

Uniformity of point samples using gap ratio

Sameer Desai

Supervisor: Dr. Arijit Bishnu

Outline

- 1 Introduction
- 2 Previous Works
- 3 Our Results
- 4 Lower Bounds
- 5 Hardness
- 6 Approximation Hardness
- 7 Approximation
- 8 Coreset
- 9 The unit square
- 10 Gap Ratio and discrepancy

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Motivation

Generating uniformly distributed points over a specific domain has applications in

- digital halftoning(Asano, 2008; Teramoto, Asano, Katoh, Doerr, 2006; Zhang et. al., 2011)
- numerical integration (Chazelle: The Discrepancy Method; Matousek: Geometric Discrepancy)
- computer graphics (Chazelle: The Discrepancy Method)
- mesh generation(Asano, Teramoto, 2007)

Motivation

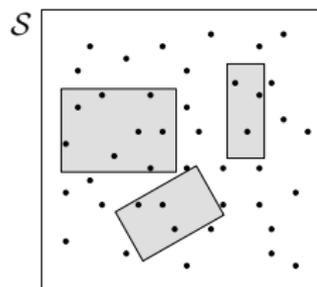
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The Problem

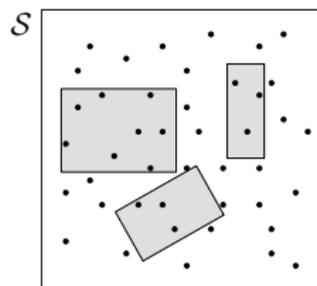
Given a metric space (\mathcal{M}, δ) , we want to choose a set $P \subseteq \mathcal{M}$ of k points such that it is **uniformly** distributed in \mathcal{M} .

Measures of Uniformity: Discrepancy



- The discrepancy $D(P)$ of a point set P is,
$$D(P) = \sup_{B \in \mathcal{R}} |n \cdot \text{vol}(B) - |P \cap B||,$$
where \mathcal{R} is the set of rectangles contained in the unit square.

Measures of Uniformity: Discrepancy



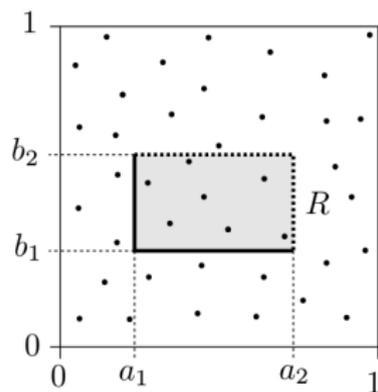
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Computing the measure

How to compute the above discrepancy measure? Seems difficult.

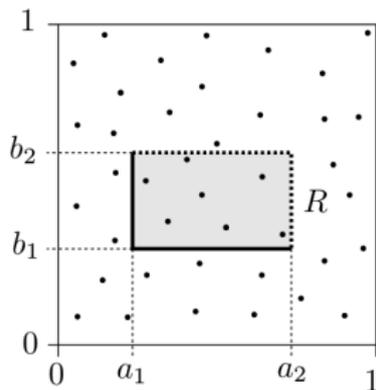
Measures of Uniformity: Discrepancy (contd.)

- What if we restrict R to be the set of all axis-parallel rectangles inside \mathcal{S} ?



Measures of Uniformity: Discrepancy (contd.)

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Solution

For n points in a square, $O(n^2 \log n)$ time is required to compute this value (Dobkin and Gunopulos, 1994).

The problem definition (Generalization)

Definition of *Gap Ratio* (A generalization of Teramoto, et. al., 2006)

- Let (\mathcal{M}, δ) be a metric space and let P be a set of n points sampled from \mathcal{M} .

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- The gap ratio for the point set P is defined as $GR_P := \frac{R_P}{r_P}$.

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The problem

We look to sample a set of points P from \mathcal{M} such that GR_P is minimised.

Making sense of the definition

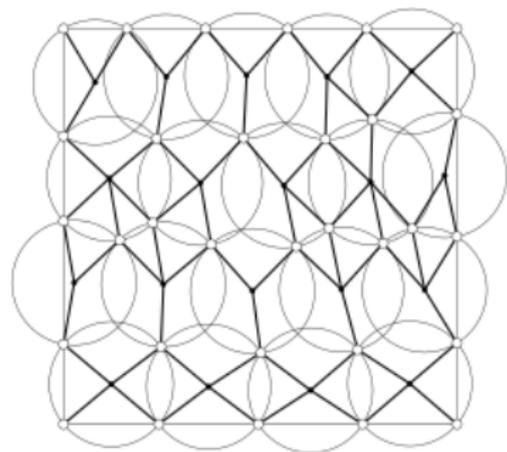


Figure: Maximum Gap = Covering Radius

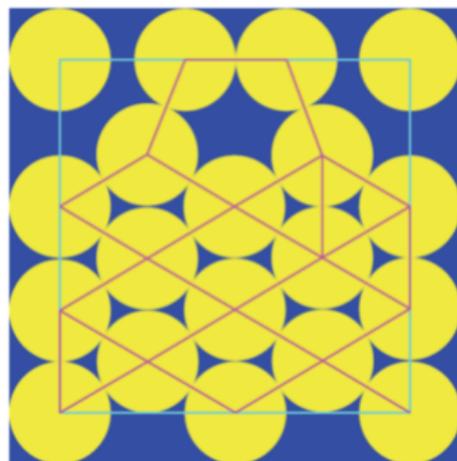


Figure: Minimum Gap = Packing Radius

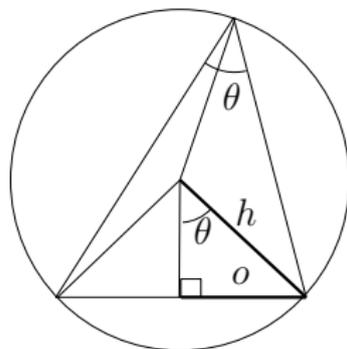
The problem definition

The optimization problem

Given a metric space (\mathcal{M}, δ) , an integer k and a parameter g , find a set $P \subset \mathcal{M}$ such that $|P| = k$ and $GR_P \leq g$.

