

On Approximate Halfspace Range Counting and Relative ϵ -Approximations

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2: Counting

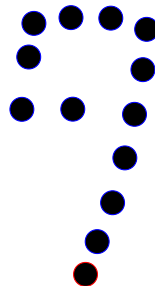
Counting

P: Set of n points in \mathbb{R}^d

Preprocess **P** for **Counting**

h: query halfspace

Output: $|\mathbf{P} \cap \mathbf{h}|$



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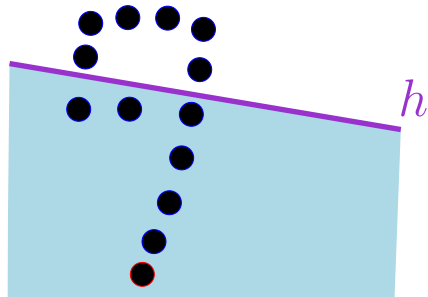
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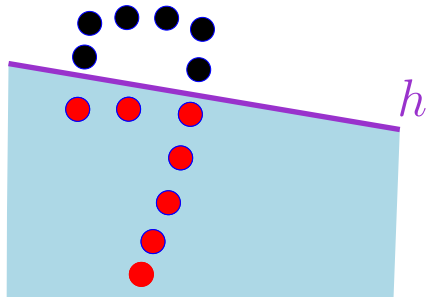
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3: Everybody knows that the dice are loaded

Already known to the Czech

(and many other nationalities):

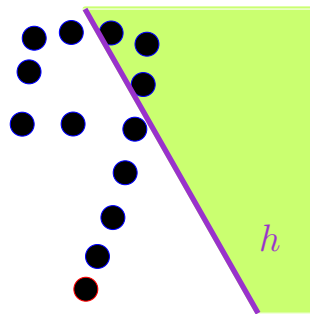
- Storage/preprocessing: $\approx O(n)$,
- Query time: $O(n^{1-1/d})$
via **Partition Trees** [Matoušek, 1992a]

Let's call this **"inefficient"**.

4: What's "Efficient"?

Halfspace Range Reporting / Emptiness

Report $\mathbf{P} \cap \mathbf{h}$ / Is $\mathbf{P} \cap \mathbf{h} = \emptyset$?



4: What's "Efficient"?

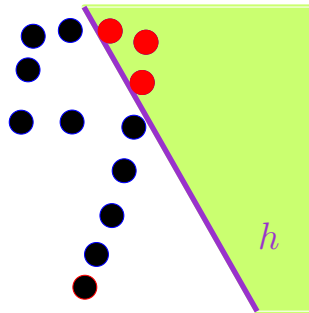
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[Matoušek, 1992b]:

Shallow Partition Trees



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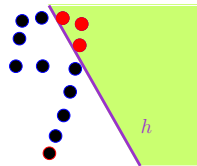
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Performance

storage/preprocessing: $\approx O(n)$,

Query time: $O(n^{1-1/\lfloor d/2 \rfloor} (+k))$

Now this is **efficient**!

4: What's "Efficient"?

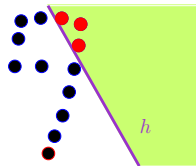
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Range Searching Performance

Storage/preprocessing: $\approx O(n)$,

	efficient Emptiness	inefficient Counting
Query time	$O(n^{1-1/\lfloor d/2 \rfloor} (+k))$	$O(n^{1-1/d})$

Interesting for $d = 3...$

5: But we want COUNTING

...and we want it now, and we don't care how!

Impossible(?)

Naa... use **Approximate Counting**:

Approx counting

Given $\epsilon > 0$

h: query halfspace

compute **C** s.t.

$$(1 - \epsilon)|P \cap h| \leq C \leq (1 + \epsilon)|P \cap h|$$

Relative error = error proportional to **weight(h)**.

