

Low-Distortion Embeddings II

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Lecture 15: Low-Distortion
Embeddings II

In the previous lecture

- Definition of embedding $f:M \rightarrow M'$ with distortion c
- Isometric embedding of l_1^d into $l_\infty^{2^d}$
 - $l_\infty^{d'}$ diameter in $O(nd')$ time
 - l_1^d diameter in $O(n2^d)$ time
- Embedding of $M=(X,D)$ into l_∞^d
 - Isometry for $d=|X|$
 - Distortion $O(c)$ for $d=|X|^{1/c}$

Today

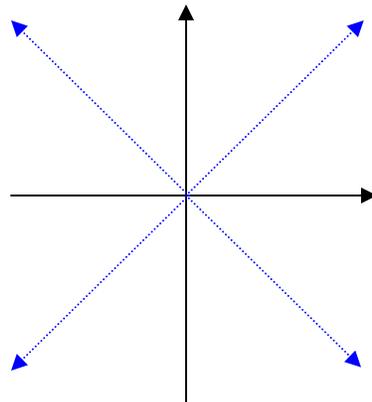
- $(1+\varepsilon)$ -distortion embedding of l_2 into l_∞
 - Approximate diameter in l_2
- $(1+\varepsilon)$ -distortion embedding of (X, l_2) , X in \mathbb{R}^d , into $l_2^{d'}$, where $d' = O(\log |X| / \varepsilon^2)$
 - $(1+\varepsilon)$ -approximate Near Neighbor in l_2^d
 - Query time: $O(d \log n / \varepsilon^2)$
 - Space: $n^{O(\log(1/\varepsilon) / \varepsilon^2)}$
- $(1+\varepsilon)$ -distortion embedding of l_2^d into $l_1^{d'}$, where $d' = O(d / \varepsilon^2 \log(1/\varepsilon))$

$(1+\varepsilon)$ -embedding of l_2 into l_∞

- We know:
 - $(1+\varepsilon)$ -embedding of l_2^d into $l_1^{O(d/\varepsilon^2 \log(1/\varepsilon))}$
 - Isometry of l_1^d into $l_\infty^{2^d}$
 - Therefore: $(1+\varepsilon)$ -embedding of l_2^d into $l_\infty^{d'}$,
where $d' = 2^{O(d/\varepsilon^2 \log(1/\varepsilon))}$
- We will improve d' to $O(1/\varepsilon)^{(d-1)/2}$

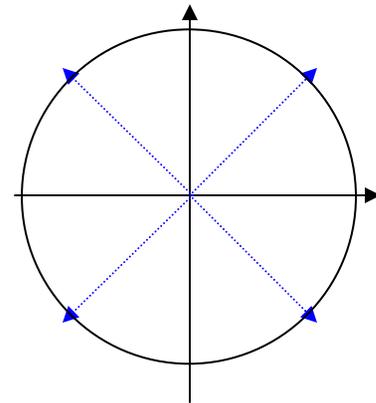
Consider $d=2$

- For embedding into l_1 we used $f(x,y)=[x+y,x-y,-x+y,-x-y]$
 - Since f linear, we have $\|f(p)-f(q)\|=\|f(p-q)\|$
 - $\|(x,y)\|_1 = |x|+|y| = \max[x+y , x-y , -x+y , -x-y]$



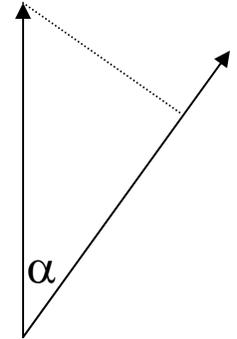
Embedding of l_2

- Again, use projections
 - Onto unit (l_2) vectors $v_1 \dots v_k$
 - Requirement: vectors are “densely” spaced
 - I.e., for any u there is v_i such that $u^*v_i \geq \|u\|_2 / (1+\epsilon)$
 - Can assume $\|u\|_2=1$
- How big is k ?



Lemma

- Consider two unit vectors u and v , such that the $\text{angle}(u,v)=\alpha$.
Then $u \cdot v \geq 1 - O(\alpha^2)$
- Proof: $u \cdot v = \cos(\alpha)$
 $\approx 1 + \alpha \cos'(\alpha) + \alpha^2 \cos''(\alpha)/2$
 $\approx 1 - \Theta(\alpha^2)$
- Therefore, suffices to use $2\pi/\varepsilon^{1/2}$ vectors to get distortion $1 + O(\varepsilon)$



Higher Dimensions

- For $d=2$ we get $d'=O(1/\varepsilon^{1/2})$
- For any d we get $d'=O_d(1/\varepsilon)^{(d-1)/2}$
 - Can “cover” a unit sphere in \mathbb{R}^d with $O_d(1/\alpha)^{d-1}$ vectors so that any v has angle $<\alpha$ with at least one of the vectors
 - The remainder is the same
- Yields an $O_d(1/\varepsilon)^{(d-1)/2}n$ – time algorithm for approximate diameter in l_2

Covering vs Packing

- Assume we want to **cover** the sphere using disks of radius α
- This can be achieved by **packing**, as many as possible, disks of radius $\alpha/2$
- How many disks can be pack ?
 - Each disk has volume $\Theta_d(\alpha/2)^{d-1}$ times smaller than the volume of the sphere
 - Inverse of that gives the packing/covering bound

Dimensionality Reduction

[Johnson-Lindenstrauss'85]: For any X in \mathbb{R}^d , $|X|=n$, there is a $(1+\varepsilon)$ -distortion embedding of (X, l_2) , into $l_2^{d'}$, where $d' = O(\log n / \varepsilon^2)$