Gap Ratio and Delaunay

Consider the Delaunay graph of a point set with gap ratio g . Let θ be the minimum angle in the delaunay triangulation.



Clearly $R \geq h$ and $r \leq o \implies \sin \theta = \frac{o}{h} \geq \frac{r}{R} = \frac{1}{g}$

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Previous Works

Up until now the problem has only been studied in an online setting where the gap ratio is minimised at every stage.

- At least $2^{\frac{\lfloor n/2 \rfloor}{\lfloor n/2 \rfloor + 1}}$ where n points are inserted in the interval $[0, 1]$
- At most 2 where n integral points are inserted in the interval $[0, m]$ where m is a positive integer and $0 < n < m$
- No online algorithm can achieve strictly less than 2.5 for a 3×3 grid (discrete space)

Previous results (Algorithms)

Algorithms:

- Linear time optimal algorithm for inserting n points in the interval $[0, 1]$
- An insertion strategy that achieves a gap ratio of at most 2 in a unit square
- An insertion strategy that achieves a gap ratio of at most $2\sqrt{2}$ in a bounded two dimensional grid

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A Summary of Results

Metric space		Lower Bounds	Hardness	Approximation Factor
General		none	Yes	2-approx. hard
Discrete	connected Graph	$\frac{2}{3}$	Yes	approx. factor 3
	Euclidean	-	-	$\frac{3}{2}$ -approx. hard
Continuous	Path Connected	1	Yes	$(1 + \epsilon)$ -approx. algo.
	Unit Square in \mathbb{R}^2	$\frac{2}{\sqrt{3}} - o(1)$	-	approx. factor $\sqrt{3} + o(1)$

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No lower bound for gap ratio

Note the **maximum gap** and **minimum gap** are **not** the maximum and minimum of the same parameter.

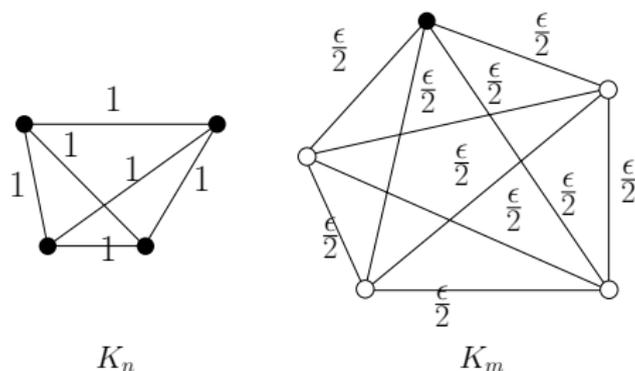


Figure: Discrete space (The black points form the sampled set.)

- Let $\mathcal{M} = V[K_n] \cup V[K_m]$ and the metric $\delta =$ edge weights of $K_n \cup K_m$ and ∞ , when there is no edge.
- $R = \epsilon/2$ and $r = 1/2$. Thus, $GR = \epsilon$.

No lower bound for gap ratio

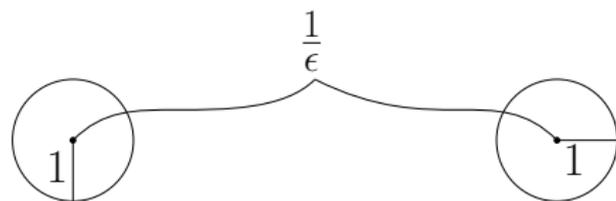
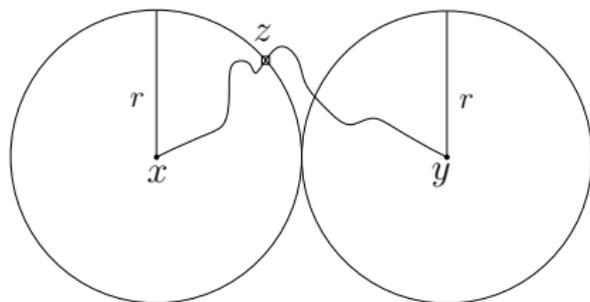


Figure: Continuous Space (The black points form the sampled set).

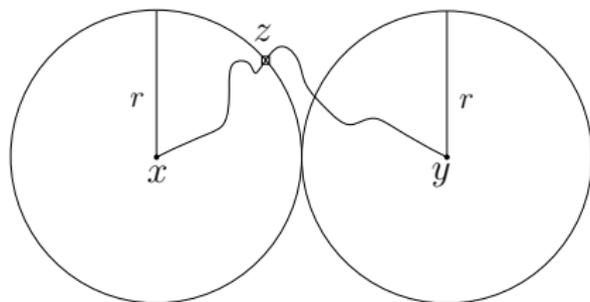
- Let \mathcal{M} be the union of two balls A and B in \mathbb{R}^d of diameter 1 with distance between their centres being $\frac{1}{\epsilon}$, where $0 < \epsilon < 1$.
- P is any two points, one in A and the other in B .
- $R \leq 1$, and $r \geq \frac{1}{2\epsilon} - \frac{1}{2}$. Thus, $GR \leq \frac{2\epsilon}{1+\epsilon}$.

Lower bound for Path-connected space is 1



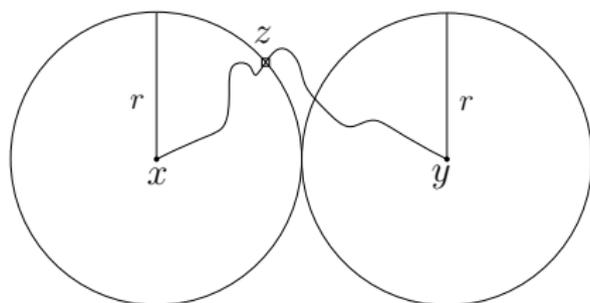
- Let x and y be the closest pair of points in a sampled set P from a path connected space (\mathcal{M}, δ) . $\delta(x, y) = 2r$.

