

A Greedy Algorithm for Low-Crossing Partitions for General Set Systems

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Preliminaries

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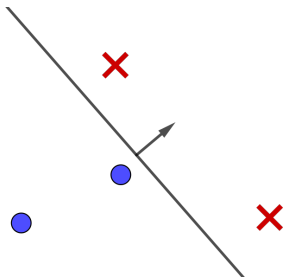
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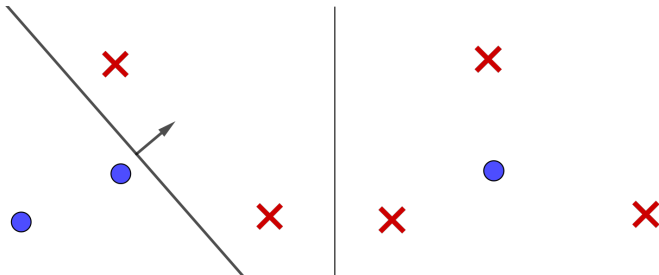


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Low-crossing Partitions

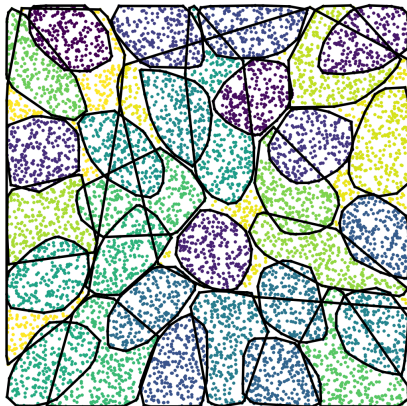
Given (X, \mathcal{F}) and a parameter $t \in [2.. \frac{n}{2}]$, find a partition of X in t parts P_1, \dots, P_t of size $O(\frac{n}{t})$ to minimize the *crossing number*:

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A partition of points in the unit cube and random halfspaces

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There are neither fast algorithms in practice in dimension > 3 to compute small crossing number partitions for geometric set systems nor for abstract set systems in any dimension.

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An algorithm that picks elements following this potential function builds partitions with crossing number $O(t^{1-1/d} \ln(t))$.

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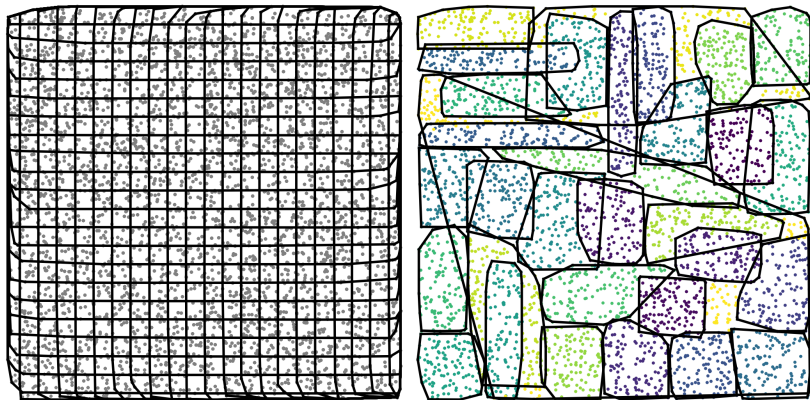
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 - We select an arbitrary element that keeps the cost of part (sum of costs of elements of the part) below the potential function.
- We double the weight of ranges crossing the part built.

Experiments

Experimental Setup

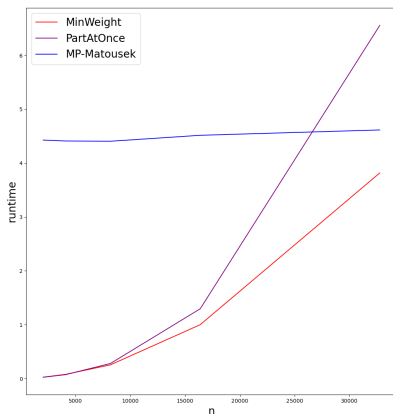
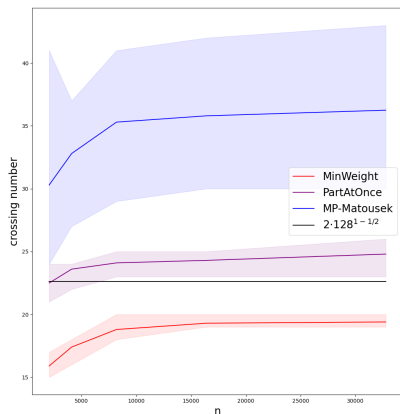
- We propose two variations of the greedy algorithms:
 - MINWEIGHT: We pick the element minimizing cost at each iteration.
 - PARTATONCE: We estimate weights with range sampling and pick the n/t elements with lowest estimated weight as a part.
- Experiments results are obtained from average over 10 runs.
- Experiments have been performed on AMD Ryzen 7 5800X (16 cores) @ 4.85 GHz (Home computer).

