

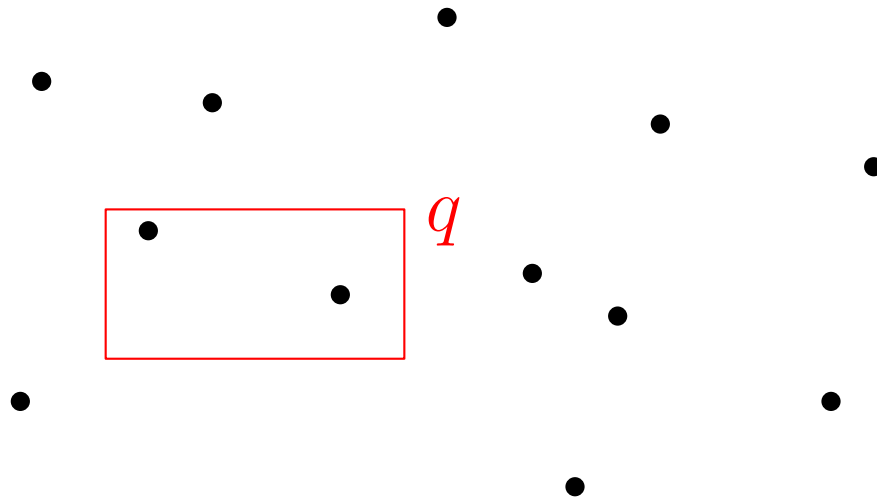
# Computational Geometry, from Low to High Dimensions

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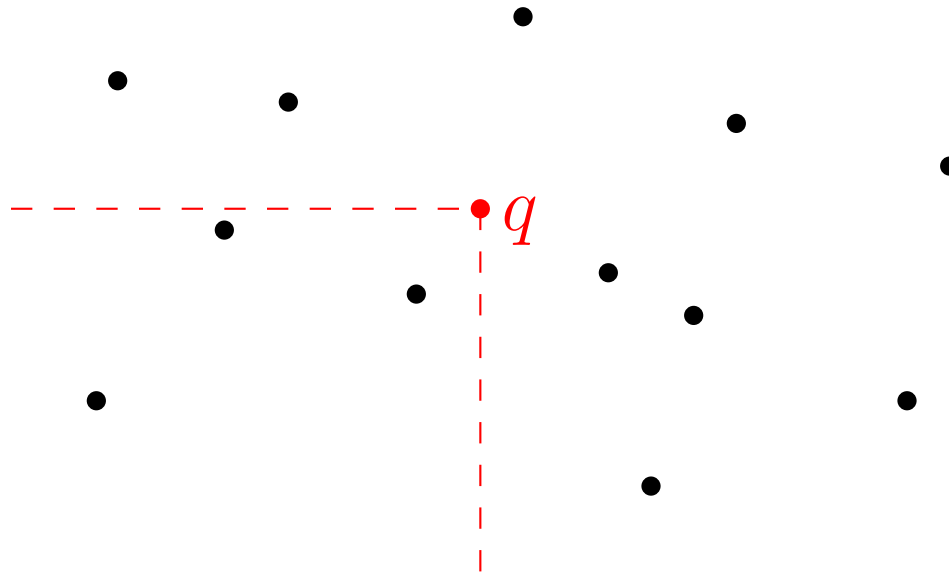
# Classic Problems in CG: Orthogonal Range Search

- preprocess  $n$  points in  $\mathbb{R}^d$ , to detect/report/count points inside any query rectangle  $q$



# Classic Problems in CG: Dominance Range Search

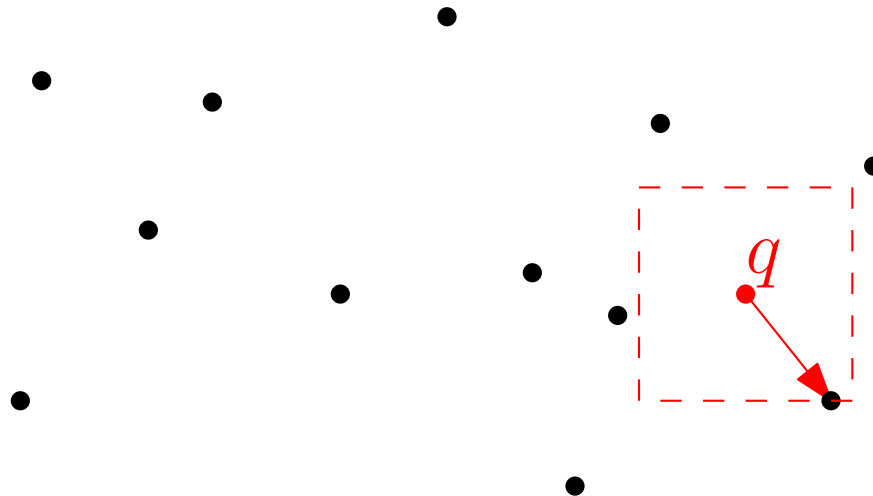
- preprocess  $n$  points in  $\mathbb{R}^d$ , to detect/report/count points dominated by any query point  $q$