Lower bound for Path-connected space is 1



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- Draw discs of radius r around x and y . The discs should be interior disjoint.

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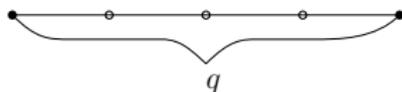
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- A point $z \in \mathcal{M}$ as shown should exist on the boundary of one of the discs because of path connectedness.

Lower Bounds: Graph (Discrete space)

- Let \mathcal{M} be the set of vertices of a connected graph and δ be the length of the shortest path.

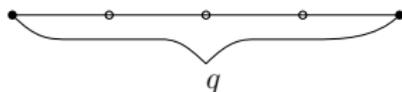
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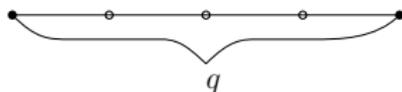
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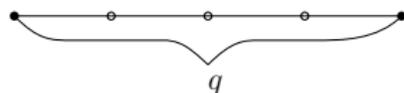
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- This gives us $\frac{R}{r} \geq f(q) = \frac{2}{q} \lfloor \frac{q}{2} \rfloor$.

$$\text{For } q \geq 2, f(q) = \begin{cases} 1 & q \text{ is even} \\ \frac{q-1}{q} & q \text{ is odd} \end{cases}$$

$q = 1$ is an exception as R is always at least 1.

$$\text{Thus } \frac{R}{r} \geq f(q) \geq \frac{2}{3}.$$

Lower Bounds: Unit square

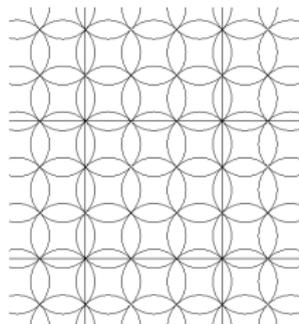
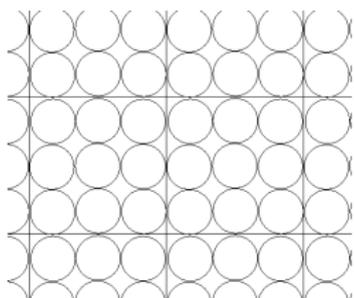
Definition (Packing and covering densities)

The density of a family \mathcal{S} of sets with respect to a set C of Lebesgue measure $\mu(C) < \infty$, is defined as $d(\mathcal{S}, C) = \left(\sum_{S \in \mathcal{S}, S \cap C \neq \emptyset} \mu(S) \right) / \mu(C)$.

Basically, Density = $\frac{\text{Total area covered}}{\text{Area of space}}$.

For an unbounded area we may just take the limit over an unbounded area.

Lower Bounds: Unit Square (contd..)



Bounds on packing and covering density of plane by congruent disks.
(Kuperberg, 1987)

Packing density: $d_1 \leq \frac{\pi}{\sqrt{12}}$ and Covering density: $D_1 \geq \frac{2\pi}{\sqrt{27}}$.

Lower Bounds: Unit Square (contd..)

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Using the upper bound on packing density

$$\frac{k\pi r^2}{(1+2r)^2} \leq \frac{\pi}{\sqrt{12}} \implies r \leq \left(\sqrt{k\sqrt{12}} - 2 \right)^{-1}$$

Lower Bounds: Unit Square (contd..)

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Using the lower bound on covering density

$$\frac{k\pi R^2}{1} \geq \frac{2\pi}{\sqrt{27}} \implies R \geq \sqrt{\frac{2}{k\sqrt{27}}}$$

Lower Bounds: Unit Square (contd..)

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Take the ratio

Thus we have $\frac{R}{r} \geq \frac{2}{\sqrt{3}} - o_k(1)$.

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Hardness Results

Theorem

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Let \mathcal{M} be a continuous metric space. Then, it is NP-hard to find a finite set $P \subset \mathcal{M}$ of cardinality k such that the gap ratio of P is at most $\frac{2}{q}$ for some $q > 2$.

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Theorem

It is NP-hard to find a set of k points in a path connected space such that gap ratio is 1.

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$\frac{3}{2}$ -Approximation Hardness on graphs

Theorem

In an unweighted graph, it is NP-hard to approximate the gap ratio better than a factor of $\frac{3}{2}$.

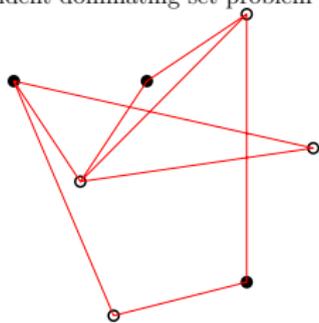
2-Approximation Hardness

Theorem

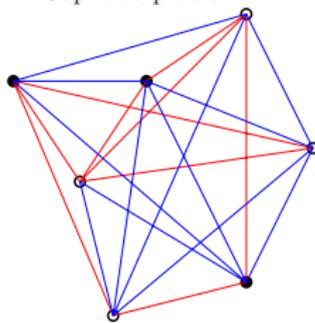
In a general metric space, it is NP-hard to approximate the gap ratio better than a factor of 2.

2-Approximation Hardness

G : Instance for the independent dominating set problem



G : Instance for the Gap ratio problem



Edge weights in G'



r	R	1	2
$\frac{1}{2}$		2	4
1		1	2

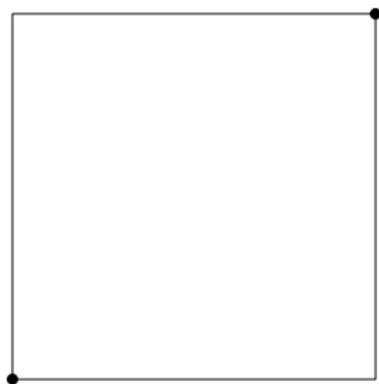
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Farthest point Method

Start with a pair of farthest points in \mathcal{M} . Call this set P_2 .