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6: Approximate Counting

- **Tricky:** If $|\mathbf{P} \cap \mathbf{h}| < 1/\epsilon \Rightarrow$ **exact answer!**
- No better than **Emptiness queries**

Goal: Approx Counting

Achieve performance same as emptiness:

Storage/preprocessing $\approx \mathbf{O}(n)$,

Query time: $\approx \mathbf{O}_\epsilon(n^{1-1/\lfloor d/2 \rfloor})$

7: Self-competing alternative approaches

[Aronov and Har-Peled, 2005]

Reduction from emptiness to approximate counting.
Estimating $|\mathbf{P} \cap \mathbf{h}|$ using random samples.

[Kaplan and Sharir, 2006]

Estimating $|\mathbf{P} \cap \mathbf{h}|$ using random permutation:
Minimum-rank element in $\mathbf{P} \cap \mathbf{h}$ is a good estimator
(General technique of [Cohen, 1997])

In both cases, repeat $O\left(\frac{1}{\epsilon^2} \log n\right)$ times
To ensure high probability (**Chernoff**)

New result.

8: Our result:

- Achieve desired performance by **combining** the reporting / emptiness data structure (**Shallow partition tree**) with **relative-error approximations** at each node

- More versatile and modular

- Avoids the “Chernoff factor” $O\left(\frac{1}{\epsilon^2} \log n\right)$

(Sharper probabilistic analysis)

- Better dependence on ϵ ; Storage independent of ϵ
- Better control: Lots of opportunities for fine-tuning
(Also a con: Messy recurrences aplenty...)

New result.

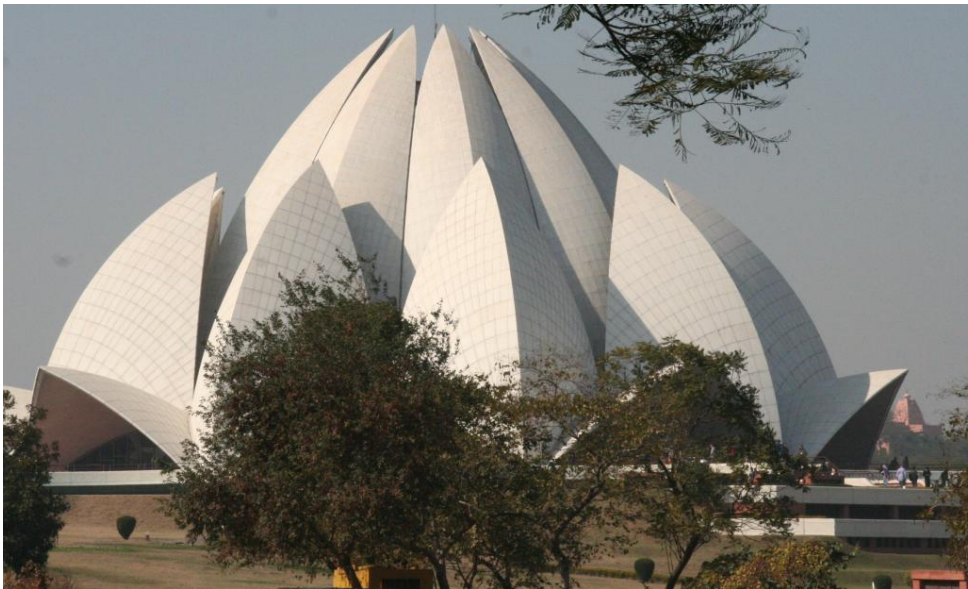
9: Approach (sketch)

- Use the shallow partition tree for emptiness / reporting
- At each node v , store a relative-error approximation A_v of P_v
- **Query with h** : At a node v of the tree:
Recurse if h crosses only a few children

$$O\left(r_v^{1-1/\lfloor d/2 \rfloor}\right)$$

- Otherwise h is not shallow: $|h \cap P_v| > |P_v|/r_v$
 \Rightarrow Can use $|A_v \cap h|$ to estimate $|P_v \cap h|$

And now for something completely different - the middle of the talk



10: Points and Spanning Trees

Good spanning tree

P: n points

Find good spanning tree for P?

