

CSE520 Computational Geometry

Lecture 22

Geometric Approximation Algorithms II

Antoine Vigneron

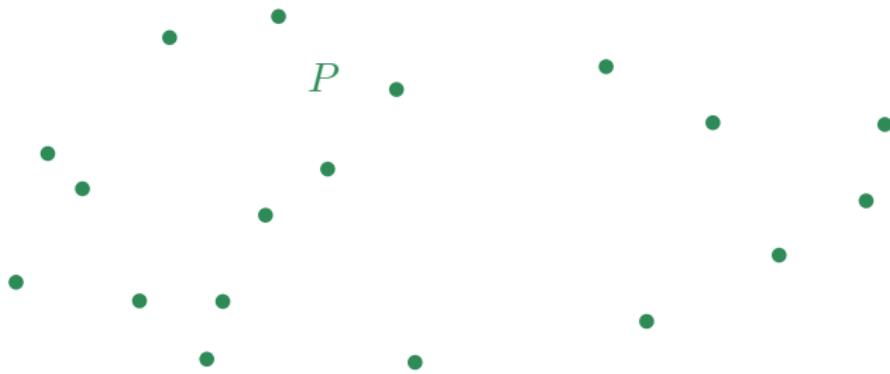
Ulsan National Institute of Science and Technology

June 15, 2020

Course Organization

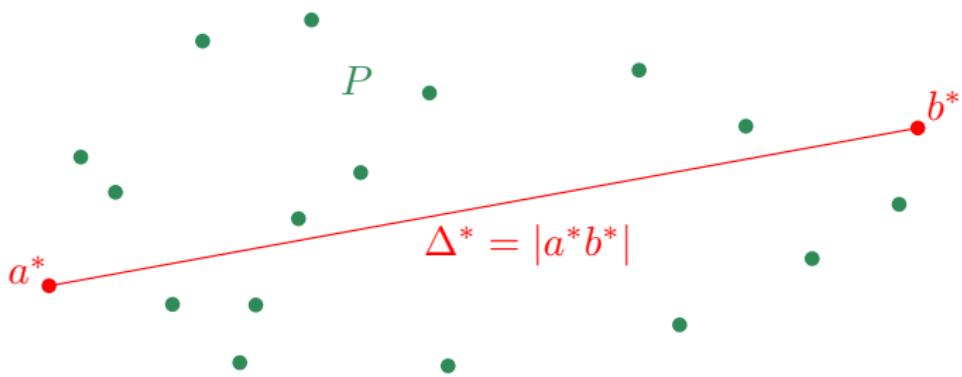
- Today, I will show how to improve the algorithm from Lecture 21 for approximating the diameter of a point set.
- References:
- Sariel Har Peled's [book](#).
- [Paper](#) by T. Chan, *Approximating the diameter, width, smallest enclosing cylinder, and minimum-width annulus*, Section 2.

The Diameter Problem



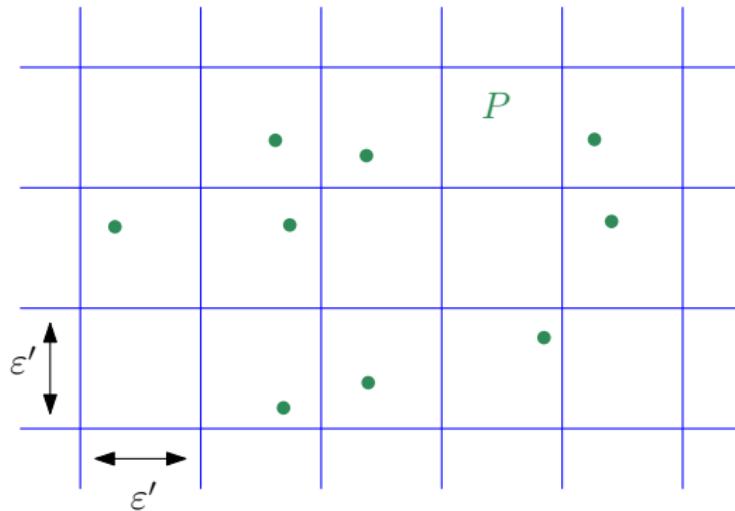
- Input: a set P of n points in \mathbb{R}^d .