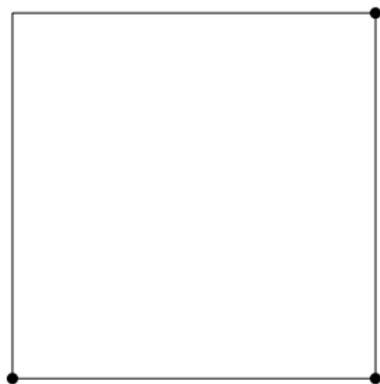
P_2



Farthest point Method

$$P_i \leftarrow P_{i-1} \cup \{p_i\} \text{ s.t., } \delta(p_i, P_{i-1}) = \sup_{x \in \mathcal{M}} \delta(x, P_{i-1})$$

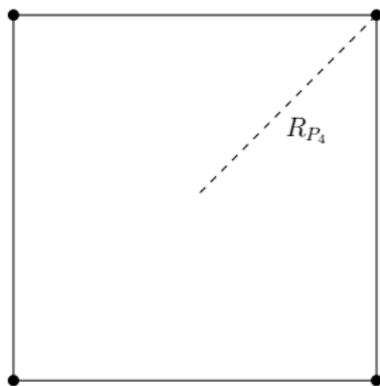
P_3



Farthest point Method

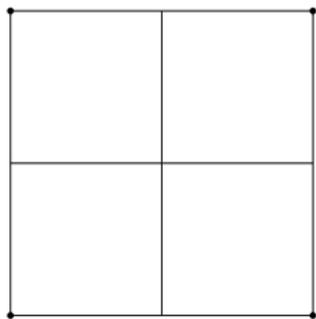
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P_4



Farthest point Method

$$P_i \leftarrow P_{i-1} \cup \{p_i\} \text{ s.t., } \delta(p_i, P_{i-1}) = \sup_{x \in \mathcal{M}} \delta(x, P_{i-1})$$



$$GR_{P_i} = \frac{R_{P_i}}{r_{P_i}} = \frac{2R_{P_i}}{R_{P_{i-1}}} \leq 2$$

The approximation result

Without loss of generality, let $P = \{p_1, \dots, p_k\}$ be the set with optimal gap ratio, and let $GR = \alpha$.

Theorem

Farthest point insertion gives the following approximation guarantees: (i) if $\alpha \geq 1$, then the approximation ratio is $\frac{2}{\alpha} \leq 2$, (ii) if $\frac{2}{3} \leq \alpha < 1$, the approximation ratio is $\frac{2}{\alpha} \leq 3$, and (iii) if $\alpha < \frac{2}{3}$, the approximation ratio is $\frac{4}{2-\alpha} < 3$.

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(k, ϵ) -coreset for set of points in \mathbb{R}^d

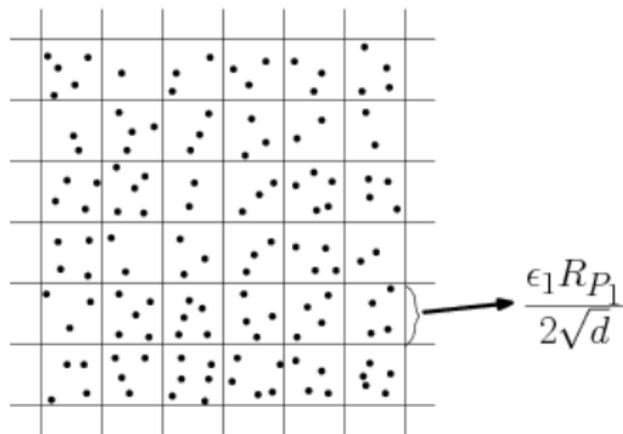
Coreset

- A (k, ϵ) – *coreset* of (\mathcal{M}, δ) for the gap ratio problem is a set (\mathcal{C}, δ) , $\mathcal{C} \subset \mathcal{M}$, such that for any $P \subsetneq \mathcal{C}$, $(|P| = k)$,
- $(1 - \epsilon) GR_P^{\mathcal{M}} \leq GR_P^{\mathcal{C}} \leq (1 + \epsilon) GR_P^{\mathcal{M}}$.

ϵ -coreset for set of points in \mathbb{R}^d

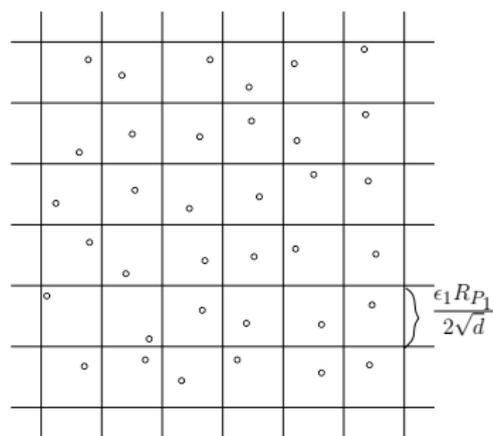
Assume $\epsilon \in (0, \frac{1}{2})$ and $\epsilon_1 := \frac{\epsilon}{3+2\epsilon} < \epsilon$.

Sample set P_1 ($|P_1| = k$) by the farthest point algorithm take a grid as below.