What is good?

A: Light, small diameter, small crossing number, and carrot.

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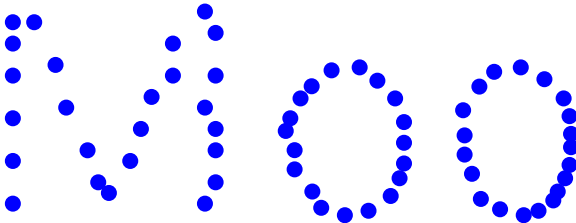
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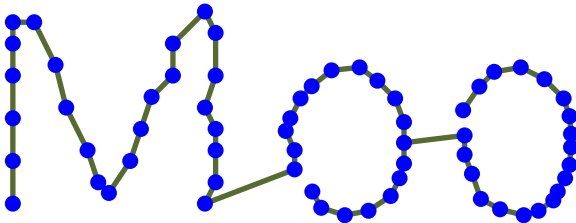
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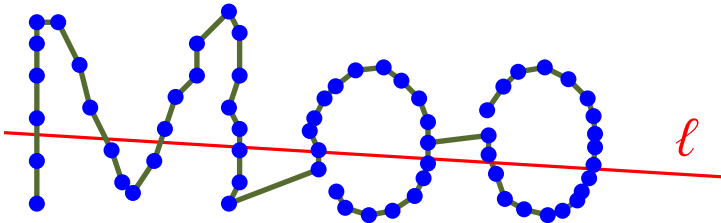
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11: Spanning tree with low crossing number

[Welzl, 1992]

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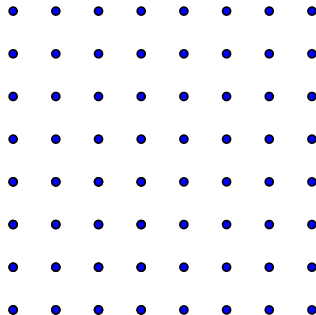
T: spanning tree for **P**

Such that any line ℓ crosses

$O(\sqrt{n})$ edges of **T**

Proof: Uses reweighting.

Mysterious.



Intuitively...

All point sets behave like they are on a grid...

[Matoušek, 1992a]: Partition trees.

11: Spanning tree with low crossing number

[Welzl, 1992]

P: n points

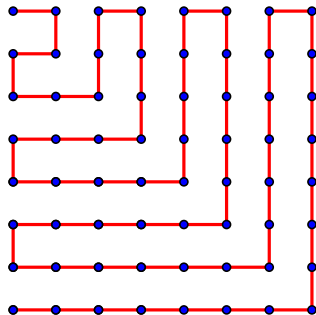
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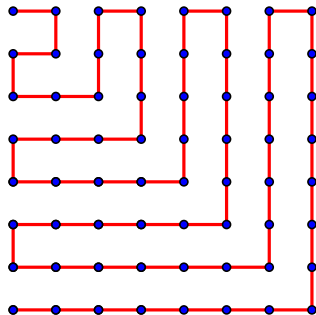
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12: Spanning tree with low crossing number

Question

Can Welzl's result be improved?

Spanning tree with relative crossing number

13: Can Welzl's spanning trees be improved?

Spanning tree with low crossing number - cont'd

Yes!

Spanning tree with relative crossing number

14: New Result

Spanning tree with relative crossing number

Theorem

T: relative spanning tree of **P** (set of **n** points).

For any line ℓ in the plane.

$k = \omega(\ell)$: # points on one side of ℓ .

$\Rightarrow \ell$ crosses $O\left(\sqrt{k} \log \frac{n}{k}\right)$

15: Half-Space Depth

Let us fix k ...