The Diameter Problem



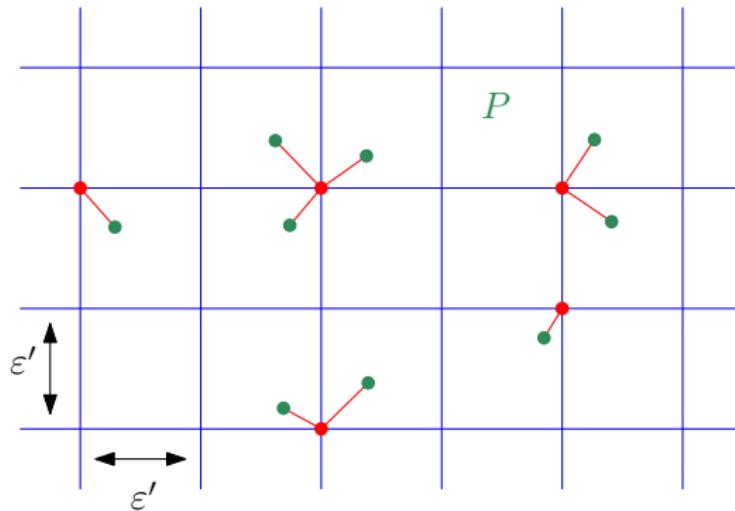
- Input: a set P of n points in \mathbb{R}^d .
- Output: the maximum distance Δ^* between any two points of P .
- $\Delta^* = \text{diam}(P)$ is the *diameter* of P .

Rounding to a Grid



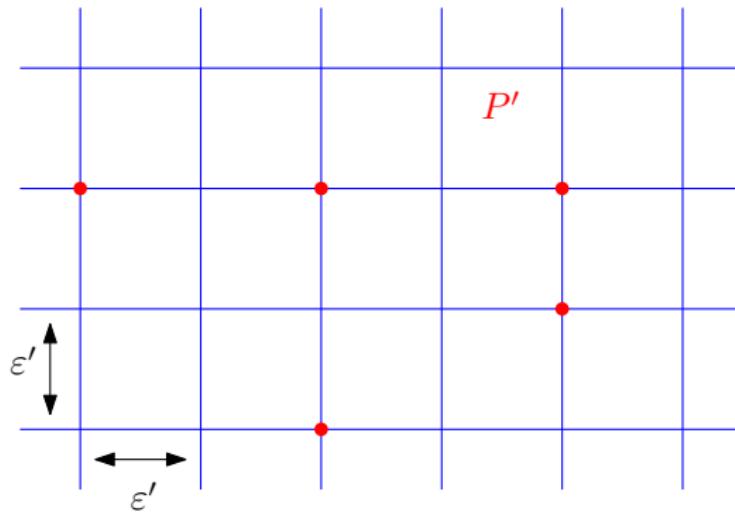
- Consider a regular grid over \mathbb{R}^d .
- The side length of the grid is ε' , to be specified later.
- Intuition: we will choose $\varepsilon' \approx \varepsilon\Delta^*$, which is the error we allow.

Rounding to a Grid



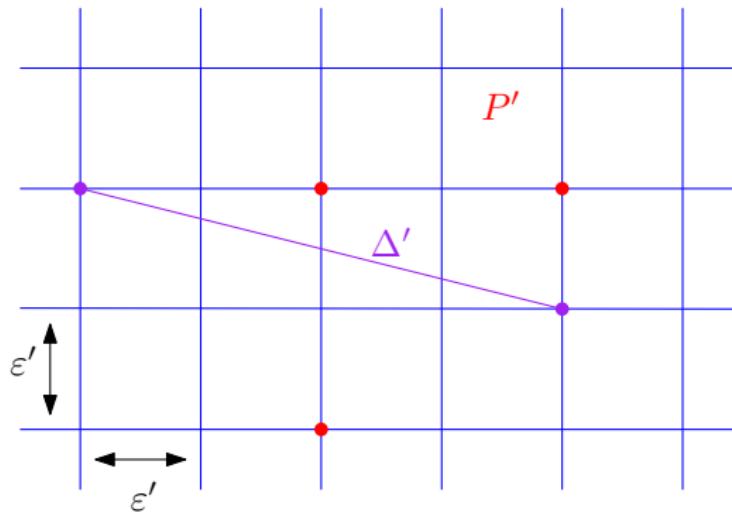
- Replace each point of P with the nearest grid point.
- This operation is called *rounding*.

Rounding to a Grid



- The grid points we obtain form the set P' .