ϵ -coreset for set of points in \mathbb{R}^d

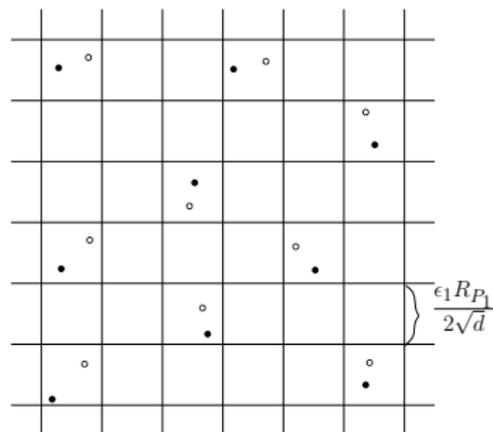
For each cell of the grid choose one point of \mathcal{M} .



$$R_{\min} \leq R_{P_1} \leq 2R_{\min} \implies O\left(k \left\lceil \frac{1}{\epsilon_1} \right\rceil^d\right) \text{ points are chosen.}$$

$(1 + \epsilon)$ -algorithm

This set has k points "close" to the optimal set. So brute force on it gives an $(1 + \epsilon)$ -approximation of the gap ratio.



- Optimal Set
- $(1 + \epsilon)$ approximation

Running time of $(1 + \epsilon)$ -algorithm

- We look at $O\left(\left(k \left\lceil \frac{1}{\epsilon_1} \right\rceil^d\right)^k\right)$ sets.
- Finding closest pair takes $O(k \log k)$.
- Computing the maximum gap takes $O((n - k)k)$ time
- Total runtime: $O\left(k^k \left\lceil \frac{1}{\epsilon_1} \right\rceil^{kd} (k \log k + (n - k)k)\right)$ time.

The runtime is comparable to the best known algorithms of k -center.

$(1 + \epsilon)$ -algorithm (Streaming)

- In the streaming case the idea is to maintain a coresets similar to the above.
- The key to the coresets selection: farthest point algorithm gave a constant approximation of the covering radius, i.e.,
$$R_{\min} \leq R_{P_1} \leq 2R_{\min}$$
- Doubling algorithm for k -centre has this property in the streaming case. To be precise, $R_{\min} \leq R_{P_1} \leq 8R_{\min}$, which allows us again to choose a grid with cell size $O\left(\frac{R_{P_1}}{2\sqrt{d}}\right)$.
- Thus we again get a core-set by choosing one point per grid cell.

To Sum up

Metric space		Lower Bounds	Hardness	Approximation Factor
General		none	Yes	2-approx. hard
Discrete	connected Graph	$\frac{2}{3}$	Yes	approx. factor 3
	Euclidean	-	-	$\frac{3}{2}$ -approx. hard
Continuous	Path Connected	1	Yes	$(1 + \epsilon)$ -approx. algo.
	Unit Square in \mathbb{R}^2	$\frac{2}{\sqrt{3}} - o(1)$	-	approx. factor $\sqrt{3} + o(1)$

Outline

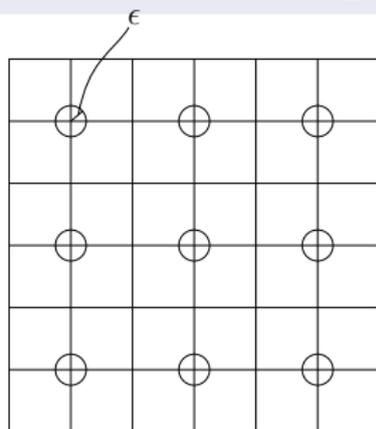
- 1 Introduction
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- 10 Gap Ratio and discrepancy

Randomised methods

Theorem

When n points are sampled using jittered sampling in a unit square, then the gap ratio is $O\left(\sqrt[d]{\frac{2}{n}}\right)$ with probability at least $\frac{1}{2}$.

- $m \times m$ grid
- Ball of radius ϵ , centred at alternate grid points
- At most $\left(\frac{m}{d}\right)^d$ balls are considered.
- A_i denotes the event that the i -th ball is hit by two points.



$$\begin{aligned}
& \mathbb{P}(2r \leq 2\epsilon) \geq \mathbb{P}(\cup A_i) \\
& \geq \sum \mathbb{P}(A_i) - \sum \mathbb{P}(A_i \cap A_j) \\
& = \sum \binom{2^d}{2} \left(\frac{c_d \epsilon^d}{2^d} / \frac{1}{m^d} \right)^2 - \sum \binom{2^d}{2}^2 \left(\frac{c_d \epsilon^d}{2^d} / \frac{1}{m^d} \right)^4 \\
& = \left(\frac{m}{d} \right)^d \binom{2^d}{2} \left(\frac{c_d \epsilon^d}{2^d} / \frac{1}{m^d} \right)^2 - \left(\left(\frac{m}{d} \right)^d \right) \binom{2^d}{2}^2 \left(\frac{c_d \epsilon^d}{2^d} / \frac{1}{m^d} \right)^4 \\
& = \left(\frac{m}{d} \right)^d \binom{2^d}{2} \left(\frac{c_d \epsilon^d m^d}{2^d} \right)^2 \\
& \quad - \frac{1}{2} \left(\frac{m}{d} \right)^d \left(\left(\frac{m}{d} \right)^d - 1 \right) \binom{2^d}{2}^2 \left(\frac{c_d \epsilon^d m^d}{2^d} \right)^4
\end{aligned}$$

$$\text{Set } \epsilon^{2d} = 2^{2d} \binom{2^d}{2}^{-1} m^{-2d} \left(\left(\frac{m}{d} \right)^d - 1 \right)^{-1} c_d^{-2}.$$

$$\text{Thus, } \mathbb{P}(2r \leq 2\epsilon) \geq \frac{1}{2} \frac{\left(\frac{m}{d} \right)^d}{\left(\left(\frac{m}{d} \right)^d - 1 \right)} > \frac{1}{2}$$

$$\text{Use } R \geq \frac{\sqrt{d}}{2m}$$

Randomised methods

Theorem

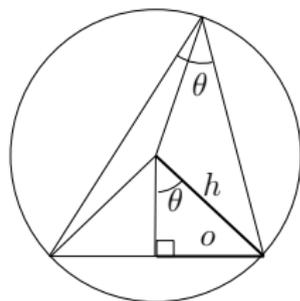
Suppose n points are sampled uniformly at random in a d -dimensional unit cube. Then the gap ratio $\rightarrow \infty$ as $n \rightarrow \infty$ with probability 1.