(orthogonal range in  $\mathbb{R}^d$  reduces to dominance range in  $\mathbb{R}^{2d}$ )

# Classic Problems in CG: $l_\infty$ Nearest Neighbor Search

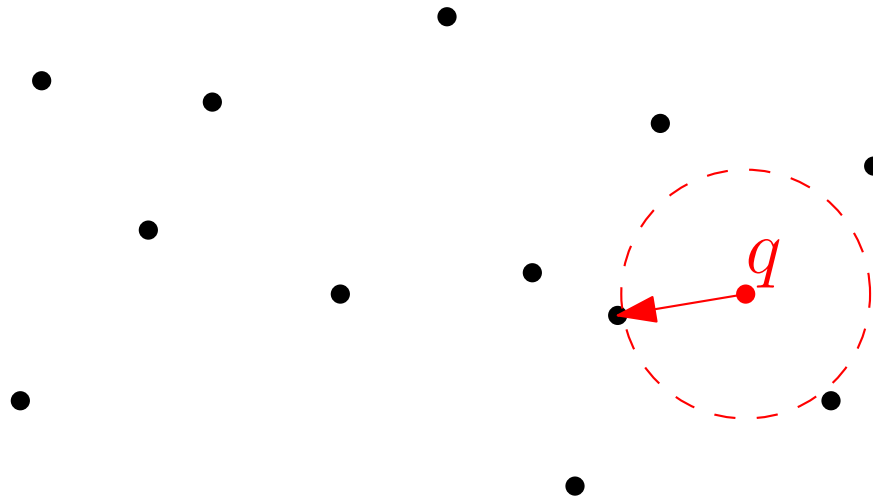
- preprocess  $n$  points in  $\mathbb{R}^d$ , to find nearest point to any query point  $q$



( $l_\infty$  nearest neighbor search reduces to orthogonal range search)

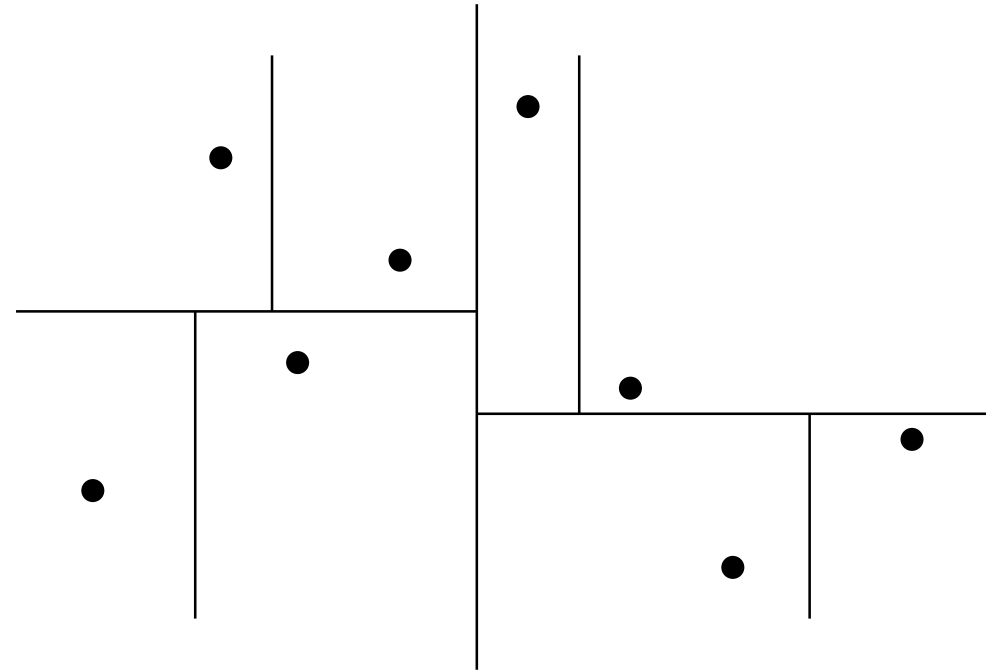
# Classic Problems in CG: $\ell_2$ Nearest Neighbor Search

- preprocess  $n$  points in  $\mathbb{R}^d$ , to find  $\ell_2$  nearest point to any query point  $q$



