

# Klee's measure problem

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# Overview

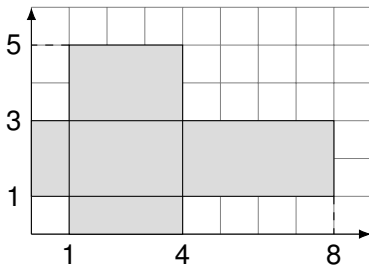
Introduction

Space partitioning

Summary

# Back to the basics

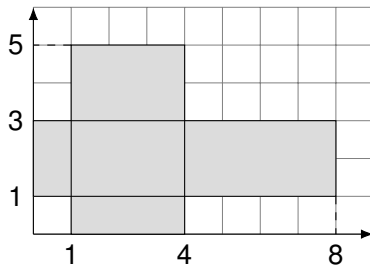
Find the colored area



$A = ?$

# Back to the basics

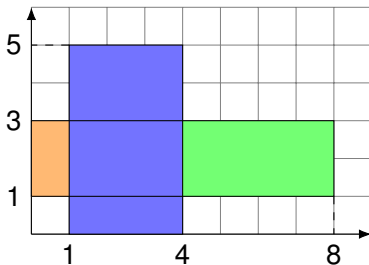
Find the colored area



$$A = 25$$

# Back to the basics

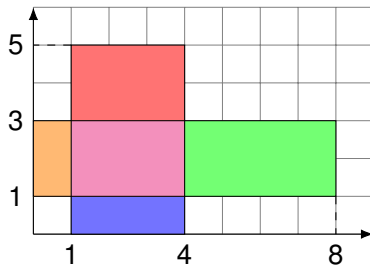
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$$A = 25 = 2 + 15 + 8$$

# Back to the basics

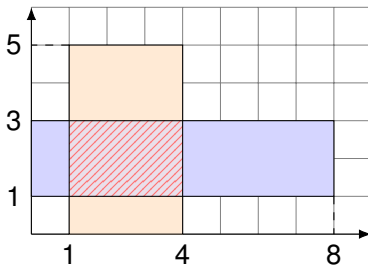
Find the colored area



$$A = 25 = 2 + 6 + 6 + 3 + 8$$

# Back to the basics

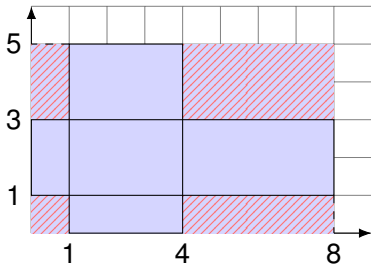
Find the colored area



$$A = 25 = 16 + 15 - 6$$

# Back to the basics

Find the colored area



$$A = 25 = 40 - 15$$



# Today's Topic

## Problem (Klee's measure problem)

*Given a set  $B = \{b_1, b_2, \dots, b_n\}$  of  $n$   $d$ -dimensional boxes, find the volume of the union of all boxes in  $B$ .*

## Definition (Hyperrectangle a.k.a. *Box*)

A hyperrectangle is a Cartesian product of finite intervals.

# What we will discuss

- ▶ Two space partitioning approaches toward Klee's measure problem.

# The first steps: 1-dim. case

Klee, 1977

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Klee, 1977

This trivial algorithm

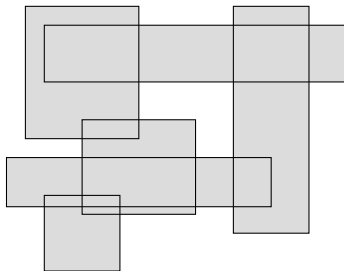
1. utilizes sweeping
2. uses no special data structures
3. runs in
  - ▶  $O(n \log n)$ -time<sup>1</sup> ( $O(n)$  w/ sorted input)
  - ▶  $\Theta(n)$ -space ( $O(1)$  w/ sorted input)

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<sup>1</sup>Actually  $O(n \log p)$ , where  $p$  is the minimum number of lines stabbing all intervals