A tight lower bound for the unit square

Theorem

Let the point set P be the vertex set of a constrained Delaunay triangulation of the square with $|P|$ points such that the maximum angle is at least $\frac{\pi}{2}$. Then $GR_P \geq \sqrt{2}$.

A tight lower bound for the unit square



Consider any triangular face Δ of the Delaunay mesh of P and any angle θ of it. Let c be the radius of the circumcircle of Δ .

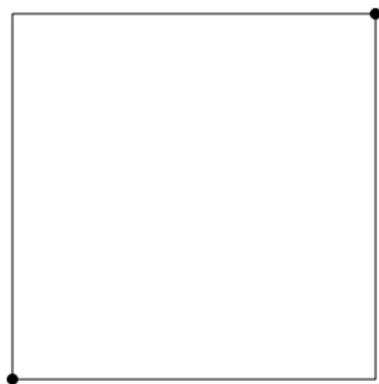
Proof

- $s \geq r$, $c \leq R$; where r : minimum gap and R : maximum gap.
- $\sin \theta = \frac{s}{c} \geq \frac{r}{R} = \frac{1}{g}$.
- Let θ_{\max} : largest angle; then, $\theta_{\max} \geq \frac{\pi}{2}$.
- Then, $g^{-1} \leq \sin \theta_{\min} \leq \sin \frac{\pi - \theta_{\max}}{2} \leq \sin \frac{\pi}{4} \leq \frac{1}{\sqrt{2}}$

Farthest Point algorithm achieves the lower bound

Initialisation: Sample a farthest pair of points.

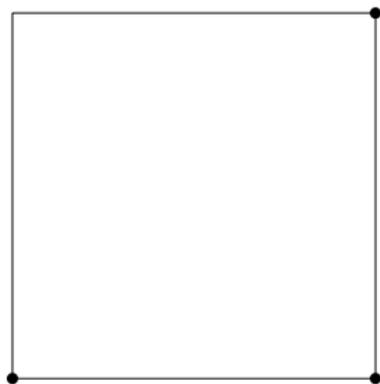
P_2



Farthest Point algorithm achieves the lower bound

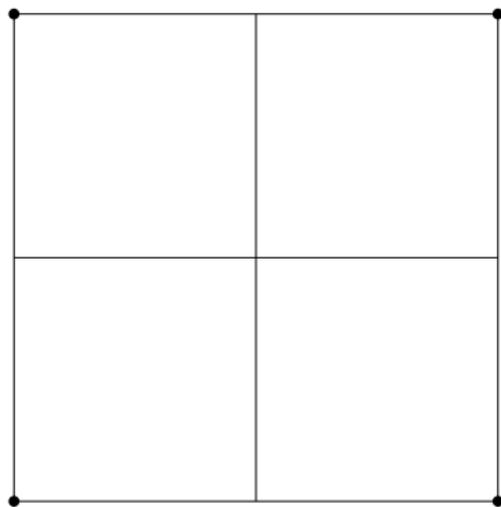
$$P_i \leftarrow P_{i-1} \cup \{p_i\} \text{ s.t., } \delta(p_i, P_{i-1}) = \sup_{x \in \mathcal{M}} \delta(x, P_{i-1})$$

P_3



Farthest Point algorithm achieves the lower bound

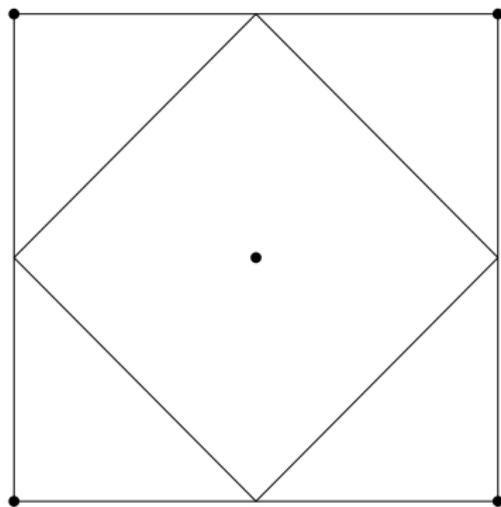
We continue for a few iterations.



Gap ratio = $\sqrt{2}$.

Farthest Point algorithm achieves the lower bound

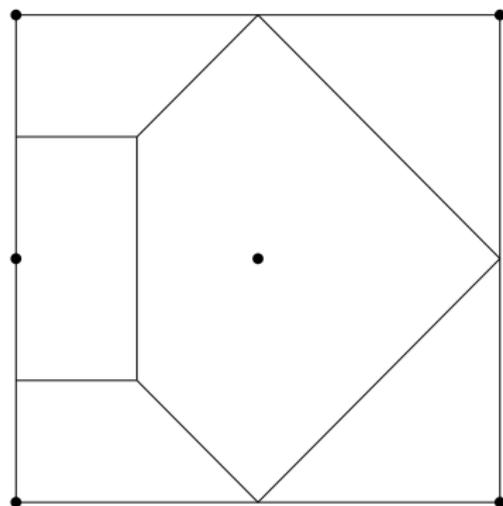
We continue for a few iterations.



Gap ratio = $\sqrt{2}$.

Farthest Point algorithm achieves the lower bound

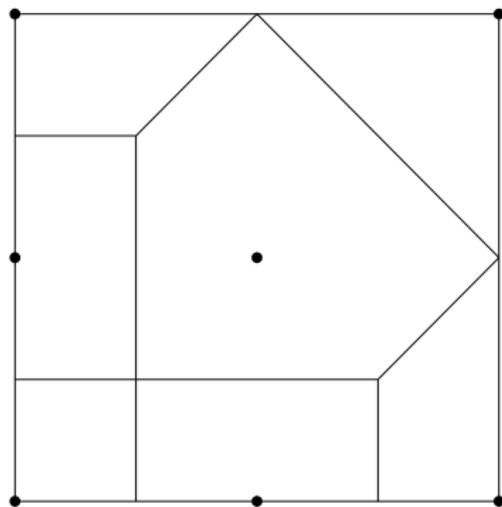
We continue for a few iterations.