# Standard Alg'ms in Low Dimensions: k-d Trees [Bentley'75]

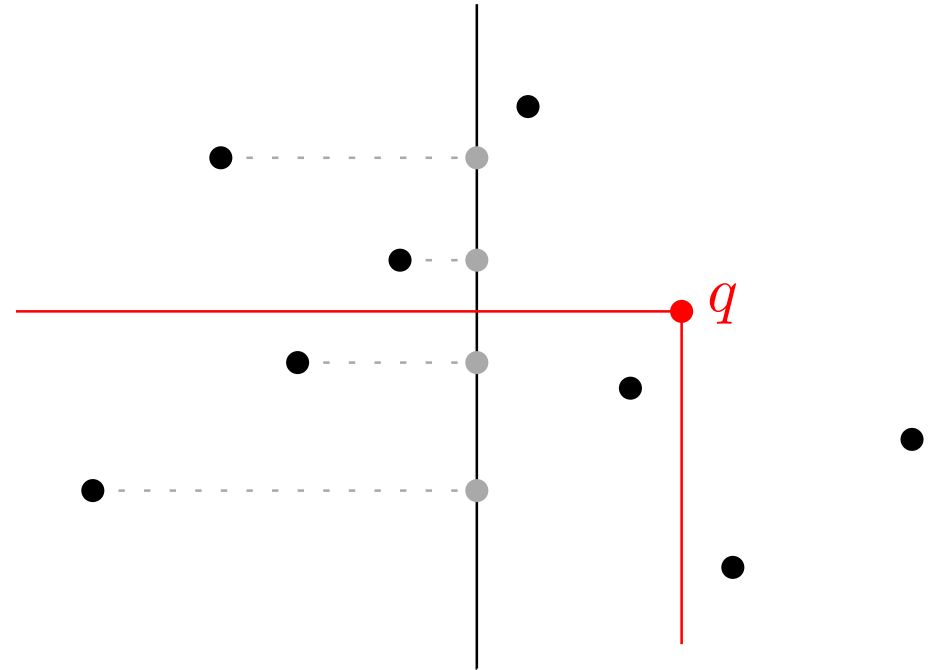
- divide by median x-coord.
- then by median y-coord.
- etc.



⇒  $O(n)$  space,  $O(n^{1-1/d})$  query time

# Standard Alg'ms in Low Dimensions: Range Trees [Lueker'78, Willard'79, Bentley'79, Lee–Wong'80]

- divide by median x-coord.
- recurse on left & right
- recurse on projection



$$\Rightarrow S_d(n) \leq 2S_d(n/2) + S_{d-1}(n)$$

$$Q_d(n) \leq Q_d(n/2) + Q_{d-1}(n)$$

$$\Rightarrow \boxed{O(n \log^d n)} \text{ space, } \boxed{O(\log^d n)} \text{ query time}$$

# Improvements for Orthogonal Range Reporting

	space	query time
Willard'85	$n \log^{d-1} n$	$\log^{d-1} n$
Chazelle [FOCS'83]	$n \log^{d-1} n / \log \log n$	$\log^{d-1} n$
Chazelle [FOCS'85]	$n \log^{d-2+\varepsilon} n$	$\log^{d-1} n$
Willard [SODA'92]	$n \log^{d-1} n / \log \log n$	$\log^{d-1} n / \log \log n$
Subramanian– Ramaswamy [SODA'95]	$n \log^{d-1} n$	$\log^{d-2} n \log^{**} n$
Alstrup–Brodal– Rauhe [FOCS'00]	$n \log^{d-2+\varepsilon} n$	$\log^{d-2} n / \log^{d-3} \log n$
Nekrich [SoCG'07]	$n \log^{d+1+\varepsilon} n$	$\log^{d-3} n / \log^{d-5} \log n$
Afshani'08	$n \log^{d+\varepsilon} n$	$\log^{d-3} n / \log^{d-5} \log n$
Karpinski–Nekrich'09	$n \log^{d-2+\varepsilon} n$	$\log^{d-3} n / \log^{d-6} \log n$
C. [SODA'11]	$n \log^{d-2+\varepsilon} n$	$\log^{d-3} n / \log^{d-5} \log n$
C.–Larsen–Pătraşcu [SoCG'11]	$n \log^{d-2+\varepsilon} n$	$\log^{d-3} n / \log^{d-4} \log n$



# Improvements for Offline Dominance Range Detection

	total time for $n$ queries
Kung–Luccio–Preparata'75	$n \log^{d-2} n$
Gabow–Bentley–Tarjan [STOC'84]	$n \log^{d-3} n \log \log n$
C.–Larsen–Pătraşcu [SoCG'11]	$n \log^{d-3} n$

# Improvements for Offline Dominance Range Counting

Bentley'80

Willard [SODA'02]

C.–Pătraşcu [SODA'10]

total time for  $n$  queries

$$n \log^{d-1} n$$

$$n \log^{d-1} n / \log \log n$$

$$n \log^{d-2+1/d} n$$

# Known Alg'ms in Low Dimensions for $\ell_2$ Nearest Neighbor Search

	space	query time
Dobkin–Lipton [STOC'74]	$n^{2^{d+1}}$	$\log n$
Clarkson [STOC'85]	$n^{\lceil d/2 \rceil + \varepsilon}$	$\log n$
Haussler–Welzl [SoCG'86]	$n$	$n^{1-1/(d(d+1)+1)+\varepsilon}$
Chazelle–Sharir–Welzl [SoCG'90]	$n$	$n^{1-1/(d+1)+\varepsilon}$
Matoušek [SoCG'91]	$n$	$n^{1-1/(d+1)} \text{polylog } n$
Matoušek [FOCS'91]	$n$	$n^{1-1/\lceil d/2 \rceil} \text{polylog } n$