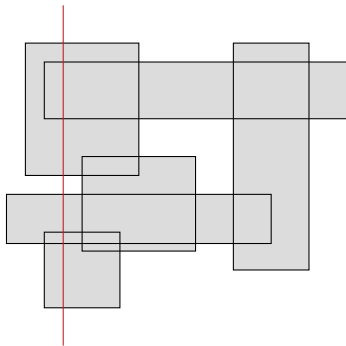
# The first steps: 2-dim. case

Bentley, 1977



# The first steps: 2-dim. case

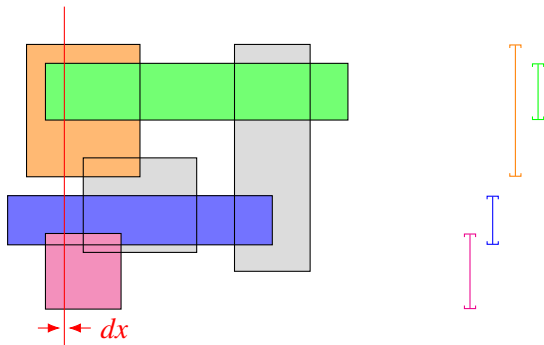
Bentley, 1977



Try sweeping!

# The first steps: 2-dim. case

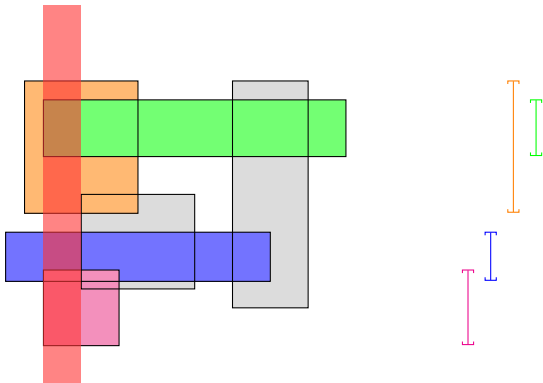
Bentley, 1977



The volume differential is  $dx \cdot (\text{measure of the cross section})$

# The first steps: 2-dim. case

Bentley, 1977



Measure intervals and compute the volume of the slab



# The first steps: 2-dim. case

Bentley, 1977

This algorithm<sup>2</sup>

1. uses sweeping
2. uses segment tree maintaining partial measure of corresponding intervals
3. runs in
  - ▶  $O(n \log n)$ -time (at most  $2n$  updates on segment tree)
  - ▶  $\Theta(n)$ -space

---

<sup>2</sup>The article is unpublished. Referred to Leeuwen and Wood, 1981 instead.

## Beyond 3-dimension

- ▶ Bentley's algorithm easily extends to  $d$ -dimensional boxes, where  $d \geq 2$  is arbitrary integer
- ▶ Shows  $O(n^{d-1} \log n)$  running time.
- ▶ Unfortunately, the known (non-tight) lower bound is  $\Omega(n \log n)$  regardless of the dimensionality.
- ▶ How can we design improved algorithms?

# Space partitioning

- ▶ *Partitioning* a given (euclidean) space.

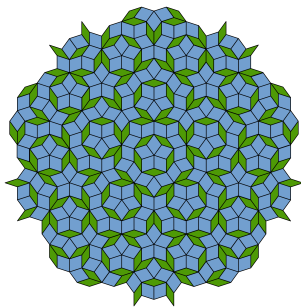


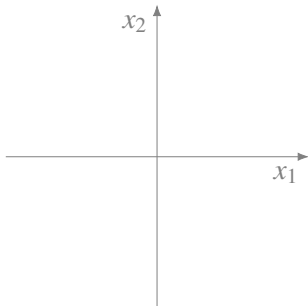
Figure: Penrose tiling

# Data structures

- ▶  $k$ -d tree
- ▶ Quadtree/Octree
- ▶ Binary Space Partitioning tree
- ▶ and many others...

