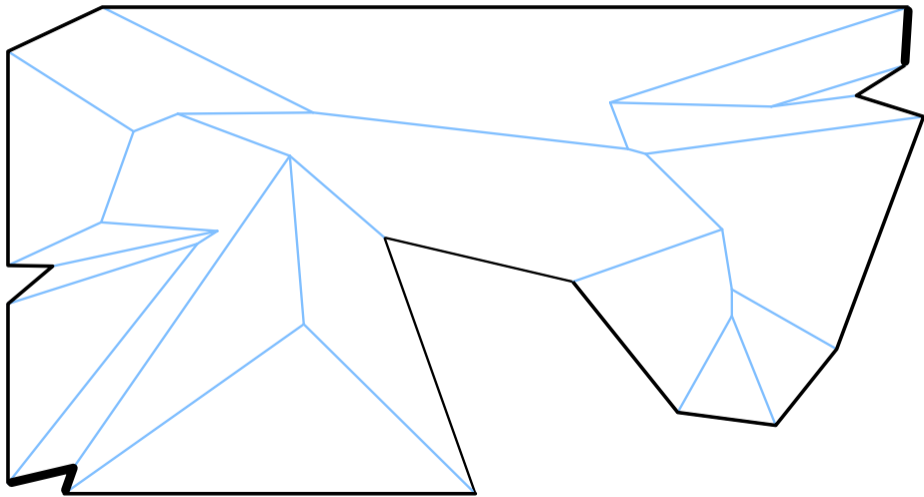
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Complexity Analysis

P consists of n vertices, r of which are reflex. The propagation of $\mathcal{W}_p^*(t, \sigma)$ results in:

- $\mathcal{O}(n)$ split/edge events,
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<i>classical</i>	<i>time</i>	<i>space</i>
	$\mathcal{O}(n^2 + r^3 + nr \log n)$	$\mathcal{O}(n)$

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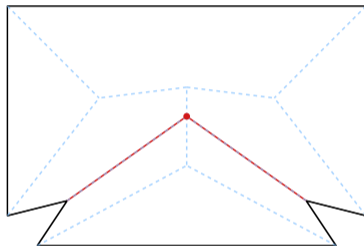
	<i>time</i>	<i>space</i>
<i>classical</i>	$\mathcal{O}(n^2 + r^3 + nr \log n)$	$\mathcal{O}(n)$
<i>trade-off</i> ⁵	$\mathcal{O}(n^2 + r^3/k + nr \log n)$	$\mathcal{O}(n + kr)$

⁵with a fixed k s.t. $1 \leq k \leq r$.

Summary and Q & A

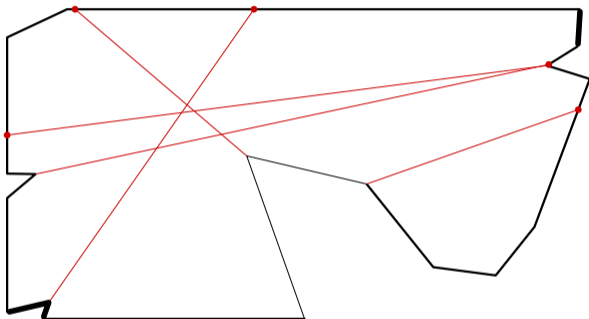
Summary

- Motorcycle Graph vs. $\mathcal{S}(P, \sigma)$.
- Induced Line Arrangement $\mathcal{A}(P)$.
- $\mathcal{W}_P^*(t, \sigma)$ as kinetic PSLG.
- Practical Candidate for an Implementation.
- Simple Space Time Trade-Off.
- $\mathcal{O}(n^2 + r^3/k + nr \log n)$ time and $\mathcal{O}(n + kr)$ space.



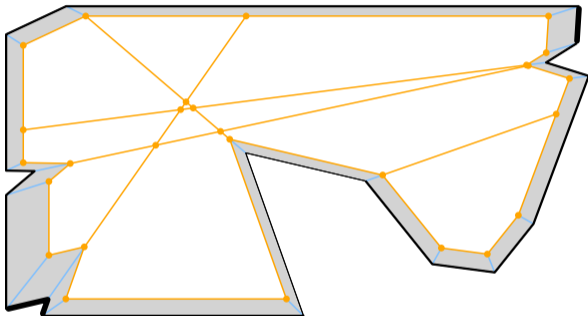
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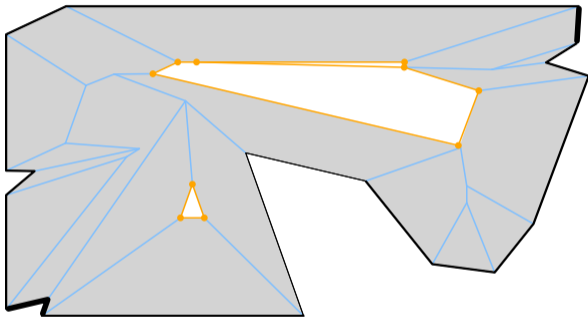
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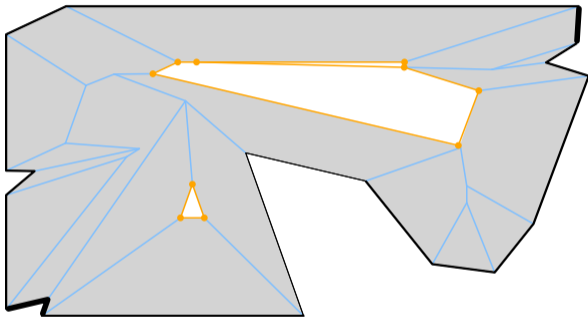
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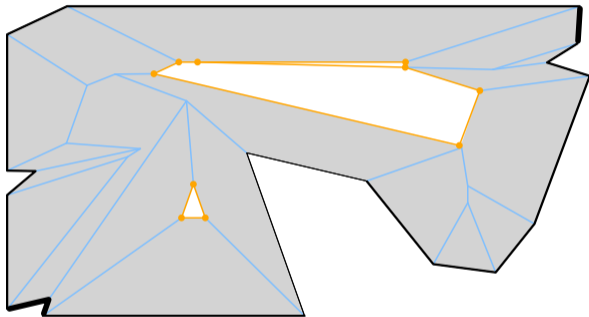
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Questions?

References I

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