# “Curse of Dimensionality”

- all these alg'ms have **exponential** dependencies in  $d$

e.g., range tree's  $O(\log^d n)$  time is sublinear only for  $d \ll \log n / \log \log n$

[C.'05: range tree still OK for  $d \ll 0.29 \log n$ ]

[Short Proof:

$$Q_d(n) \leq Q_d(n/2) + Q_{d-1}(n) \Rightarrow O\left(\binom{\log n + d}{d}\right)]$$

# In Very High Dimensions...

- Let  $M(n, d, n)$  be time to **multiply**  $n \times d$  &  $d \times n$  **matrix**
- **Folklore**: offline  $\ell_2$  nearest neighbor search in  $O(M(n, d, n))$  time  
e.g.,  $O^*(n^2)$  for  $d \ll n^{0.30}$  [Coppersmith'82 ... Le Gall'12]
- **Matoušek'91**: offline dominance range search in  $O(M(n, ds, n) + dn^2/s)$  time  
e.g.,  $O^*(n^2)$  for  $d \ll n^{0.15}$  (by picking  $s = d$ )

# Focus of This Talk

For what  $d$  can we obtain **subquadratic** **exact** alg'ms?

# Part I: Range Trees Strike Back

(or, How CG can help nongeometric problems...)

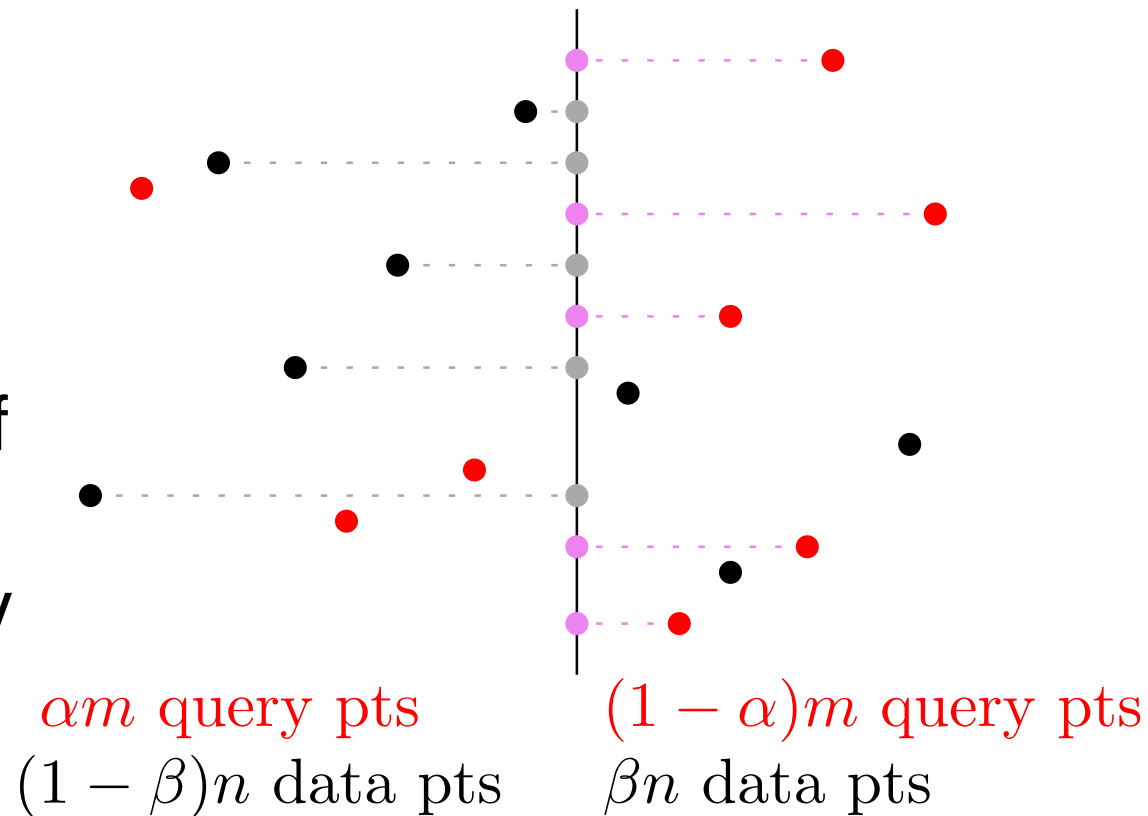
# Surprisingly...

