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Approximation Algorithms for

Diameter, Width, and Related Problems

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Timothy M. Chan

School of CS

Univ of Waterloo

# INTRODUCTION

## Why Approximation?

efficiency

e.g. dD width:  $O(n^{\lceil d/2 \rceil}) \rightarrow O(n)$

allows for tougher versions

e.g. dynamic, kinetic, outliers, etc.

## This Talk

focuses on just 2 specific problems  
& on fixed  $d$

surveys a number of known alg's  
& some new ones

+ previews results in "data stream"  
model

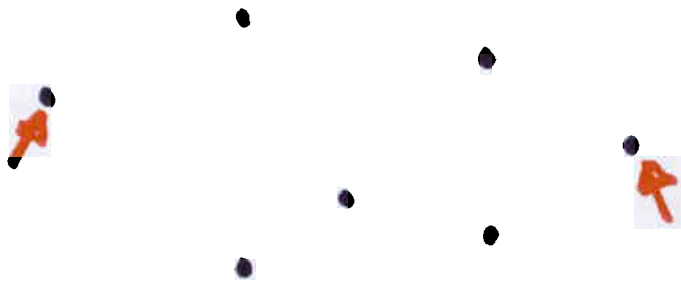
### 3 Strategies to Approx. Schemes

[from Schuurman, Woeginger's survey ~'07]

- \* 1. structuring (e.g. rounding) the input
- \* 2. structuring ( " ) the output
- 3. structuring the execution of an alg'm

# THE DIAMETER PROBLEM

given  $n$  pts  $P$  in  $\mathbb{R}^d$ ,  
find farthest pair



## Quick Const-Factor Alg's

- a)  $s = \text{any pt}$   
 $t = \text{farthest neighbor of } s$

Approx factor: 2

Pf:  $|pq| \leq |ps| + |sq| \leq 2|st|$

□

b) [Egecioglu, Kalantari '89]

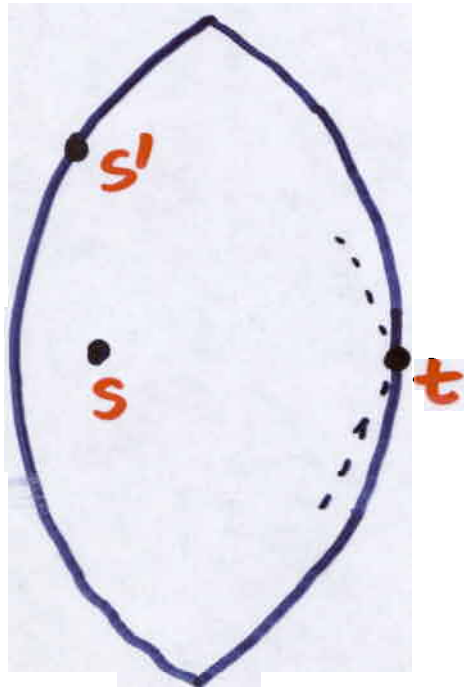
$s$  = any pt

$t$  = farthest neighbor of  $s$

$s'$  = farthest neighbor of  $t$

Approx factor:  $\sqrt{3}$

Pf:



draw lune  
with radius  
 $r = |s't|$

$$\text{Diam}(P) \leq \text{Diam}(\text{lune}) = \sqrt{3} r$$

□

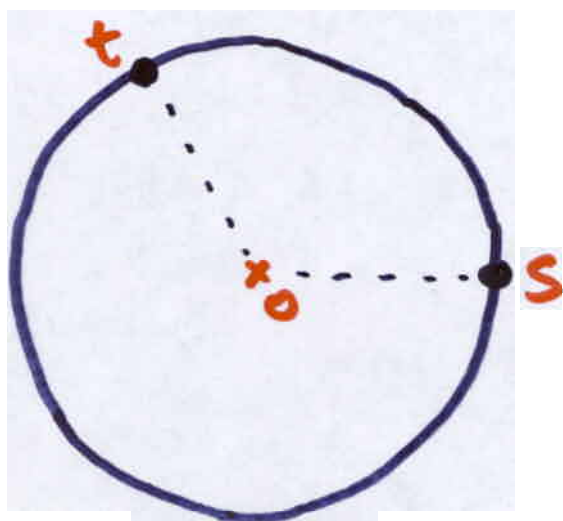


c) [Goel, Indyk, Varadarajan '01]