Rounding to a Grid

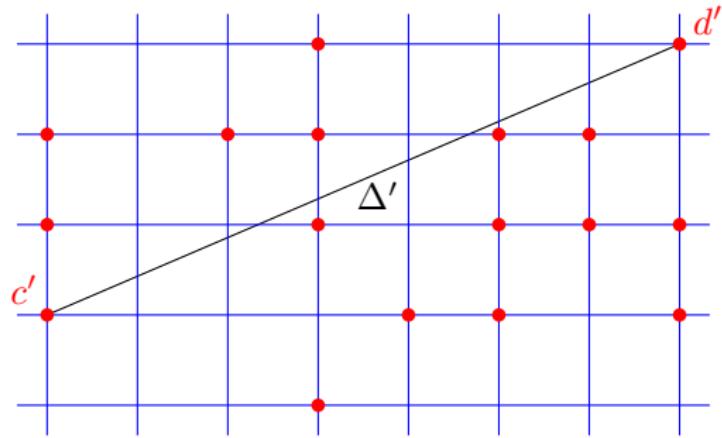


- Compute $\Delta' = \text{diam}(P')$ by brute force.

Rounding to a Grid

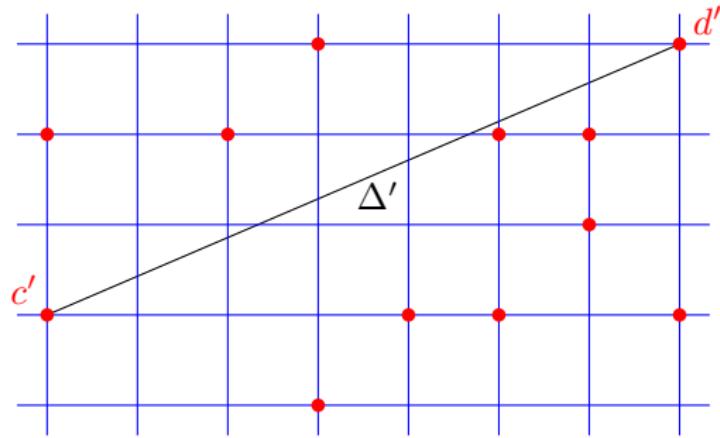
- Let Δ_0 be the 2-approximation of Δ^* that we computed in linear time.
- In the previous lecture, we saw that if we choose $\varepsilon' = \frac{\varepsilon\Delta_0}{4\sqrt{d}}$, then $\Delta = \Delta' - \varepsilon'\sqrt{d}$ satisfies $\Delta \leq \Delta^* \leq (1 + \varepsilon)\Delta^*$, and $|P'| = O(1/\varepsilon^d)$.
- P' can be computed in linear time, so Δ can be computed in $O(1/\varepsilon^{2d})$ time by brute force.
- We now explain how to improve this result by a simple observation.

Grid Cleaning: Example in \mathbb{R}^2



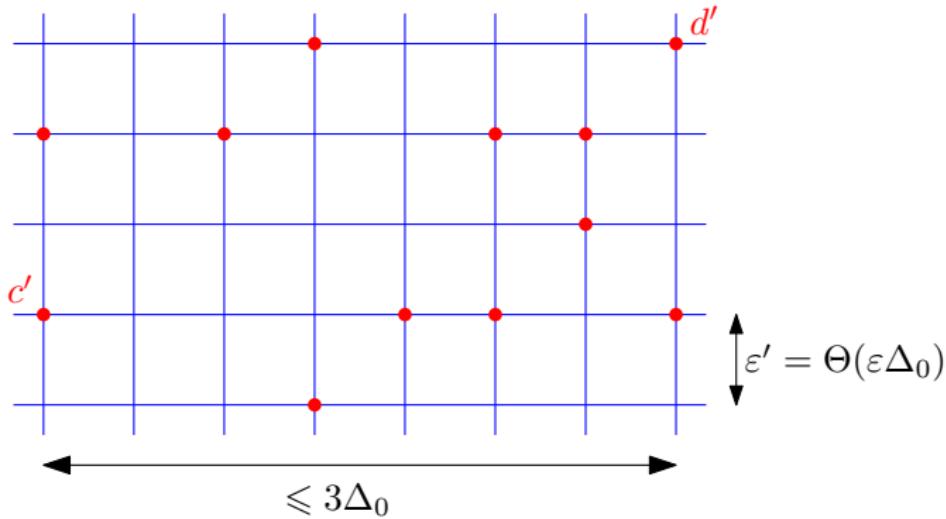
- c' is the lowest point on a vertical line
- d' is the highest point on a vertical line