- **range-tree**-like divide&conquer can still work well **beyond log dimensions!**
- **Impagliazzo–Lovett–Paturi–Schneider'14:**  
offline dominance range search in  $n^{2-1/\tilde{O}(c^{15})}$  time  
for  $d = c \log n$
- **C. [SODA'15]:** improves to  $n^{2-1/\tilde{O}(c)}$  time  
(e.g., subquadratic for  $d \ll \log^2 n$ )



# Range-Tree-Like Method for Offline Dominance [Impagliazzo–Lovett–Paturi–Schneider'14]

- divide by some vertical hyperplane
- recurse on left & right
- recurse on projection of left data pts with **projection** of right query pts



$$\Rightarrow T_d(m, n) = T_d(\alpha m, (1 - \beta)n) + T_d((1 - \alpha)m, \beta n) + T_{d-1}((1 - \alpha)m, (1 - \beta)n)$$

- pick dividing hyperplane s.t.  $\alpha = \beta$

# Analysis

[C. (SODA'15)]

$$T_d(m, n) = T_d(\alpha m, (1-\alpha)n) + T_d((1-\alpha)m, \alpha n) \\ + T_{d-1}((1-\alpha)m, (1-\alpha)n)$$

- How to solve recurrence: guess!
- suppose  $T_d(m, n) \leq (1 + \delta)^d (nm^{1-\varepsilon} + mn^{1-\varepsilon})$
- induction goes through by picking  $\delta \approx 1/c^2$ ,  $\varepsilon \approx 1/c$   
 $\Rightarrow \boxed{n^{2-1/\tilde{O}(c)}} \text{ time}$

# Remark: Online Dominance?

- C.'17(?):  $O(n^{1+\varepsilon})$  space/preproc. time,  $n^{1-1/\tilde{O}(c)}$  expected query time

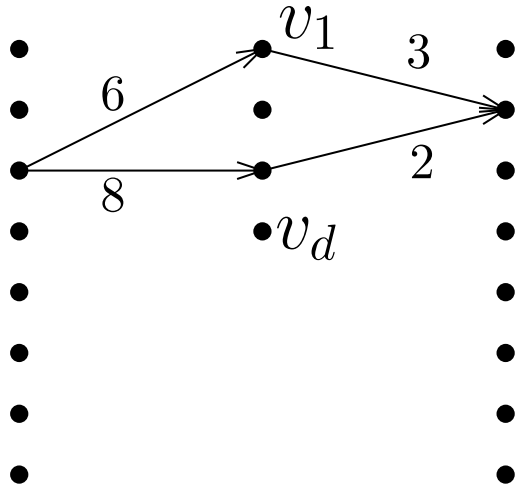
by **k-d-tree**-like divide&conquer (lopsided, randomized, with secondary structures for lower-dimensional projections...)

# Application 1: All-Pairs Shortest Paths (APSP)

- given a real-weighted dense graph with  $n$  vertices, compute shortest path from  $s$  to  $t$  for every pair  $(s, t)$
- **Textbook** [Floyd–Warshall]:  $O(n^3)$  time
- Subcubic?

# From APSP to Dominance [C.'05]

- **Main Case:**  $n \times d \times n$  tripartite graph



**Subproblem:** which shortest paths go through  $v_1$ ?

- same as reporting all pairs  $(s, t)$  s.t.  $\forall k = 1, \dots, d,$   
 $w(s, v_1) + w(v_1, t) \leq w(s, v_k) + w(v_k, t)$ , i.e.,  
 $w(s, v_1) - w(s, v_k) \leq w(v_k, t) - w(v_1, t)$   
 $\Rightarrow$  offline dominance for  $n$  pts in  $d$  dimensions!

- implies APSP alg'm in  $\tilde{O}(n^3 / \log^2 n)$  time
- C.'17(?): combine with **bit packing** tricks  
⇒ APSP alg'm in  $\tilde{O}(n^3 / \log^3 n)$  time (combinatorial)
- extends earlier  $\tilde{O}(n^3 / \log^3 n)$  combinatorial alg'm for **Boolean matrix multiplication** [C. (SODA'15)]

# Application 2: 0-1 Integer Linear Programming (ILP)

- Find an assignment of  $n$  variables  $x_1, \dots, x_n \in \{0, 1\}$  to satisfy  $cn$  given constraints
- Beating brute-force  $2^n$  time?

# From 0-1 ILP to Dominance

[Impagliazzo–Lovett–Paturi–Schneider'14]

- same as finding  $(x_1, \dots, x_{n/2}), (x_{n/2+1}, \dots, x_n) \in \{0, 1\}^{n/2}$   
s.t.  $\forall i = 1, \dots, cn,$

$$a_{i,1}x_1 + \dots + a_{i,n}x_n \leq b_i, \text{ i.e.,}$$

$$a_{i,1}x_1 + \dots + a_{i,n/2}x_{n/2} \leq b_i - a_{i,n/2+1}x_{n/2+1} - \dots - a_{i,n}x_n$$

$\Rightarrow$  offline dominance for  $2^{n/2}$  pts in  $cn$  dimensions!

- implies 0-1 ILP alg'm in  $(2^{n/2})^{2-1/\tilde{O}(c)} =$   
 $2^{(1-1/\tilde{O}(c))n}$  time



# Part II: The Polynomial Method

(or, How nongeometric techniques can help CG...)