Depth Contour

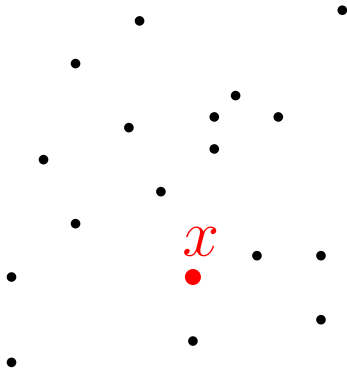
P : n points

x : a point.

Half-space depth of x :

Min number of points of P on a half plane of x .

\mathcal{D}_k : Contour of depth k .



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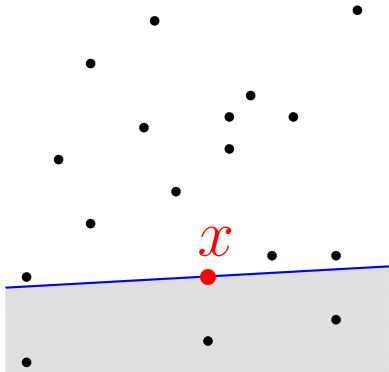
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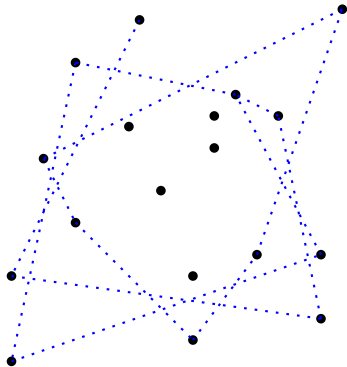
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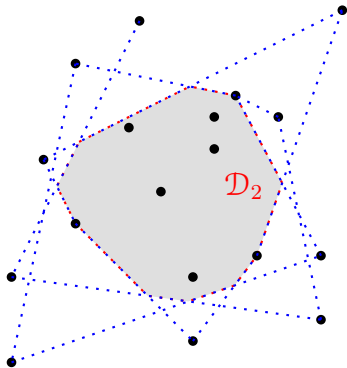
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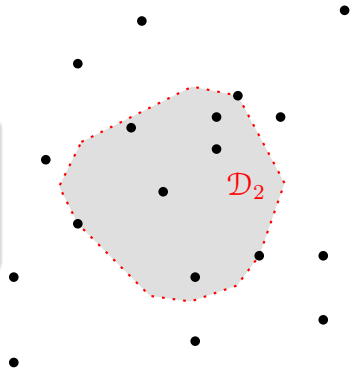
Fix k

Observations

ℓ a line

$$|\ell^+ \cap \mathbf{P}| > 2k \Rightarrow \ell \text{ intersects } \mathcal{D}_k$$

$$|\ell^+ \cap \mathbf{P}| \leq k \Rightarrow \ell \text{ avoids } \mathcal{D}_k$$



\mathcal{D}_k approximates points of depth $\leq k$.

Find spanning tree for $\mathbf{Q} = \mathbf{P} \setminus \mathcal{D}_k$

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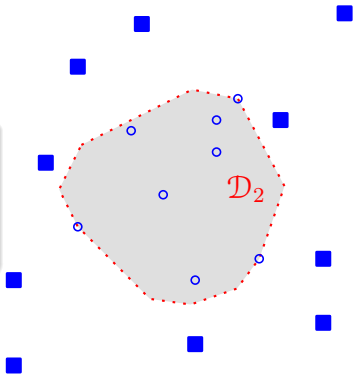
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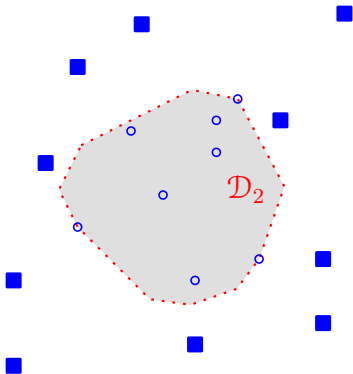
17: Half-Space Depth - cont'd

Idea:

$$Q = P \setminus \mathcal{D}_k$$

Partition Q into sets of size k :

- S_1, \dots, S_m .
- Any line crosses few sets.
- $T_i \leftarrow$ good spanning tree for S_i
[Welzl, 1992]
- Connect trees together into one big hairy tree.