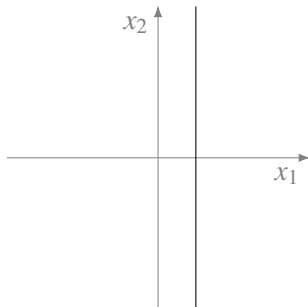
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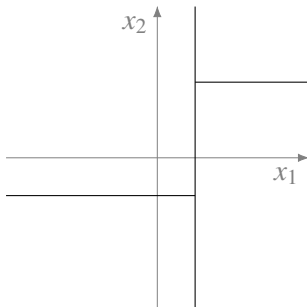
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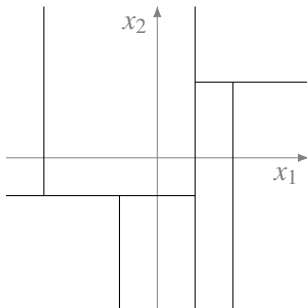
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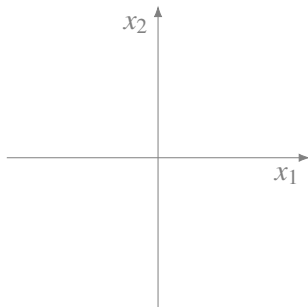
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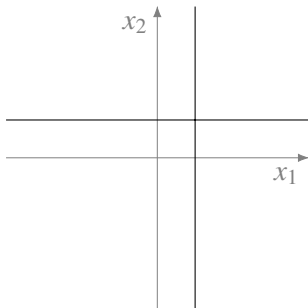
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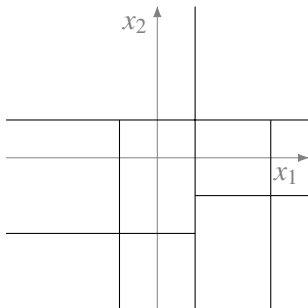
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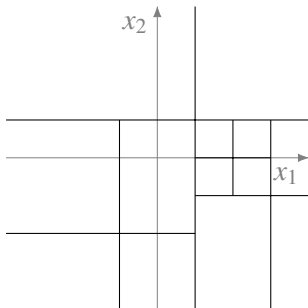
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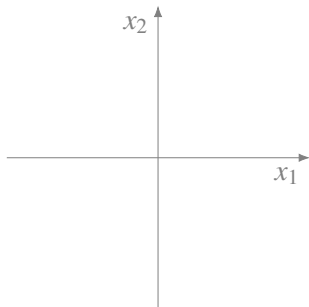
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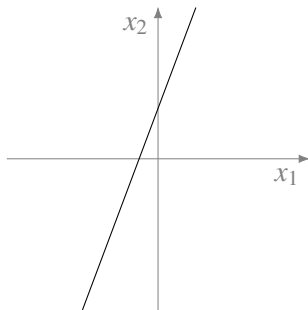
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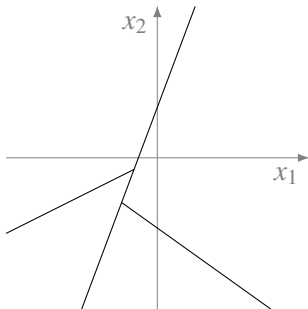
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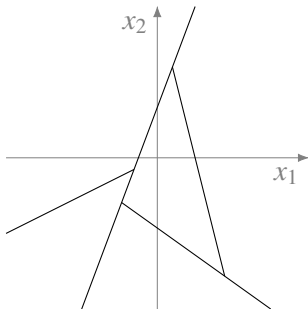
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# Data structures

- ▶  $k$ -d tree
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- ▶ **BSP tree**
- ▶ and many others...





# Which one is effective?

- ▶ Problem-by-problem.
- ▶ Key is *how* to partition the space.  
Note: the sweeping algorithms effectively partition the space into slabs in which some “characteristics” are preserved.
- ▶ We will see two approaches for the Klee’s measure problem.

# Approach 1

Overmars and Yap, 1991

## Key ideas

The linear increase ( $O(n^{d-1} \log n)$ ) w.r.t. dimensionality is due to the recursive structure of the computation.

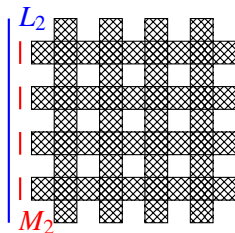
Suppose we sweep the whole space along the last ( $x_d$ ) axis. Can we maintain the  $(d - 1)$ -dim. cross section in a single data structure?

# Idea



Figure: Trellis (취빙)

## Idea



The volume is  $\prod L_i - \prod (L_i - M_i)$  and computing  $L_i$ 's and  $M_i$ 's are simple problems.

Divide the space into this trellis pattern!

# Data structure

## Orthogonal partition tree

- ▶ A balanced binary tree.
- ▶ A node  $\alpha$  has an associated region  $C_\alpha$ ;  $C_{\text{root}}$  is the whole space.
- ▶ For any two children  $\alpha_1$  and  $\alpha_2$  of a node  $\alpha$ ,  $\{C_{\alpha_1}, C_{\alpha_2}\}$  is a partition of  $C_\alpha$ .