Grid Cleaning: Example in \mathbb{R}^2



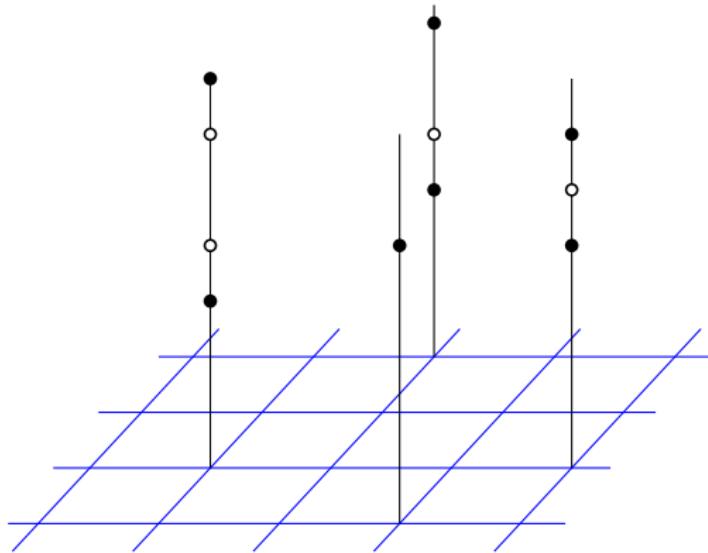
- We only keep the highest and the lowest point on each vertical line.
- Then run the brute-force algorithm.

Analysis



- When $d = 2$, only $O(1/\varepsilon)$ points remain after grid cleaning, because there are $O(1/\varepsilon)$ vertical lines containing a point of P' .
- So we are left with $O(1/\varepsilon)$ points instead of $O(1/\varepsilon^2)$.
- The algorithm runs in $O(n + 1/\varepsilon^2)$ time instead of $O(n + 1/\varepsilon^4)$.

Analysis



- In higher dimension, consider rounded points that coincide in their first $(d - 1)$ coordinates.
- Keep only highest and lowest. Then only $O((1/\varepsilon)^{d-1})$ points remain.
- Compute their diameter by brute force.

Analysis

- Grid cleaning can be done in $O(1/\varepsilon^{d-1})$ time:
- WLOG, suppose that the smallest value of the i th coordinate of the points in P' is 0 for all i .
- Construct a $(d - 1)$ -dimensional array $L[0 \dots E][0 \dots E] \dots [0 \dots E]$ where $E = \Theta(1/\varepsilon)$.
- $L[k_1][k_2] \dots [k_{d-1}]$ records the lowest point on the vertical line through the point $(k_1\varepsilon', k_2\varepsilon', \dots, k_{d-1}\varepsilon')$.
- This array has $O(1/\varepsilon^{d-1})$ cells.
- Same with the highest point.
- So rounding + grid cleaning yields a running time $O(n + 1/\varepsilon^{2d-2})$, instead of $O(n + 1/\varepsilon^{2d})$.

Projecting on Lines

- We measure angles in radian.
- That is, an angle is in $[0, 2\pi]$.

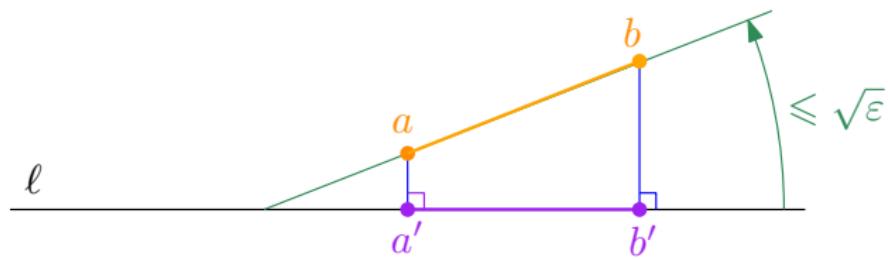
Property

For any α ,

$$1 - \frac{\alpha^2}{2} \leq \cos \alpha \leq 1.$$

- Idea: We get a relative error ε by choosing α to be roughly $\sqrt{\varepsilon}$.