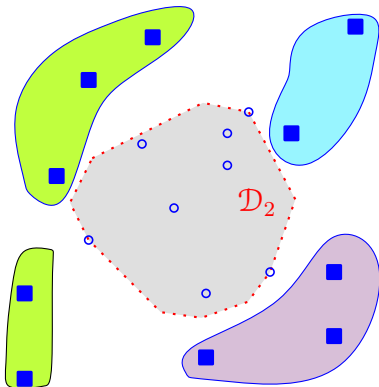
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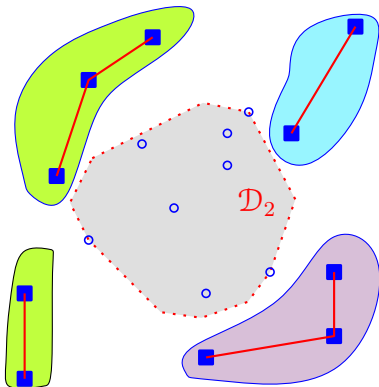
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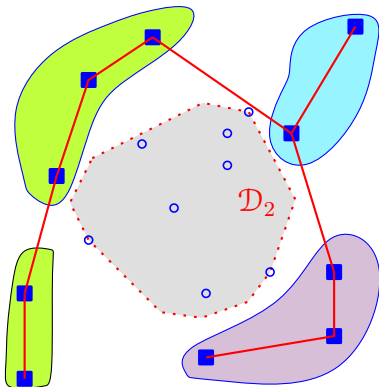
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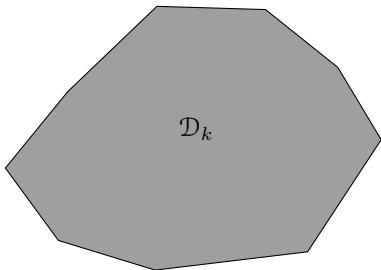
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18: Half-Space Depth - cont'd

Construction

- How to generate sets?
- Dobkin-Kirkpatrick....
- Any line crosses $O(\log \frac{n}{k})$ sets.
- Any line crosses $O(\sqrt{k} \log(n/k))$ edges spanning tree.

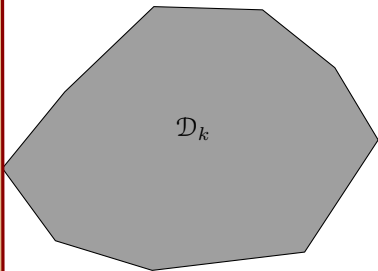


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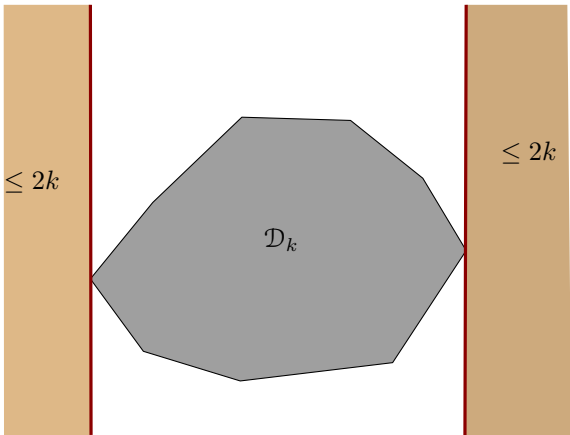
$\leq 2k$