$B =$  smallest enclosed ball (exact/approx)  
w. center  $o$ , radius  $r$

$s =$  a pt on  $\partial B$

$t =$  a pt on  $\partial B$  w.  $\angle sot \geq 90^\circ$



Factor:  $\sqrt{2}$  (roughly)

Pf:  $\text{Diam}(P) \leq 2r$

$|st| \geq \sqrt{2}r \quad \square$

# Rounding the Input (i.e. Points)

[ Barequet, Har-Peled '99 ]

1.  $D = \text{const-factor approx.}$
2. round pts to grid of side  $\epsilon D$
3. return diameter of rounded pts

Approx factor:  $1 + O(\epsilon)$

Pf: additive error is  $O(\epsilon D)$ .  $\square$

Runtime: 2. gives  $O(1/\epsilon^d)$  pts  
3. by brute force

$$\Rightarrow O(n + 1/\epsilon^{2d})$$

Rmks: can improve to  $O(n + 1/\epsilon^{2(d-1)})$   
idea works for k-center, MDST,  
etc.

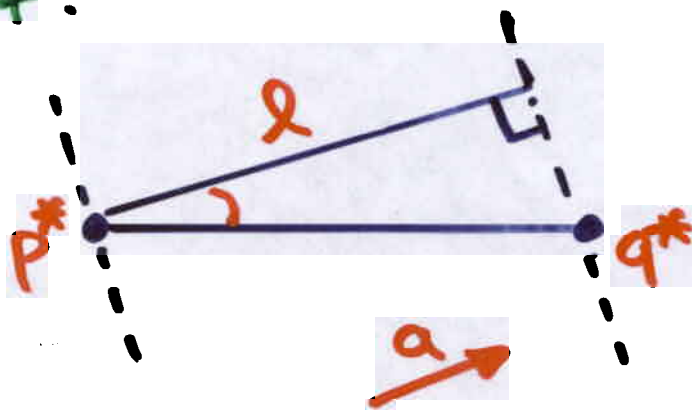
# Rounding "Output" (i.e. Direction)

[Agarwal, Matoušek, Suri '92]

1. form  $O(1/\delta^{d-1})$  vectors  $A$  s.t.  
 $\forall x, \exists a \in A, \angle(a, x) = O(\delta)$
2. find extreme pt pair  $p_a q_a$   
along each direction  $a \in A$

Factor:  $1 + O(\delta^2)$  (set  $\delta = \sqrt{\epsilon}$ )

Pf:



let  $a \in A$

$$\angle(a, p^*q^*) = O(\delta)$$

$$|p_a q_a| \geq \ell = |p^*q^*| / \cos O(\delta) \quad \square$$

Runtime:

$$O(n / \epsilon^{(d-1)/2})$$

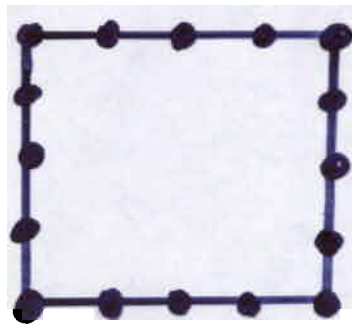


## Options for the vectors $A$ :

- Pack pts on unit sphere [e.g. Yao '82]



- Or grid pts on unit hypercube's bdry



[e.g. Agarwal,  
Har-Peled '01]

## Rmks:

- idea works for nearest neighbors, MST, et
- can slightly improve runtime by data structures for extreme pts

## Combo: Rounding Pts AND Directions [Chan'00]

Runtime:  $O(n + 1/\epsilon^{d-1} + n'/\epsilon^{(d-1)/2})$   
with  $n' = O(1/\epsilon^{d-1})$

$$\Rightarrow O(n + 1/\epsilon^{3(d-1)/2})$$

## More Alg's [Chan'00] (skip)

2 ways to get runtime near  
 $O(n + 1/\epsilon^d)$

idea - recurse in dimension ...

## DIGRESSION: "Discrete Voronoi Diagrams"

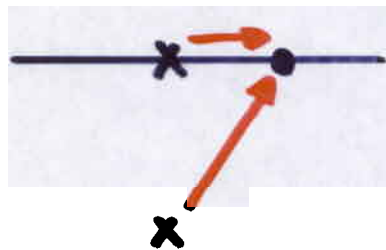
given  $n$  grid pts in  $\{1, \dots, E\}^d$ ,  
find nearest neighbor (NN) of  
every grid pt in  $\{1, \dots, E\}^d$

### Rmks:

- different from "approx. VD's"
- image processing apps
- studied by **Bren, Gil, Kirkpatrick, Warnan '95**

### An Alg'm

idea - recurse in dimension



Rmk: works for farthest neighbors,  
or extreme pts along grid directions

## BACK TO DIAMETER . . .

New (Final?) Alg'm

just run Combo

but use above method to find  
the extreme pts

$\Rightarrow \boxed{O(n + 1/\epsilon^d)}$  time

(or more precisely,  $O(n + 1/\epsilon^{d-(3/2)})$ )

Open:  $O(n + 1/\epsilon^{(d-1)/2})$  time??

comparison w. practical methods?

