Advanced Algorithms

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Exercise 08

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1 ℓ_1 diameter in small dimension

The diameter of a pointset P is defined as $\operatorname{diam}(P) = \max_{u,v \in P} d(u,v)$. The goal of the exercise is to show an algorithm for *exact* computation of diameter in ℓ_1^d metric, when d is small (i.e. $o(\log n)$).

- 1. Show an exact embedding $\phi: \ell_1^d \to \ell_\infty^{2^d}$.
- 2. Show an algorithm that finds ℓ_1^d diameter of poinset P, |P| = n, in $\mathcal{O}(n \cdot 2^d)$ time.

2 Johnson-Lindenstrauss

Consider following useful result by Johnson–Lindenstrauss:

Lemma 1 For $P \subset \mathbb{R}^n$, |P| = n, and for $0 < \varepsilon \leq 1$, there is $k = \mathcal{O}(\log n/\varepsilon^2)$ and map $\varphi : \ell_2^n \to \ell_2^k$, such that:

 $\forall_{p,q \in P} \quad \|p - q\|_2 \le \|\phi(p) - \phi(q)\|_2 \le (1 + \varepsilon) \cdot \|p - q\|_2$

More surprisingly, such mapping can be easily constructed. It can be shown, that for appropriate k, and picked u.a.r. matrix $A \in \{-1, 1\}^{k \times n}$, a map $\phi(p) = A \cdot p$ works (w.h.p., and up to some fixed scaling factor).

1. Show that random matrix A preserves squared norm in expectation, that is

$$\mathbb{E}[\|\phi(p)\|_{2}^{2}] = k\|p\|_{2}^{2}$$

where the expectation is taken across choices of A.

2. For the harder part, that is *concentration*: weaker bound (i.e. $\mathcal{O}(\text{poly}(\log n)/\varepsilon^2))$ can be derived from known concentration bounds OR

find and read the proof of online.

3 Sketching for Hamming distance

Consider a following sketching problem: Alice and Bob hold each text, i.e. $T_A, T_B \in \Sigma^n$. They want to send relatively short messages (that is, *sketches*) to Charlie, so that Charlie can compute *Hamming distance* of T_A and T_B .

- 1. Consider case of $\Sigma = \{0, 1\}$. Show that Alice and Bob can encode messages of $\mathcal{O}(\log n/\varepsilon^2)$ words so that Charlie decodes $1 + \varepsilon$ approximation of Hamming distance.
- 2. Consider random projection $\varphi : \Sigma \to \{0, 1\}$. Show that it preserves Hamming distance in expectation (up to factor 1/2).
- 3. Show that averaging over many projections $\varphi_1, \varphi_2, \ldots, \varphi_k$ for some $k = \mathcal{O}(\log n/\varepsilon^2)$ preserves Hamming distance up to $1 + \varepsilon$ factor, w.h.p.

- 4. Combine all of above ideas to obtain a communication protocol:
 - Alice and Bob send messages of length $\mathcal{O}(\log^2 n/\varepsilon^4)$ words,
 - Charlie can decode, with high probability, an $(1+\varepsilon)$ approximation of the Hamming distance.
- 5. (*) Show that by smarter combining of J-L lemma and random projections, one can reduce the size of messages to $\mathcal{O}(\log n/\varepsilon^2)$ words.

4 Applications of the sparsest cut

4.1 Approximating edge expansion

For undirected, unweighted graph G and set $S \subset V$ s.t. $|S| \leq |V|/2$, the *edge expansion* of S is defined to be $|\delta S|/|S|$. Show an $\mathcal{O}(\log n)$ approximation algorithm for finding *minimal expansion* in graph G.

4.2 Minimum bisection cut

A bisection cut is a cut (S, S') such that |S| = |S'| = n/2. A r-balanced cut is a cut where $r \cdot n \leq |S| \leq (1 - r) \cdot n$. Our goal is to (pseudo) approximate minimal bisection cut. Consider following procedure:

- Initialize $U \leftarrow \emptyset, W \leftarrow V$.
- Until |U| < n/3, repeat:
- X is the approximate small edge expansion cut in G_W
- $U \leftarrow U \cup X, W \leftarrow W \setminus X.$
- return (U, W)

Show, that this procedure returns an 1/3-balanced cut, with size being $\mathcal{O}(\log n)$ factor from minimal bisection cut.