# Data structure

## Orthogonal partition tree for Klee's measure problem

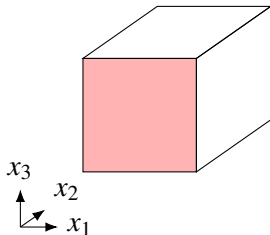
- ▶  $M_\alpha$  is the total measure under the subtree rooted at  $\alpha$ .
- ▶  $T_\alpha$  is the number of boxes covering whole  $C_\alpha$  but not  $C_{parent(\alpha)}$ .
  - ▶ If  $T_\alpha > 0$  ( $C_\alpha$  is fully covered),  $M_\alpha = V(C_\alpha)$
  - ▶ Else  $M_\alpha = M_{left(\alpha)} + M_{right(\alpha)}$
- ▶ A leaf  $\lambda$  maintains a set  $B_\lambda$  of boxes that intersect with the interior of  $C_\lambda$  but do not cover  $C_{parent(\lambda)}$ .

# Terms

## Definition ( $i$ -boundary)

For a  $d$ -box, its  $i$ -boundaries are its  $(d - 1)$ -dim. faces perpendicular to  $x_i$ -axis. Note that a  $d$ -box has two  $i$ -boundaries for all  $1 \leq i \leq d$ .

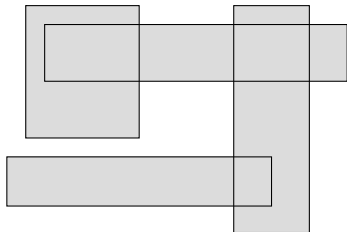
## Example



The red face  $[0, 1] \times \{0\} \times [0, 1]$  is a 2-boundary of the unit cube.

# Partition strategy

3-dim. case; 2-dim. projection

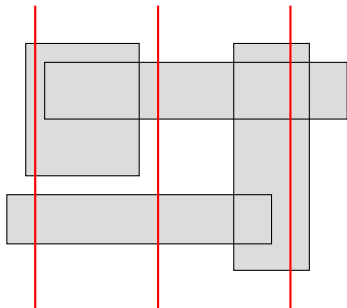


1. Split  $x_1$ -axis into  $2\sqrt{n}$  intervals such that each contains at most  $\sqrt{n}$  1-boundaries.
2. For each 1-boundaries contained in a slab, split the slab along its 2-boundaries
3. For all other 2-boundaries, split along the  $\sqrt{n}$ -th 2-boundaries.



# Partition strategy

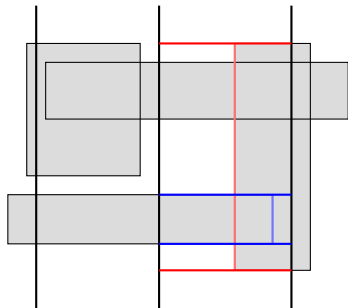
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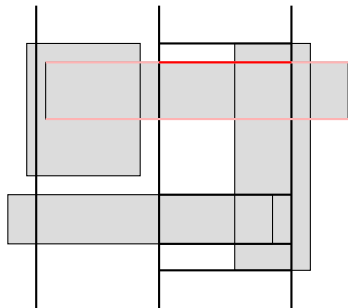
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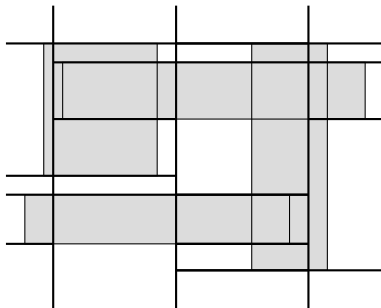
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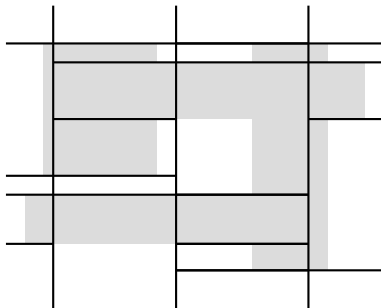
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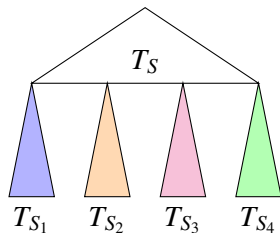
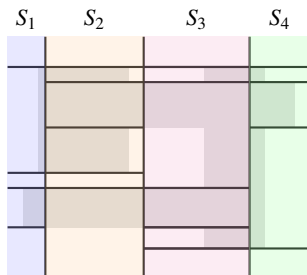
# Characteristics