[Har-Peled '02, Malandain, Boissonat '02]



To find NN of every pt in a kD grid:

1. solve problem for pts in each horizontal slice of G
2. for each vertical line Q of G,  
let  $p_1, \dots, p_E$  be the NNs from 1.  
construct the ID VD of  $p_1, \dots, p_E$  at Q  
look up answers

Runtime:

ID VD reduces to 2D CH,  
computable by Graham scan in  $O(E)$  time

$$T_k(n) = \sum_{i=1}^E T_{k-1}(n_i) + O(E^{k-1} \times E)$$

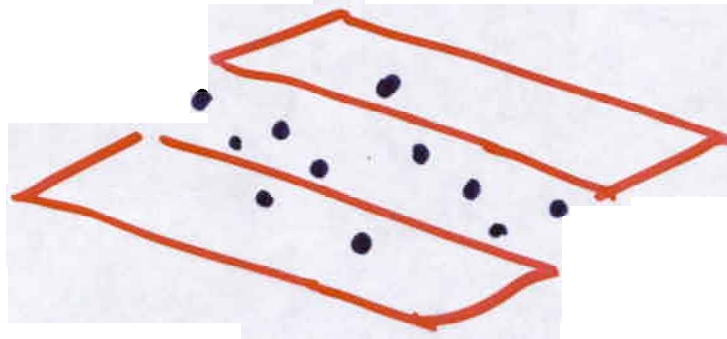
$$\Rightarrow T_d(n) = \boxed{O(E^d)}$$



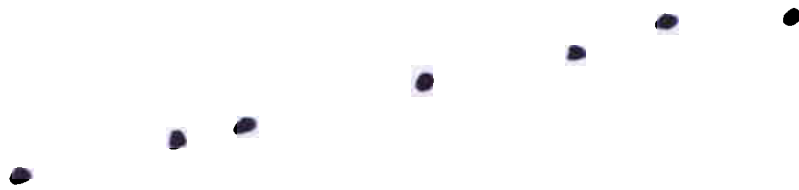
# THE WIDTH PROBLEM

given  $n$  pts  $P$  in  $\mathbb{R}^d$ ,

find 2 parallel hyperplanes  
enclos.  $P$ , with min distance



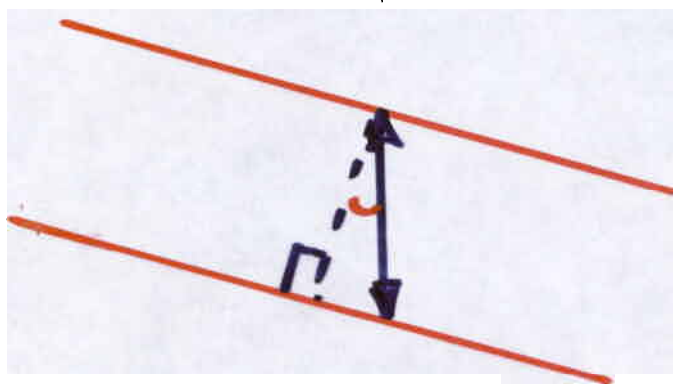
Rounding Pts : doesn't seem to work ...



Rounding Directions: not work either?

# First Alg'm [Duncan, Goodrich, Ramos '97]

1. form vectors  $A$  as before
2. for each  $a \in A$ ,  
find 2 parallel hyperplanes (any direction) enclos.  $P$ , minimizing distance along direction  $a$



← this is LP

↑  $a$

Factor:  $1 + O(\delta^2)$  (set  $\delta = \sqrt{\epsilon}$ )

Pf: let  $a \in A$ ,  $\angle(a, a^*) = O(\delta)$

dist along  $a \leq \text{dist along } a^* / \cos O(\delta)$

Runtime:  $O(n / \epsilon^{(d-1)/2})$

(or slightly better by data structures for LP)

## Rounding Pts (1st version)

1. let  $s, t, W$  be as before
2. draw grid lines **parallel** to  $\overleftrightarrow{st}$  at distance  $\epsilon W, 2\epsilon W, \dots, W$
3. round pts to lie on grid lines
4. return width of rounded pts