Gap ratio = 2.

Farthest Point algorithm achieves the lower bound

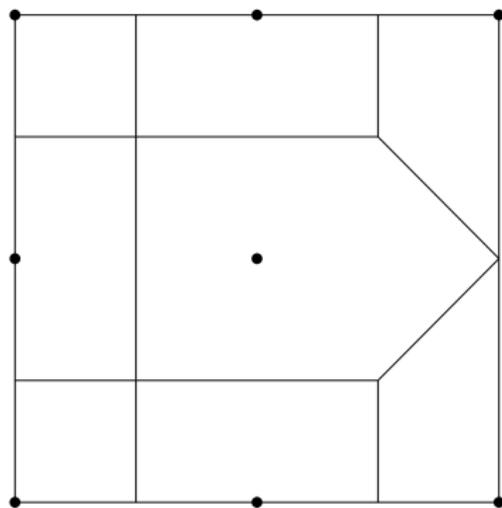
We continue for a few iterations.



Gap ratio = 2.

Farthest Point algorithm achieves the lower bound

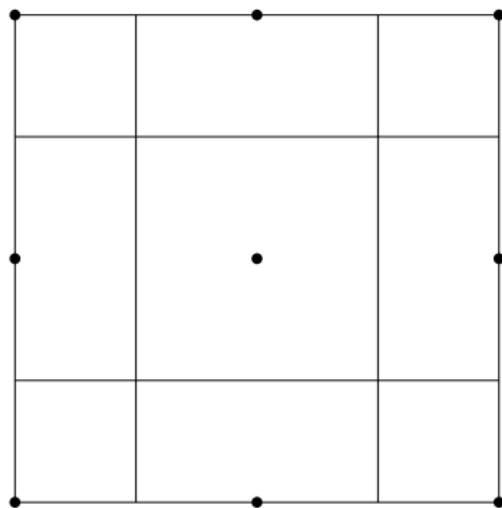
We continue for a few iterations.



Gap ratio = 2.

Farthest Point algorithm achieves the lower bound

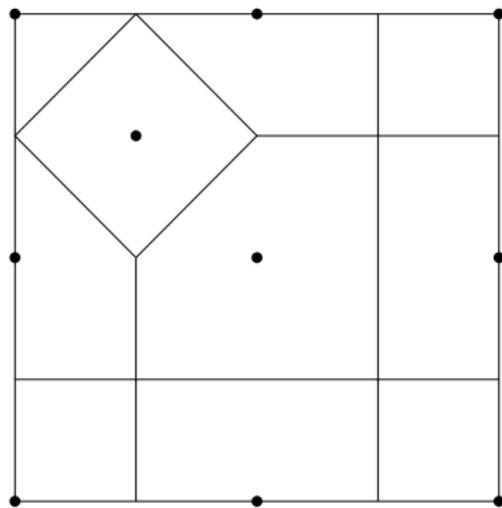
We continue for a few iterations.



Gap ratio = $\sqrt{2}$.

Farthest Point algorithm achieves the lower bound

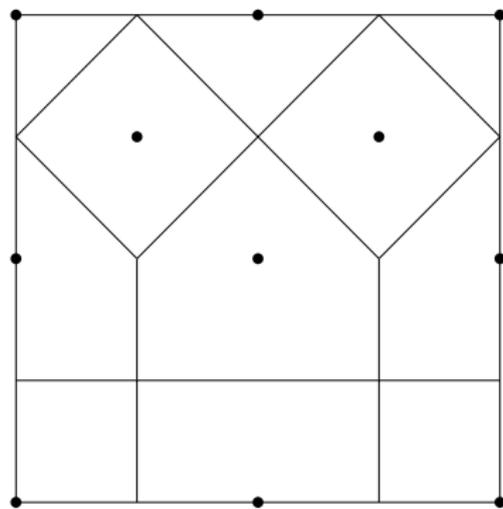
We continue for a few iterations.



Gap ratio = 2.

Farthest Point algorithm achieves the lower bound

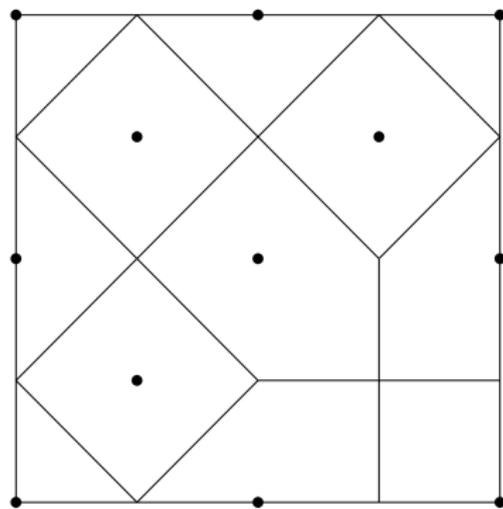
We continue for a few iterations.



Gap ratio = 2.

Farthest Point algorithm achieves the lower bound

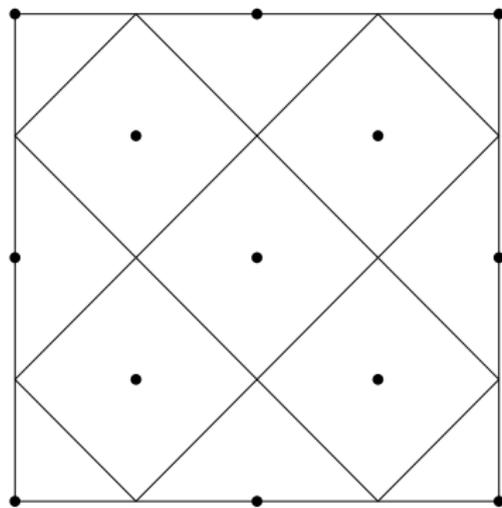
We continue for a few iterations.