# Recent Breakthrough

- Williams [STOC'14]:  
APSP in  $n^3 / 2^{\Omega(\sqrt{\log n})}$  rand. time
- Abboud–Williams–Yu [SODA'15]:  
offline dominance range detection in **Boolean** case in  $n^{2-1/O(\log c)}$  rand. time for  $d = c \log n$   
e.g.,  $n^2 / 2^{\Omega(\sqrt{\log n})}$  for  $d \ll 2^{\sqrt{\log n}}$   
(derandomized by C.–Williams [SODA'16])
- C.'17(?): same for offline dominance in **general** case

# Polynomial Method for Offline Boolean Dominance [Abboud–Williams–Yu (SODA'15)]

- Given group of  $s$  data pts  $\mathbf{x} \in (\{0, 1\}^d)^s$  & a query pt  $y \in \{0, 1\}^d$ , define predicate

$$F(\mathbf{x}, y) = [\text{all } s \text{ pts in } \mathbf{x} \text{ are not dominated by } y]$$

- **Goal:** evaluate  $F$  over all  $n/s$  groups  $\mathbf{x}$  & all  $n$  query pts  $y$
- **Approach:** express  $F$  as a multivariate **polynomial!**

# Polynomial Method for Offline Boolean Dominance [Abboud–Williams–Yu (SODA'15)]

- e.g.,  $F(x, y) = x_1y_2 + 5x_2y_1y_2 + 4x_1x_2y_1$   
 $= (x_1, 5x_2, 4x_1x_2) \cdot (y_2, y_1y_2, y_1)$
- goal reduces to computing dot products between  $n/s$  vectors &  $n$  vectors with  
dimension  $d' = \#$  monomials in  $F$   
i.e., **multiply** an  $n/s \times d'$  with  $d' \times n$  **matrix**  
 $\Rightarrow \tilde{O}(n^2/s)$  time for  $d' \ll (n/s)^{0.1}$  [Coppersmith'82]

# Polynomial Method for Offline Boolean Dominance [Abboud–Williams–Yu (SODA'15)]

- **New Problem:** express

$$\begin{aligned} F(\mathbf{x}, \mathbf{y}) &= [\text{all } s \text{ pts in } \mathbf{x} \text{ are not dominated by } \mathbf{y}] \\ &= \bigwedge_{i=1}^s \bigvee_{j=1}^d (x_{ij} \wedge \overline{y_j}) \end{aligned}$$

as a polynomial with small # monomials

- aim for small degree...

# OR Polynomial [Razborov–Smolensky'87]

- **Subproblem:** express  $\bigvee_{j=1}^d z_j$  as a polynomial
- **Easy Sol'n:**  $1 - \prod_{j=1}^d (1 - z_j)$  (but degree =  $d$ , too big!)

# OR Polynomial [Razborov–Smolensky'87]

- **Subproblem:** express  $\bigvee_{j=1}^d z_j$  as a polynomial
- **Rand. Sol'n:** take random vector  $\mathbf{r} \in \{0, 1\}^d$   
return  $\sum_{j=1}^d r_j z_j \pmod{2}$
- degree = 1!
- 1-sided error prob. = 1/2
- can lower error prob. to  $1/s$  by repeating  $\log s$  times & taking product  $\Rightarrow$  degree  $\approx \log s$

$$F(\mathbf{x}, \mathbf{y}) = \bigwedge_{i=1}^s \bigvee_{j=1}^d (x_{ij} \wedge \overline{y_j})$$

- apply Razborov–Smolensky twice

(for top AND, use de Morgan & const error prob.)

- degree  $\approx \log s$

- # monomials  $\approx s \cdot \binom{d}{\log s}$

$$= s \cdot \binom{c \log n}{\alpha \log n} \quad \text{for } d = c \log n, \quad s = n^\alpha$$

$$\leq (c/\alpha)^{O(\alpha \log n)}$$

$$= n^{O(\alpha \log(c/\alpha))}$$

$$\ll (n/s)^{0.1} \quad \text{for } \alpha \approx \boxed{1/O(\log c)}$$



# Offline Dominance: From Boolean to General [C:17(?)]

- **Idea**: combine with **range-tree** divide&conquer!
- assume all but the first  $j$  coordinates are in  $\{1, \dots, b\}$

$$T_{d,j}(n) = \begin{cases} n^{2-1/O(\log(bc))} & \text{if } j = 0 \\ bT_{d,j}(n/b) + T_{d,j-1}(n) & \text{else} \end{cases}$$

$\Rightarrow$   $n^{2-1/O(\log c)}$  (by **guessing**, picking  $b = c^{O(1)} \dots$ )

# More Developments

- **Alman–Williams [FOCS'15]**: offline nearest/farthest neighbor search in **Hamming** case in  $n^{2-1/\tilde{O}(c)}$  rand. time for  $d = c \log n$
- **Alman–C.–Williams'16(?)**: improves to  $n^{2-1/\tilde{O}(\sqrt{c})}$   
e.g., subquadratic for  $d \ll \log^3 n$