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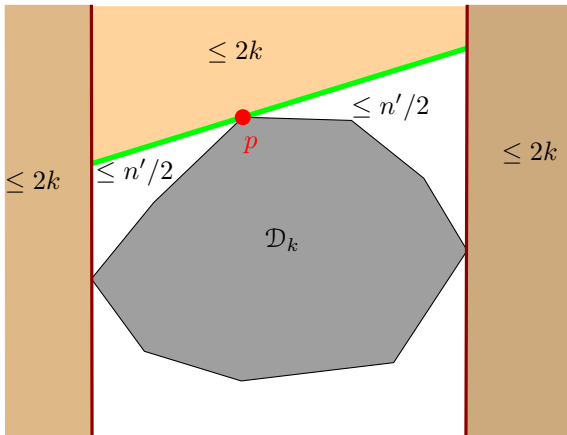
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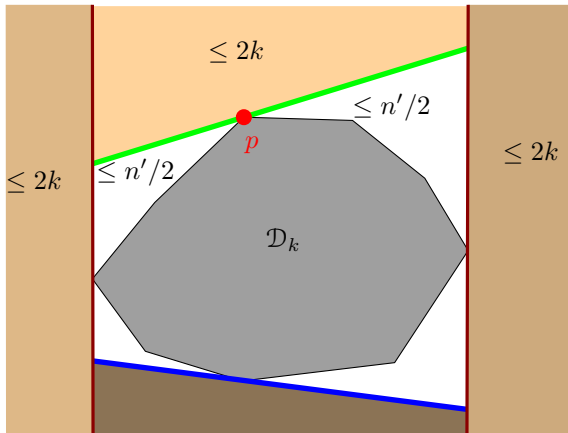
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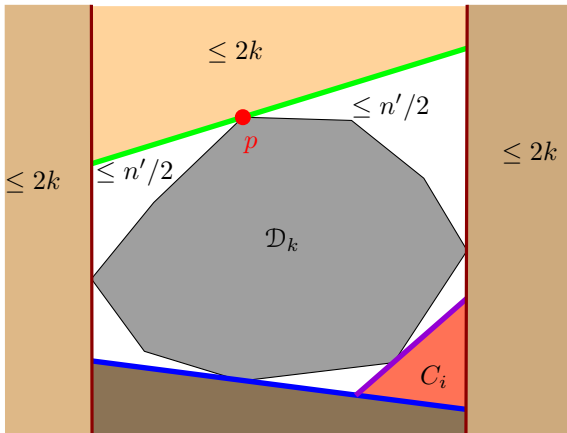
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19: If k is not known...

Onion peeling.

Algorithm

Set $k_i = 2^i$. $P_0 = P$, and $i = 1$.

$Q_i = P_{i-1} \setminus \mathcal{D}(P_{i-1}, k_i)$

$T_i \leftarrow$ spanning tree of Q_i .

$P_i = P_{i-1} \setminus Q_i$.

$T =$ connect $T_1, T_2 \dots$ together.

Crossing number of T , for a line of weight k :

$$\sum_i^{\lg k} O\left(\sqrt{2^i} \log(n/2^i)\right) = O(\sqrt{k} \log(n/k)),$$

20: Conclusions

- A lot of other results in the paper.
- **Open:** Extend sensitive spanning tree to higher dimensions.
- **Open:** Find an alt alg/proof of the Welzl [**Welzl, 1992**] result.
(Related to partitions.)
- **Open:** Better understanding of discrepancy.



Aronov, B. and Har-Peled, S. (2005).

On approximating the depth and related problems.

In *Proc. 16th ACM-SIAM Sympos. Discrete Algorithms*, pages 886–894.



Cohen, E. (1997).

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Randomized incremental constructions of three-dimensional convex hulls and planar voronoi diagrams, and approximate range counting.

In *Proc. 17th ACM-SIAM Sympos. Discrete Algorithms*, pages 484–493.



Matoušek, J. (1992a).

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Welzl, E. (1992).

On spanning trees with low crossing numbers.

In *Data Structures and Efficient Algorithms, Final Report on the DFG Special Joint Initiative*, volume 594 of *Lect. Notes in Comp. Sci.*, pages 233–249. Springer-Verlag.