Gap ratio = 2.

Farthest Point algorithm achieves the lower bound

We continue for a few iterations.



Gap ratio = $\sqrt{2}$.

Farthest Point algorithm achieves the lower bound

Note that

- Every new point is a Voronoi vertex of the point set from the previous iteration.
- Gap ratio becomes $\sqrt{2}$ when the Voronoi diagram tiles the unit square with congruent square cells.
- Point samples of size n , where $n = k^2$ or $n = k^2 + (k - 1)^2$ for some k achieve $\sqrt{2}$.

Theorem

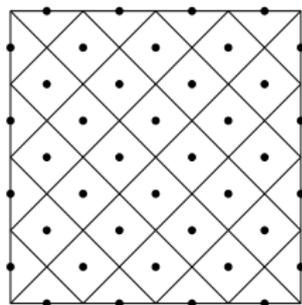
If the set P of size $n \geq 4$ points is sampled from the unit square using the farthest point algorithm then the gap ratio is $\sqrt{2}$ if and only if $n = G_j$ where,

$$G_j = \begin{cases} 4^{i-1} + 2^i + 1 & j = 2i - 1 \\ 2 \cdot 4^{i-1} + 2^i + 1 & j = 2i \end{cases}$$

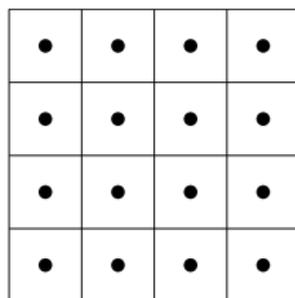
for $i \in \mathbb{N}$.

Other point sets which achieve the lower bound

Not all the grid patterns occur in the farthest point pattern, for eg.:- the 4×4 grid does not occur in the farthest point method. Also grid patterns which do not have the corners of the square do not occur in the farthest point algorithm.



(a)



(b)

Figure: (a) Point sample of size $40 = 2 \cdot 5 \cdot 4 = 2 \cdot k \cdot (k - 1)$ with gap ratio $= \sqrt{2}$; (b) A 4×4 sample with no point on the boundary

Lattices: 1-parameter family

$$\Delta : \mathbb{R}^d \rightarrow \mathbb{R}$$

$$\Delta(x) = \langle x, \mathbf{1} \rangle = \sum_{i=1}^d x_i$$

$$D = \frac{\delta-1}{d}$$

$$T_\delta : \mathbb{R}^d \rightarrow \mathbb{R}^d$$

$$x \mapsto x + D\Delta(x) \cdot \mathbf{1}$$

$$L_\delta := T_\delta(\mathbb{Z}^d)$$

Lattices: 1-parameter family

$$g^2 = \begin{cases} \frac{(d^2-1)+(d^2+2)\delta^2+(d^2-1)\delta^4}{3d^2\delta^2} & 0 \leq \delta \leq \frac{1}{\sqrt{d+1}} \\ \frac{(d^2-1)+(d^2+2)\delta^2+(d^2-1)\delta^4}{3(d+\delta^2-1)} & \frac{1}{\sqrt{d+1}} \leq \delta \leq 1 \\ \begin{cases} \frac{d^2-1+\delta^2}{d+\delta^2-1} & d \text{ is odd} \\ \frac{d^2-2+\delta^2}{d+\delta^2-1} + \frac{1}{\delta^2(d+\delta^2-1)} & d \text{ is even} \end{cases} & 1 \leq \delta \leq \sqrt{d+1} \\ \begin{cases} \frac{d^2-1+\delta^2}{2d} & d \text{ is odd} \\ \frac{d^2-2+\delta^2}{2d} + \frac{1}{2d\delta^2} & d \text{ is even} \end{cases} & \sqrt{d+1} \leq \delta \end{cases}$$

Lattices: Random construction

Theorem

Suppose the fundamental region of a lattice is obtained by using two unit vectors at an angle chosen uniformly at random from $[0, \frac{\pi}{2}]$. Then,

$$\begin{aligned} & \mathbb{P}(\text{Gap Ratio of lattice} \geq g) \\ = & \begin{cases} \frac{1}{2} - \frac{2}{\pi} \arccos\left(\frac{2}{g^2} - 1\right) + \frac{2}{\pi} \arcsin\left(\frac{1}{g}\right) & g \in \left(\frac{2}{\sqrt{3}}, \sqrt{2}\right) \\ \frac{2}{\pi} \arcsin\left(\frac{1}{g}\right) & g \geq \sqrt{2} \end{cases} \end{aligned}$$

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Gap Ratio and discrepancy (1D)

Discrepancy in 1 dimension

$$D_N = \sup_{0 \leq a < b \leq 1} \left| b - a - \frac{A([a, b], P)}{N} \right|$$

Theorem

For N points in the interval $[0, 1]$ with discrepancy $D_N < \frac{2}{n}$,

$$GR \leq \frac{D_N}{\frac{1}{N} - \frac{D_N}{2}} = \frac{4}{2 - ND_N} - 2.$$

Gap Ratio and discrepancy (1D)

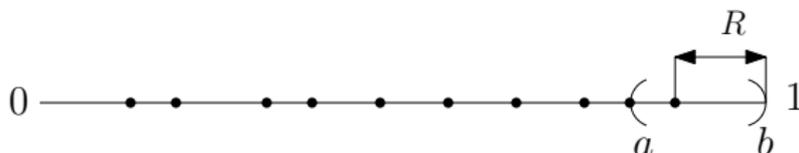
Proof

We can bound r by looking at the smallest interval with exactly 2 points i.e., $-D_N \leq 2r - \frac{2}{N}$. Thus, $r \geq \frac{1}{N} - \frac{D_N}{2}$.



Proof

We can bound the largest interval $[a, b)$ which contains exactly no point as, $b - a \leq D_N$. Thus, we have $R \leq D_N$.



Thanks !