Proof

- Need to show that for any pair p, q in X , we have $\|f(p)-f(q)\| \approx S \|p-q\|$
 - $X \approx C$ means $X=(1 \pm \varepsilon)C$
- Our mapping: $f(u)=Au$, A “random”
- Sufficient to show that for a *fixed* $u=p-q$, where p, q in X , the prob. that $\|Au\| \approx S\|u\|$ does **not** hold is at most $1/n^2$
 - Because #pairs $\{p, q\}$, times $1/n^2$, is at most $1/2$
- In fact, by linearity of A we can assume $\|u\|=1$, so we just need to show $\|Au\| \approx S$

Normal Distribution

- Normal distribution:
 - Range: $(-\infty, \infty)$
 - Density: $f(x) = e^{-x^2/2} / (2\pi)^{1/2}$
 - Mean=0, Variance=1
- Basic facts:
 - If X and Y independent r.v. with normal distribution, then $X+Y$ has normal distribution
 - $\text{Var}(cX) = c^2 \text{Var}(X)$
 - If X, Y independent, then $\text{Var}(X+Y) = \text{Var}(X) + \text{Var}(Y)$

Back to embedding

- We map $f(u)=Au=[a^1*u,\dots,a^d*u]$, where each entry of A has normal distribution
- Consider $Z=a^i*u = a^i*u=\sum_j a_{ij} u_j$
- Each term $a_{ij} u_j$
 - Has normal distribution
 - With variance u_j^2
- Thus, Z has normal distribution with variance $\sum_j u_j^2 = 1$
- This holds for each a^j

What is $\|Au\|_2$

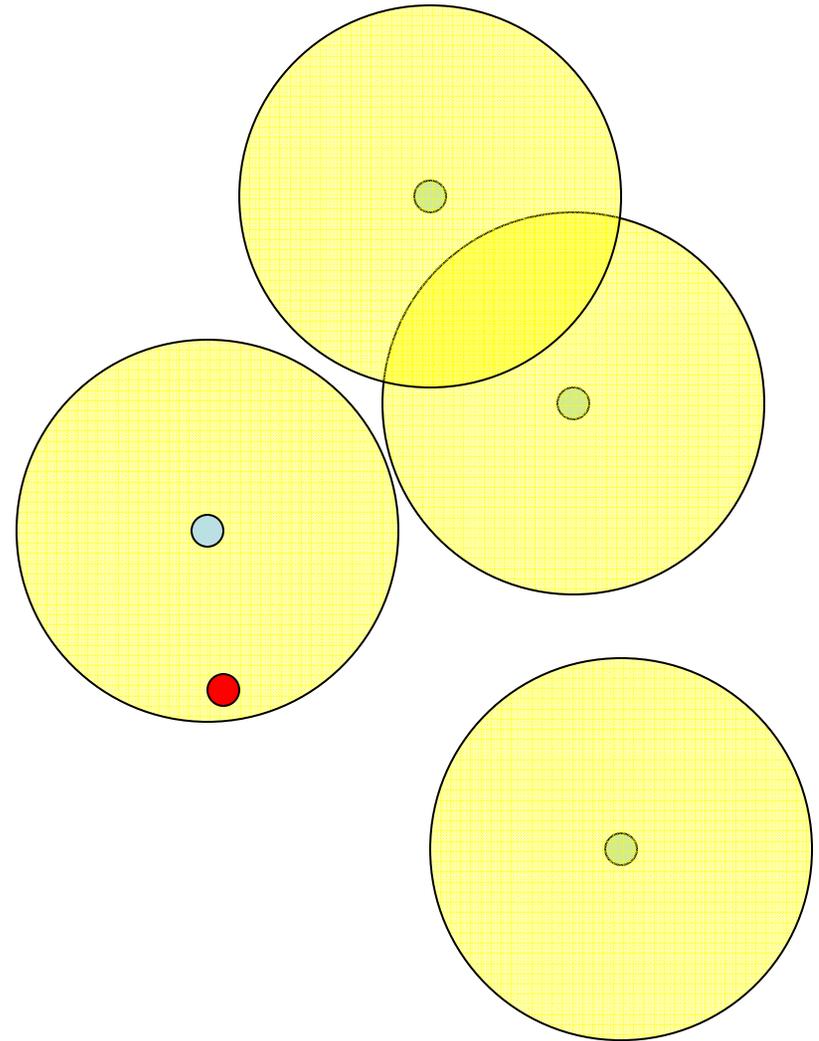
- $\|Au\|^2 = (a^1 * u)^2 + \dots + (a^{d'} * u)^2 = Z_1^2 + \dots + Z_{d'}^2$
where:
 - All Z_i 's are independent
 - Each has normal distribution with variance=1
- Therefore, $E[\|Au\|^2] = d' * E[Z_1^2] = d'$
- By Chernoff-like bound
$$\Pr[|\|Au\|^2 - d'| > \epsilon d'] < e^{-B d' \epsilon^2} < 1/n^2$$
for some constant B
- So, $\|Au\|_2 \approx (d')^{1/2}$ with probability $1 - 1/n^2$

Application to Near Neighbor

- Suppose we have an algorithm with:
 - $O(d)$ query time
 - $O(1/\varepsilon)^d$ n space
- Then we get:
 - $O(d \log n / \varepsilon^2)$ query time
 - $n^{O(\log(1/\varepsilon)/\varepsilon^2)}$ space

$O(1/\epsilon)^d$ n space NN

- Assume $r=1$



Grid

- Impose a grid with side length= $\varepsilon/d^{1/2}$
- Parameters:
 - Cell diameter: ε
 - #cells/ball: $O(1/\varepsilon)^d$
- Store all cells touching a ball