Factor:  $1 + O(\epsilon)$

Pf: additive error is  $O(\epsilon W)$   $\square$

Runtime: 2. gives  $O(1/\epsilon)$  pts

$$\Rightarrow O(n + 1/\epsilon)$$

## Rounding Pts (2<sup>nd</sup> version)

1. let  $s, t$  be as before,  $D = |st|$
2. draw grid lines **orthogonal** to  $\overleftrightarrow{st}$  at distance  $\epsilon D, 2\epsilon D, \dots, D$
- 3, 4. as before

Factor:  $1 + O(\epsilon)$

Pf: for each direction  $a$ ,  
width along  $a$  changes by  
 $O(\epsilon \cdot \text{width of } \overleftrightarrow{st} \text{ along } a)$   
 $\leq O(\epsilon \cdot \text{width of } P \text{ along } a)$   $\square$

Runtime:  $O(n + 1/\epsilon)$

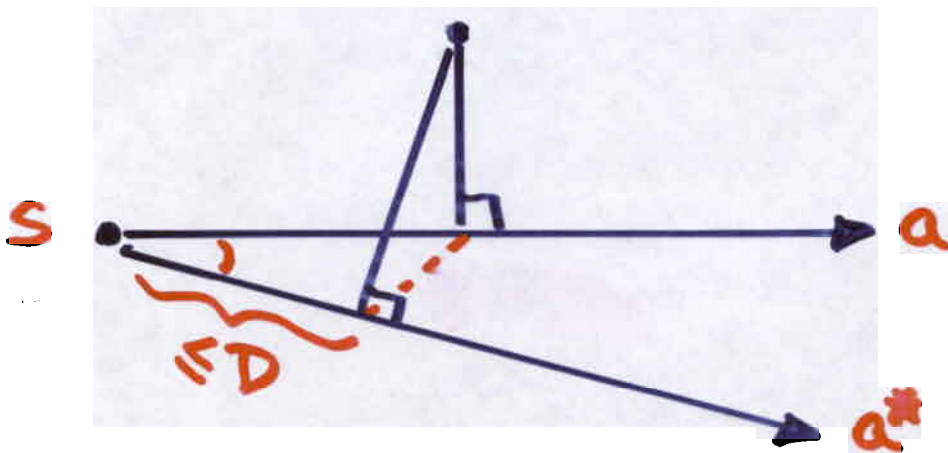


# Rounding Directions [Agarwal, Aronov, Sharir '97]

1. let  $s, t, W, D$  be as before
2. find width along directions at angle  $\epsilon W/D, 2\epsilon W/D, \dots, O(W/D)$  with normal of  $\overrightarrow{st}$

Factor:  $1 + O(\epsilon)$

Pf: say  $\angle(a, a^*) = O(\delta)$



$$| \text{width along } a^* - \text{width along } a | \leq O(D \cdot \epsilon W/D) = O(\epsilon W) \quad \square$$

Runtime:  $O(n/\epsilon)$



## Moral

rounding can work, but need grid over boxes (rotated) instead of hypercubes

## Generalization to $dD$

[ Agarwal, Har-Peled '01 ] (sketch)

take a bounding box [ Barequet, Har-Peled '99 ]  
rotate & re-scale it into hypercube  
now use a standard grid approach  
(round pts and/or directions)

Runtime:  $O(n + 1/\epsilon^{2(d-1)})$

or by a combo,  $O(n + 1/\epsilon^{3(d-1)/2})$

## New Result

run Agarwal, Har-Peled's alg'm  
with our discrete VD method

$$\Rightarrow \boxed{O(n + \frac{1}{\epsilon^{d-1}})} \text{ time}$$

Rmks:

- A.H.'s alg'm approximates width along **all directions** ("extent fn")
- leads to **"core sets"** for min enclos. cylinder / annulus / cylind. shell, kinetic problems, etc.

[Har-Peled, Varadarajan '01]

# DATA STREAM COMPUTATION

## The Model

goal - handle massive data sets

alg'm can make just one pass over input

space is limited (can't store whole input)

## Example: Diameter

recall alg'm by rounding direction

$O(1/\epsilon^{(d-1)/2})$  space &  
time per pt

Let's Try 2D Width ...

traditional alg's, e.g. Janardan '93

$O(n)$  space,  $O((\log n)/\sqrt{\epsilon})$  time  
per pt

Agarwal, Har-Peled, Varadarajan '03

$O((\log^2 n)/\sqrt{\epsilon})$  space & time

Hershberger, Suri [MPDS '03]

(needs bounded ratio  $W/D$ )

Question: const space?