# Polynomial Method for Offline Hamming Farthest Neighbor [Alman–Williams (FOCS'15)]

- again consider group of  $s$  data pts  $\mathbf{x} \in (\{0, 1\}^d)^s$  & a query pt  $y \in \{0, 1\}^d$
- **New Problem:** express

$$\begin{aligned} F(\mathbf{x}, y) &= [\text{one of the } s \text{ pts in } \mathbf{x} \text{ has dist. } > t \text{ from } y] \\ &= \bigvee_{i=1}^s \left[ \sum_{j=1}^d (x_{ij} - y_j)^2 > t^2 \right] \end{aligned}$$

as a polynomial with small # monomials/degree

# Threshold Polynomial: Method 1

[Alman–Williams (FOCS'15)]

- **Subproblem:** express  $\left[ \sum_{j=1}^d z_j > t \right]$  as a polynomial

- **Idea 1:** take rand. sample  $R$  of size  $d/2$

- degree  $D(d) \approx D(d/2) + \sqrt{d \log s}$

↑

interpolating polynomial

$$\approx \sqrt{d \log s} \text{ (optimal)}$$

$$F(\mathbf{x}, \mathbf{y}) = \sum_{i=1}^s \left[ \sum_{j=1}^d (x_{ij} - y_j)^2 > t^2 \right]$$

• degree  $\approx \sqrt{d \log s}$

• # monomials  $\approx s \cdot \binom{d}{\sqrt{d \log s}}$

$$\approx s \cdot \binom{c \log n}{\sqrt{c\alpha} \log n} \quad \text{for } d = c \log n, \quad s = n^\alpha$$

$$\leq (c/\alpha)^{O(\sqrt{c\alpha} \log n)}$$

$$= n^{O(\sqrt{c\alpha} \log(c/\alpha))}$$

$$\ll (n/s)^{0.1} \quad \text{for } \alpha \approx \boxed{1/\tilde{O}(c)}$$

# Threshold Polynomial: Method 2

[Alman–C.–Williams'16(?)]

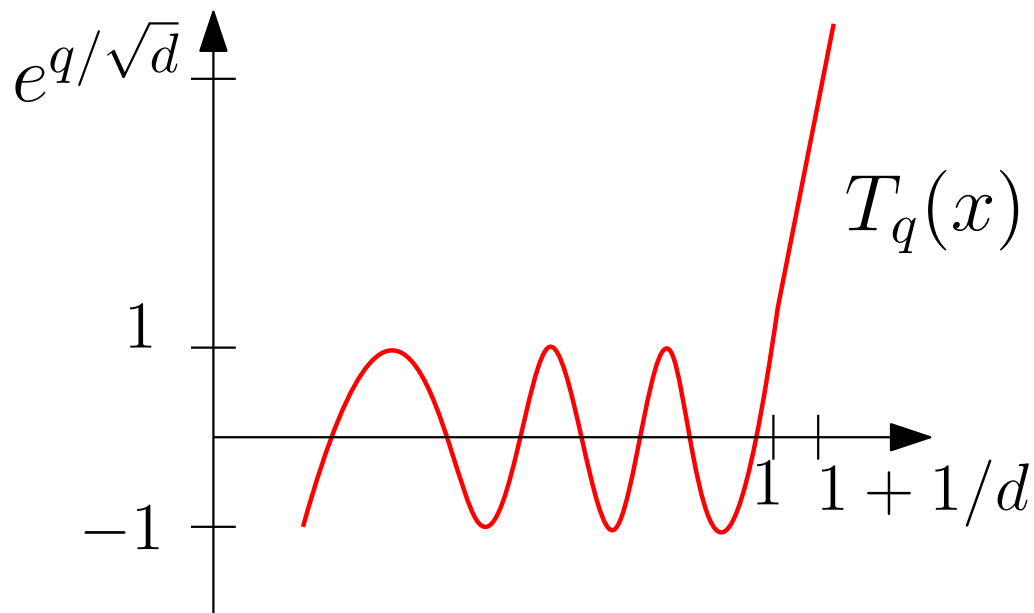
- **Subproblem:** express  $\left[ \sum_{j=1}^d z_j > t \right]$  as a polynomial
- **Idea 2:** just take  $T_q \left( \sum_{j=1}^d z_j / t \right)$ , where  $T_q$  is the degree- $q$  **Chebyshev polynomial (1854)!**

$$\begin{aligned} T_q(x) &= \cos(q \arccos(x)) \\ &= \cosh(q \operatorname{arcosh}(x)) \\ &= \sum_{i=0}^{\lfloor q/2 \rfloor} \binom{q}{2i} (x^2 - 1)^i x^{q-2i} \end{aligned}$$

# Threshold Polynomial: Method 2

[Alman–C.–Williams'16(?)]