For 2-dimension case,

1. The space is divided into  $O(n)$  cells.  
:  $2\sqrt{n}$  slabs split into  $4\sqrt{n}$  cells.
2. Each box of  $B$  partially covers at most  $O(\sqrt{n})$  cells.  
: Each vertical line(1-boundary) cut through  $4\sqrt{n}$  cells and horizontal line cut through  $2\sqrt{n}$  slabs.
3. No cell contains vertices in its interior.  
: The vertices are in some horizontal boundaries.  
: The boxes are in trellis pattern
4. Each cell has at most  $O(\sqrt{n})$  boxes partially covering it.  
: A cell contains at most  $\sqrt{n}$  1-boundaries and a slab contains at most  $\sqrt{n}$  2-boundaries.

# Data structure

## Orthogonal partition tree



# Characteristics

## Orthogonal partition tree

For 2-dimension case,

1. The tree has  $O(n)$  leaves.
2. Each box stored in at most  $O(\sqrt{n})$  leaves.
3. No  $C_\lambda$ 's contain vertices in its interior.
4. Each leaf stores at most  $O(\sqrt{n})$  boxes.
5. Each box influences at most  $O(\sqrt{n} \log n)$   $T_\alpha$ 's.  
: from 1 and 2.



# Analysis

## 3-dim. case

- Box insertion**
- ▶ A box is stored in at most  $O(\sqrt{n})$  leaves ( $O(\sqrt{n} \log n)$ )
  - ▶  $M_\lambda$  is computed from two segment trees for each axis. ( $O(\sqrt{n} \log n)$ )
  - ▶  $M_\alpha$  and  $T_\alpha$  is updated for all nodes  $\alpha$  between the leaves  $\lambda$  and the root ( $O(\sqrt{n} \log n)$  updates)

**Box deletion** Similar to inserting analysis

**Measure query**  $M_{\text{root}}$  is the answer.  $O(1)$ .

## Theorem

*The algorithm runs in  $O(n \sqrt{n} \log n)$ -time.*

# Extend to higher dimension

## Partition strategy

1. Split  $x_1$ -axis into  $2\sqrt{n}$  intervals such that each contains at most  $\sqrt{n}$  1-boundaries.
2. For each boxes whose 1-boundaries contained in a 1-slab, split the slab along its 2-boundaries
3. For all others, split the 1-slab along the  $\sqrt{n}$ -th 2-boundaries.
4. For each boxes whose 1- and 2- boundaries in a 2-slab, split the slab along the 3-boundaries.
5. For all others, split the 2-slab along the  $\sqrt{n}$ -th 3-boundaries.
6. (repeat)

# Extend to higher dimension

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# Summary

Overmars and Yap, 1991

For  $d$ -dim. Klee's measure problem, this algorithm

1. runs in  $O(n^{d/2} \log n)$ -time with the partition tree.
2. uses  $O(n^{d/2})$  space  
: This can be reduced to  $O(n)$  (only segment trees) by interleaving the measuring step with the partitioning.

# Approach 2

Chen, 2013

## Key ideas

The logarithmic factor ( $O(n^{d/2} \log n)$ ) comes from maintaining the tree while sweeping.

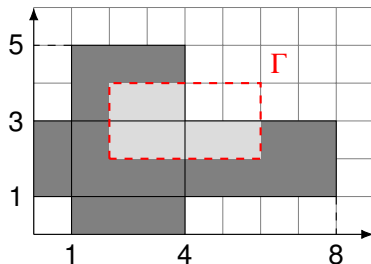
Can we design a different partitioning method that is free from maintaining a tree?

## Modified problem

### Problem (Modified version of the problem)

For a set of  $d$ -boxes  $B$  and an open box  $\Gamma$ , find the complement volume of union of  $B$  within the domain  $\Gamma$ .