Let's Start Over in 2D...

A Const-Factor Alg'm (Agarwal, Aronov, Sharir '97)

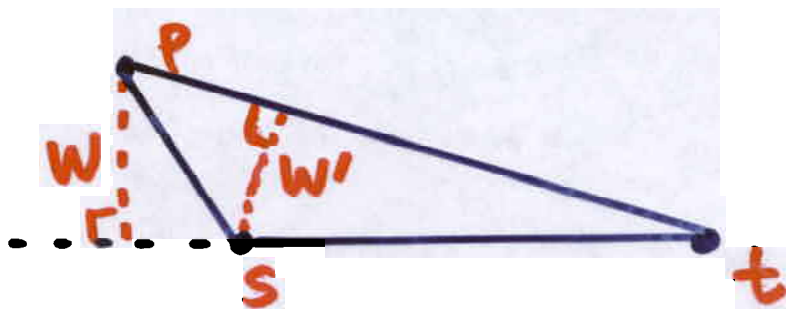
$s$  = any pt

$t$  = farthest neighbor of  $s$

$W$  = farthest distance to  $\overleftrightarrow{st}$

Factor: 4

Pf:  $\text{Width}(P) \leq 2W$



$$W |st| = w' |pt|$$

$$|pt| \leq 2|st| \Rightarrow w' \geq W/2$$

$$\text{Width}(P) \geq \text{Width}(\{s, t, P\})$$

$$\geq W/2$$

□



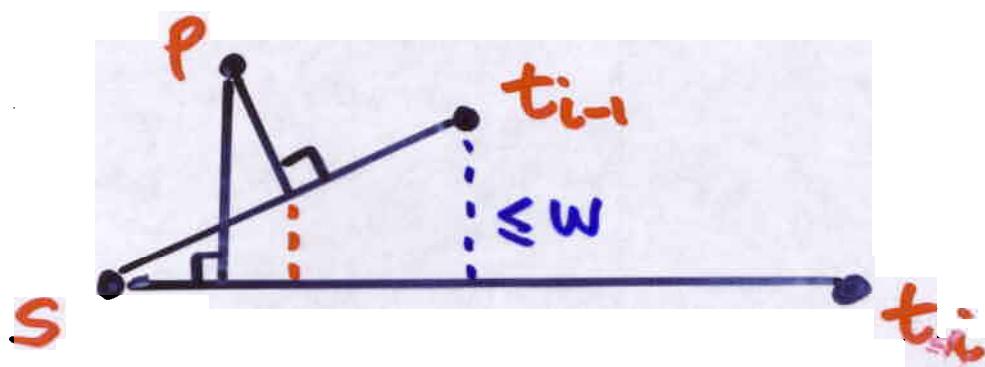
Approx factor:  $O(1)$

Pf:  $\text{Width}(P) \geq W/c$  as before

let  $t_i = i$ th version of  $t$   
since  $p$  is processed

know  $|st_i| \geq 2 |st_{i-1}|$

$|spt| \leq 2 |st_i|$



$$d(p, \overrightarrow{st_i}) - d(p, \overrightarrow{st_{i-1}}) \leq \frac{|spt|}{|st_{i-1}|} W$$

$$\Rightarrow d(p, \overrightarrow{st_\infty}) - d(p, \overrightarrow{st_i}) \leq \left(2 + 1 + \frac{1}{2} + \frac{1}{4} + \dots\right) W$$

$$\Rightarrow \text{Width}(P) \leq c' W$$

□

# A New, One-Pass, Const-Factor Alg'm !

$s, t =$  first two pts

for each  $p \in P$ ,

if  $|sp| \leq 2|st|$

$$W = \max\{W, d(p, \overleftrightarrow{st})\}$$

$$\text{else } W = \max\{W, d(t, \overleftrightarrow{sp})\}$$

$$t = p$$

Space:  $O(1)$

## Refinement & Generalization to $dD$ (skip)

Space (& Time):

$$O\left(\left(\frac{1}{\epsilon} \log \frac{1}{\epsilon}\right)^{d-1}\right)$$

Rmk:

works for the core sets of

Agarwal, Har-Peled, Varadarajan

( $\Rightarrow$  min enclosed ball, annulus, etc.)

## CONCLUSION

diameter & width are good "test cases" for geom. approx. techniques

Open: e.g. best dynamic data structures?