Projecting on Lines



$$|a'b'| \leq |ab| \leq (1 + \varepsilon)|a'b'|$$

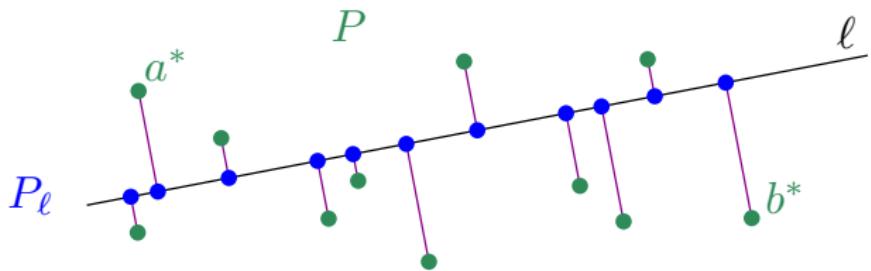
Projecting on Lines

- Assume that the angle between line ab and line ℓ is at most $\sqrt{\varepsilon}$.
- a' (resp. b') is the orthogonal projection of a (resp. b) into ℓ .
- Then $|a'b'| \leq |ab|$ and

$$\begin{aligned}|ab| &\leq \frac{|a'b'|}{\cos \sqrt{\varepsilon}} \\&\leq |a'b'| \times \frac{1}{1 - \varepsilon/2} \\&= |a'b'| \times \left(1 + \frac{\varepsilon}{2} + \frac{\varepsilon^2}{4} + \dots\right) \\&\leq |a'b'|(1 + \varepsilon)\end{aligned}\quad \text{since } \varepsilon < 1$$

- In other words, $|a'b'|$ is a $(1 + \varepsilon)$ -factor approximation of $|ab|$.

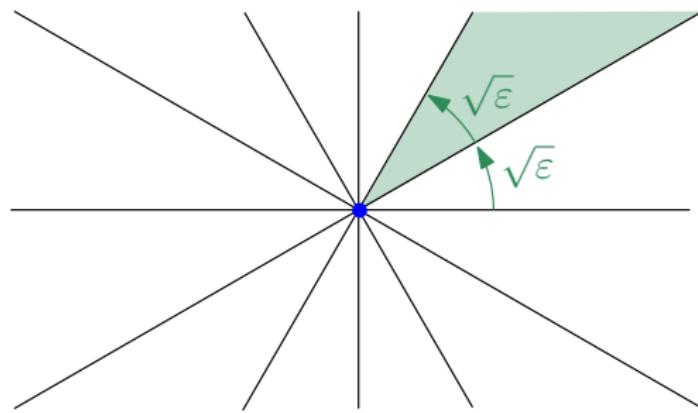
Approach



- P_ℓ is obtained by projecting P onto a line ℓ .
- Compute $\text{diam}(P_\ell)$.
- Can be done in $O(n)$ time: Find maximum and minimum along ℓ .

Approach

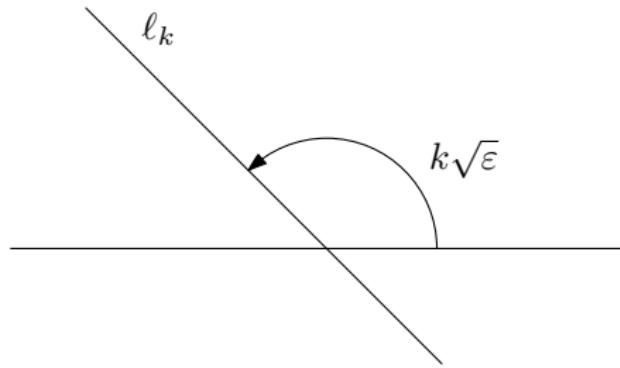
- If the angle between ℓ and a^*b^* is less than $\sqrt{\varepsilon}$, then $\text{diam}(P_\ell)$ is a $(1 + \varepsilon)$ -factor approximation of Δ^* .
- How can we find a line that makes an angle $\sqrt{\varepsilon}$ with a^*b^* ?



$O(1/\sqrt{\varepsilon})$ cones with
angular diameter $\sqrt{\varepsilon}$

- Take several lines. In the plane: angle $k\sqrt{\varepsilon}$ for each $k \in [0, \pi/\sqrt{\varepsilon}]$.

Algorithm in \mathbb{R}^2



- For each integer $k \in [0, \pi/\sqrt{\varepsilon}]$, we denote by ℓ_k a line that makes angle $k\sqrt{\varepsilon}$ with horizontal. Project P onto ℓ_k , obtaining P_{ℓ_k} .
- Then we will prove that $\Delta = \max_k (\text{diam}(P_{\ell_k}))$ is a $(1 + \varepsilon)$ -factor approximation of $\text{diam}(P)$.