### Example



$$A^C(\Gamma) = 2$$

# Algorithm

1: **function** MEASURE( $B, \Gamma$ )

**Given:**  $C$  is a fixed, small constant

2:     **if**  $|B| < C$  **then return** the answer directly.

3:     *Simplify*  $B$

4:     *Cut*  $\Gamma$  into two disjoint boxes  $\Gamma_L$  and  $\Gamma_R$

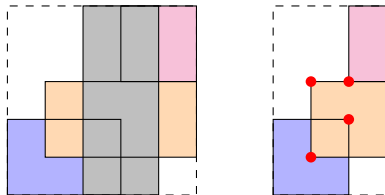
5:     **return** MEASURE( $\{b \cap \Gamma_L \mid b \in B\}, \Gamma_L$ )  
          + MEASURE( $\{b \cap \Gamma_R \mid b \in B\}, \Gamma_R$ )

6: **end function**



# Simplification

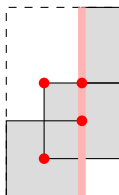
- ▶ Remove all slabs (a box of  $\{x \mid a \leq x_i \leq b\}$  form in  $\Gamma$ ) and adjust  $B$  and  $\Gamma$ .
- ▶ This costs linear time per axis.
- ▶ Note that the complement volume is preserved and all remaining boxes have a  $(d - 2)$ -face intersecting with  $\Gamma$ .



# Partitioning

## 2-dim. case

- ▶ Split  $\Gamma$  into two open boxes at the median of  $x_1$ -coord of all  $(d - 2)$ -faces
- ▶ Swap axis number



# Partitioning

## 3+-dim. case

- ▶ Assign a weight  $2^{(i+j)/d}$  on all  $(d - 2)$ -faces perpendicular to  $x_i$  and  $x_j$ -axes.
- ▶ The weight is bounded in  $[1, 4]$ .

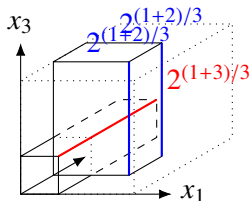


Figure: 1-faces of 3-d boxes intersecting the open domain

# Partitioning

## 3+-dim. case

- Find a weighted median  $m$  among the intersection of the  $(d - 2)$  faces and the  $x_1$ -axis, and cut  $\Gamma$  through the hyperplane  $x_1 = m$ .

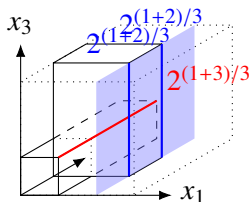


Figure: 1-faces of 3-d boxes intersecting the open domain

# Partitioning

## 3+-dim. case

- ▶ Shift axis indices ( $1 \rightarrow d \rightarrow (d - 1) \rightarrow \dots \rightarrow 3 \rightarrow 2 \rightarrow 1$ ).
- ▶ Effectively a  $k$ -d tree.

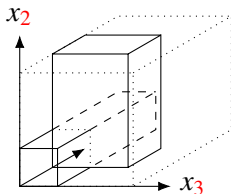


Figure: 1-faces of 3-d boxes intersecting the open domain

# Partitioning

## Weight decrease

- ▶ After cutting,  $(d - 2)$ -faces not perpendicular to  $x_1$ -axis ( $i, j \neq 1$ ) will have weight  $2^{(i-1+j-1)/d}$ , decreased by  $2^{2/d}$ .
- ▶  $(d - 2)$ -faces perpendicular to  $x_1$  axis ( $j \neq 1$ ) will have weight  $2^{(d+j-1)/d}$ , increased by  $2^{(d-2)/d}$ . Note that these faces are split into smaller domains by half, reducing the total weight by half.

# Analysis

- Simplifying
  - ▶  $O(n)$  for each axis to identify a slab
  - ▶  $O(n)$  for each axis to adjust box boundaries
- Cut
  - ▶ Finding (weighted) median is  $O(n)$  after sorting
  - ▶ The total weight is decreased by  $2^{2/d}$  for each cutting step.

## Theorem

*The algorithm runs in  $O(n^{d/2})$ -time.*

# Algorithm Summary

Dim.	Time complexity
1	$\Theta(n \log n)^3$
2	$\Theta(n \log n)$
3+	$\Omega(n \log n) - O(n^{d/2})$

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<sup>3</sup> $O(n \log p)$  where  $p$  is min. #lines stabbing all intervals.