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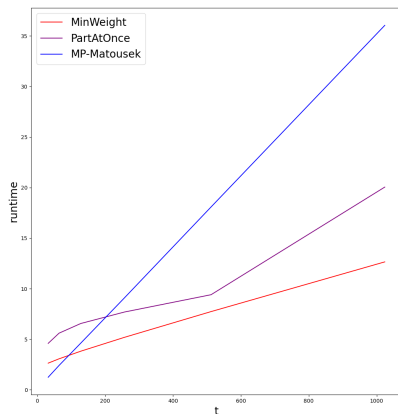
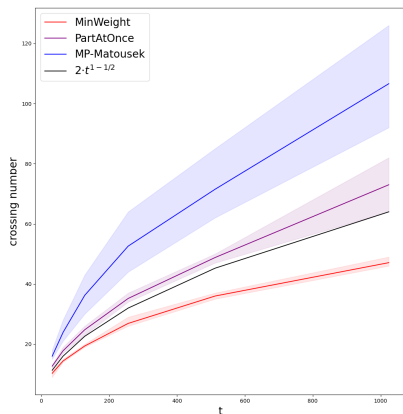
$n = 8192, t = 32$, crossing number : 10

Comparison with previous methods (n)



Comparison of crossing number and runtime of our algorithms for varying n and $t = 128, d = 2$.

Comparison with previous methods (t)



Comparison of crossing number and runtime of our algorithms for varying t and $n = 32768, d = 2$.

Real world datasets: ArXiv co-authorship graph

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Input t	$t^{4/5}$	MINWEIGHT		PARTATONCE	
		$\kappa_{\mathcal{F}}$	runtime (s)	$\kappa_{\mathcal{F}}$	runtime (s)
50	23	12	0.973	21	0.182
100	40	16	1.08	28	0.256
200	69	19	1.17	34	0.373
500	144	26	1.46	41	0.74

VC-dimension: 5 [CCDV24]

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Thank You!

More results: ε -approximations

ε – approximation

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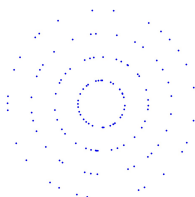
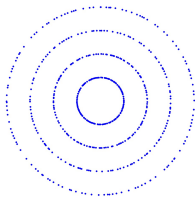
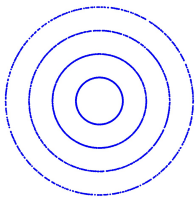
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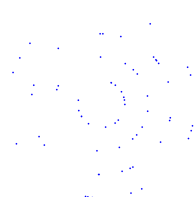
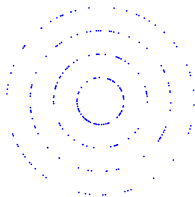
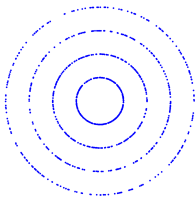
[STZ06] proved that it is possible to construct an ε -approximation of size $O\left(\frac{d}{\varepsilon^{\frac{2d}{d+1}}}\right)$ using low-crossing partitions for set systems with VC-dim $\leq d$.

ε -approximation experiments

our algorithm



random sample
[Tal94]



$$|A| = \frac{n}{4}$$

$$|A| = \frac{n}{16}$$

$$|A| = \frac{n}{64}$$

References

References I



David Coudert, Mónika Csikós, Guillaume Ducoffe, and Laurent Viennot.

Practical Computation of Graph VC-Dimension.

In *Symposium on Experimental Algorithms (SEA)*, July 2024.



B. Chazelle.

Cutting hyperplanes for divide-and-conquer.

Discrete & Computational Geometry, 9, February 1993.






S. Har-Peled.



Constructing planar cuttings in theory and practice.

SIAM J. Comput., 29, 2000.

References II

-  Yi Li, Philip M. Long, and Aravind Srinivasan.
Improved Bounds on the Sample Complexity of Learning.
Journal of Computer and System Sciences, 62(3):516–527,
May 2001.
-  Jiří Matoušek.
Efficient partition trees.
Discrete & Computational Geometry, 8:315–334, 1992.
-  Michael Matheny and Jeff M Phillips.
Practical low-dimensional halfspace range space sampling.
In *26th Annual European Symposium on Algorithms (ESA)*.
Schloss Dagstuhl-Leibniz-Zentrum fuer Informatik, 2018.

References III

-  Subhash Suri, Csaba D Toth, and Yunhong Zhou.
Range counting over multidimensional data streams.
Discrete & Computational Geometry, 36:633–655, 2006.
-  Michel Talagrand.
Sharper bounds for gaussian and empirical processes.
The Annals of Probability, pages 28–76, 1994.