Algorithm in \mathbb{R}^2 : Analysis

- Projecting onto a particular ℓ_k takes time $O(n)$.
- Computing $\text{diam}(P_{\ell_k})$ takes time $O(n)$.
- There are $O(1/\sqrt{\varepsilon})$ such lines.
- Overall running time: $O(n/\sqrt{\varepsilon})$.

Algorithm in \mathbb{R}^2 : Proof

- Let θ be the angle of a^*b^* with horizontal.
- There exists k such that $k\sqrt{\varepsilon} \leq \theta < (k+1)\sqrt{\varepsilon}$.
- The angle between a^*b^* and ℓ_k is at most $\sqrt{\varepsilon}$.
- So $\text{diam}(P_{\ell_k})$ is at least $\Delta^*/(1+\varepsilon)$.
- On the other hand, the algorithm only looks at distances between two projected points, which are always smaller than Δ^* .
- So we have

$$\frac{1}{1+\varepsilon}\Delta^* \leq \max_k (\text{diam}(P_{\ell_k})) \leq \Delta^*$$

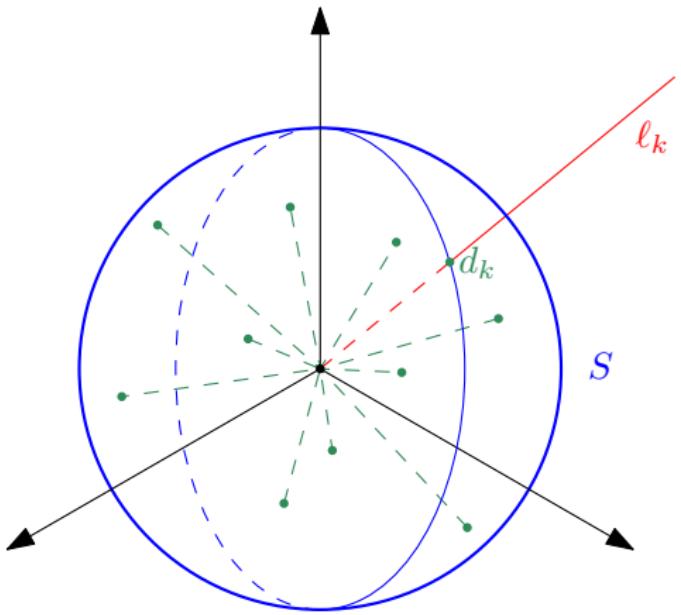
which means that

$$\frac{1}{1+\varepsilon}\Delta^* \leq \Delta \leq \Delta^*.$$

Algorithm in \mathbb{R}^2 : Proof

- It implies that $(1 - \varepsilon)\Delta^* \leq \Delta \leq \Delta^*$.
- We say that these inequalities mean that Δ is a $(1 - \varepsilon)$ -approximation of Δ^* .
- According to the definition from the previous lecture, we wanted to prove that $\Delta^* \leq \Delta \leq (1 + \varepsilon)\Delta^*$.
- There is not a big difference: In both cases, the relative error is ε .
- For instance, when $\varepsilon = 0.01$, in both cases, we make a 1% error.

Generalization in \mathbb{R}^d



Generalization in \mathbb{R}^d

- Problem: Find a set of directions that approximates well the set of all directions.
- Reformulation:
- Let S be the unit sphere in \mathbb{R}^d .
- Find $D \subset S$ with small cardinality such that $\forall x \in S$ there is a point $d_k \in D$ such that $|d_k x| \leq \sqrt{\varepsilon}$.
- Each point $d_k \in D$ is associated with the line ℓ_k through the origin and d_k .
- d_k handles a cone of direction with angular radius $\sqrt{\varepsilon}$.
- Such a set D with cardinality $O(1/\varepsilon^{(d-1)/2})$ can be computed efficiently.
- So the algorithm runs in time $O(n/\varepsilon^{(d-1)/2})$.

Combining the two Techniques

- Running times:
 - Grid + cleaning: $O(n + 1/\varepsilon^{2d-2})$.
 - Projections: $O(n/\varepsilon^{(d-1)/2})$.
- Improvement:
 - First round to P' and do grid cleaning.
 - We are left with $O(1/\varepsilon^{d-1})$ points.
 - Project on lines.
 - Overall running time: $O(n + 1/\varepsilon^{3(d-1)/2})$
- Technical problem: the relative error is now bounded by $\approx 2\varepsilon$. How can we solve it?