- **Subproblem:** express  $\left[ \sum_{j=1}^d z_j > t \right]$  as a polynomial
- **Idea 2:** just take  $T_q \left( \sum_{j=1}^d z_j / t \right)$ , where  $T_q$  is the **degree- $q$  Chebyshev polynomial (1854)!**



false  $\Rightarrow$  output  $\in [-1, 1]$

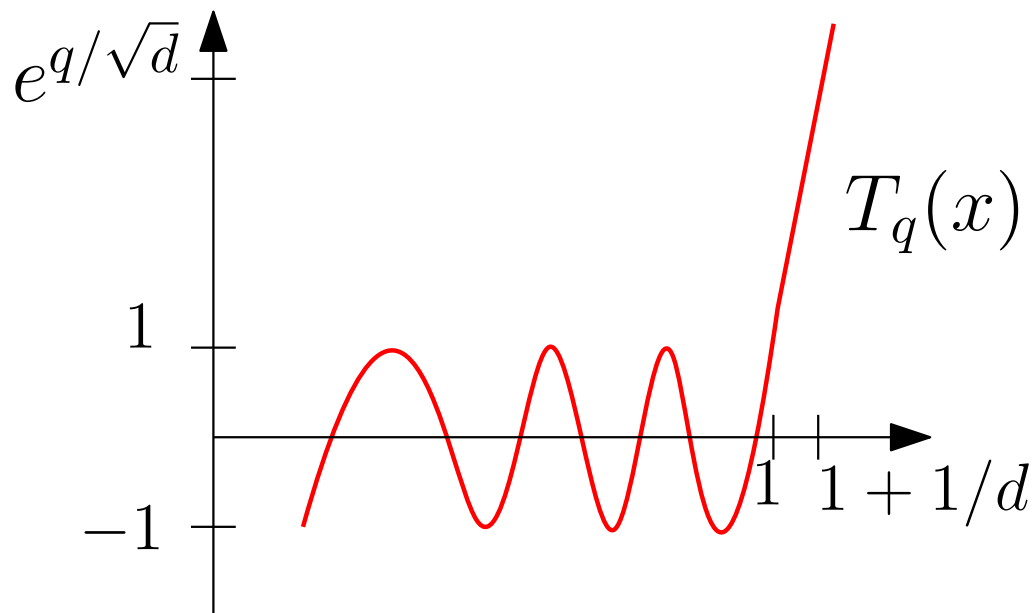
true  $\Rightarrow$  output  $> 100s$

by picking  $q \approx \sqrt{d} \log s$

# Threshold Polynomial: Method 2

[Alman–C.–Williams’16(?)]

- **Subproblem:** express  $\left[ \sum_{j=1}^d z_j > t \right]$  as a polynomial
- **Idea 2:** just take  $T_q \left( \sum_{j=1}^d z_j / t \right)$ , where  $T_q$  is the degree- $q$  “discrete” Chebyshev polynomial (1864)!



false  $\Rightarrow$  output  $\in [-1, 1]$

true  $\Rightarrow$  output  $> 100s$

by picking  $q \approx \sqrt{d \log s}$

(no improvement, sadly...)





$$F(\mathbf{x}, \mathbf{y}) = \sum_{i=1}^s \left[ \sum_{j=1}^d (x_{ij} - y_j)^2 > t^2 \right]$$

• degree  $\approx d^{1/3} \log^{2/3} s$

• # monomials  $\approx s \cdot \binom{d}{d^{1/3} \log^{2/3} s}$

$$= s \cdot \binom{c \log n}{c^{1/3} \alpha^{2/3} \log n} \quad \text{for } d = c \log n, \quad s = n^\alpha$$

$$\leq (c/\alpha) O(c^{1/3} \alpha^{2/3} \log n)$$

$$= n^{O(c^{1/3} \alpha^{2/3} \log(c/\alpha))}$$

$$\ll (n/s)^{0.1} \quad \text{for } \alpha \approx \boxed{1/\tilde{O}(\sqrt{c})}$$

# Remark: Offline **Approximate** Nearest Neighbor Search

- **LSH** [Indyk–Motwani'98, Andoni–Indyk'06]:  
 $\tilde{O}(n^{1+1/(1+\varepsilon)^2})$  rand. time
- **Data-dependent LSH** [Andoni–Indyk–Nguyen–Razenshteyn'14, Andoni–Razenshteyn'15]:  
 $\tilde{O}(n^{1+1/(2(1+\varepsilon)^2-1)}) = n^{2-\Omega(\varepsilon)}$
- G. Valiant [FOCS'12]:  $n^{2-\Omega(\sqrt{\varepsilon})}$
- Alman–C.–Williams'16(?):  $n^{2-\tilde{\Omega}(\varepsilon^{1/3})}$

# Final Remarks

- further consequences [Alman–C.–Williams'16(?)]:
  - faster exponential alg'ms for **MAX-SAT** with  $cn$  constraints
  - **circuit lower bounds** (for depth-2 threshold circuits. . .)
- many open problems:
  - **online** nearest neighbor search, for **general**  $\ell_2$ ?
  - on threshold polynomials:  $o(d^{1/3})$  degree?
  - on range-tree-like methods: lower bounds?

THE END