

Lecture 18

In today's lecture, we present some of the results contained in [1, 2].

1 Good news: A general theorem on embeddings with slack

We will show how to obtain embeddings into ℓ_p with small dimension where all but an ϵ -fraction of the embedded distances have small distortion. This result has various applications in networking. The embedding is achieved by randomly choosing a small set of beacons, constructing a good embedding for these nodes (using, e.g., Bourgain's embedding), and then extending this embedding to the remaining nodes. Thus, we need the following definition:

Definition 1. A family \mathcal{X} of metrics is subset-closed if any metric in \mathcal{X} restricted to any subset of nodes is also in \mathcal{X} .

Definition 2. Given ϵ , an embedding $\varphi : V \rightarrow V'$ has distortion D with ϵ -slack if a set of all but an ϵ -fraction of edges has distortion at most D under φ .

Theorem 1. Consider a fixed space ℓ_p , $p \geq 1$. Let \mathcal{X} be a subset-closed family of finite metric spaces such that for any $n \geq 1$ and any n -point metric space $X \in \mathcal{X}$ there exists an embedding $\varphi_X : X \hookrightarrow \ell_p$ with distortion $\alpha(n)$ and dimension $\beta(n)$.

Then there exists a universal constant $C > 0$ such that for any metric space $X \in \mathcal{X}$ and any $\epsilon > 0$ we have an embedding into ℓ_p with ϵ -slack, distortion $\alpha(\frac{C}{\epsilon} \log \frac{1}{\epsilon})$ and dimension $\beta(\frac{C}{\epsilon} \log \frac{1}{\epsilon}) + C \log \frac{1}{\epsilon}$.

A (modified) Bourgain's theorem gives $\alpha(n) = O(\log n)$ and $\beta(n) = O(\log^2 n)$. Using an improved result by Bartal gives the following corollary.

Corollary 2. Any metric space $X \in \mathcal{X}$ can be embedded into ℓ_p with ϵ -slack, distortion $O(\log \frac{1}{\epsilon})$ and dimension $O(\log \frac{1}{\epsilon})$.

Proof. Consider some metric $X = (V, d) \in \mathcal{X}$, where V is a set of n nodes. Given $\epsilon > 0$ let $\hat{\epsilon} = \epsilon/20$, and $t = 100 \log \frac{1}{\hat{\epsilon}}$. The first step is to sample $\frac{t}{\hat{\epsilon}}$ beacons, B , uniformly at random from V .

Let g be a contracting embedding from B into ℓ_p with distortion $\alpha(\frac{t}{\hat{\epsilon}})$ and dimension $\beta(\frac{t}{\hat{\epsilon}})$. We will extend g to all of V by defining, $\forall u \in V$, $f(u) = g(b)$, where $b \in B$ is the beacon closest to node u . Also, let

$$\{\sigma_j(u) | u \in V, 1 \leq j \leq t\} \tag{1}$$

be independently sampled $\{0, 1\}$ -valued Bernoulli random variables. For all $u \in V$ and $j = 1, \dots, t$, define the function

$$h_j(u) = \sigma_j(u) \rho_u(\hat{\epsilon}) t^{-1/p}. \tag{2}$$

Definition 3. Let $\rho_u(\epsilon)$ be the radius of the smallest ball around u that contains at least ϵn nodes.

The embedding $\varphi(u)$ will be the concatenation of $f(u)$ and $(h_1(u), \dots, h_t(u))$, giving a dimension of $\beta(\frac{t}{\hat{\epsilon}}) + t$.

Let E be the set of all unordered node pairs. We will remove from consideration three sets D_1 , D_2 , and D_3 of “difficult” node pairs, whose distortion we will not bound, and will show that they are of size only $O(\epsilon)$.

$$D_1 = \{(u, v) | d(u, v) < \max\{\rho_u(\hat{\epsilon}), \rho_v(\hat{\epsilon})\}\} \quad (3)$$

$$D_2 = \{(u, v) | d(u, B) \geq \rho_u(\hat{\epsilon}) \vee d(v, B) \geq \rho_v(\hat{\epsilon})\}. \quad (4)$$

At an intuitive level, D_1 is the set of node pairs that are too close together, and D_2 is the set of node pairs that are too far from the beacons. The third set, D_3 , will be given later. Note that $|D_1| \leq \hat{\epsilon}n^2$, since for any node $u \in V$ there are at most $\hat{\epsilon}n$ nodes in the ball of radius $\rho_u(\hat{\epsilon})$ around u . Also, for any node $v \in V$ we have:

$$\Pr[d(u, B) \geq \rho_u(\hat{\epsilon})] \leq \Pr[\forall v \in B_{\rho_u(\hat{\epsilon})}(u), v \notin B] \leq \left(1 - \left(\frac{t}{\hat{\epsilon}}\right) \frac{1}{n}\right)^{\hat{\epsilon}n} \leq e^{-t} \leq \hat{\epsilon}, \quad (5)$$

so $\mathbb{E}[|D_2|] \leq \hat{\epsilon}n^2$ and by Markov’s inequality we have $|D_2| \leq 2\hat{\epsilon}n^2$ with probability at least $1/2$.

1.1 Upper Bound on $d(\varphi(u), \varphi(v))$

Let $G' = E \setminus (D_1 \cup D_2)$. We will now upper bound $d(\varphi(u), \varphi(v))$ for all $(u, v) \in G'$. Since $G' \subseteq \overline{D_1}$, $d(u, v) > \rho_u(\hat{\epsilon})$ and $d(u, v) > \rho_v(\hat{\epsilon})$. Since $G' \subseteq \overline{D_2}$, $d(u, B) < \rho_u(\hat{\epsilon})$ and $d(v, B) < \rho_v(\hat{\epsilon})$. Together these imply that $d(u, B) < d(u, v)$ and $d(v, B) < d(u, v)$. We thus have:

$$\begin{aligned} \|\varphi(u) - \varphi(v)\|_p^p &= \|f(u) - f(v)\|_p^p + \sum_{j=1}^t |h_j(u) - h_j(v)|^p \\ &= \|g(b_u) - g(b_v)\|_p^p + \sum_{j=1}^t |h_j(u) - h_j(v)|^p \\ &\leq (d(b_u, b_v))^p + \sum_{j=1}^t |h_j(u) - h_j(v)|^p \\ &\leq (d(b_u, u) + d(u, v) + d(v, b_v))^p + \sum_{j=1}^t |h_j(u) - h_j(v)|^p \\ &< (3d(u, v))^p + \sum_{j=1}^t |t^{-1/p} \max\{\rho_u(\hat{\epsilon}), \rho_v(\hat{\epsilon})\} - 0|^p \\ &< (3^p + 1)(d(u, v))^p. \end{aligned} \quad (6)$$

Where we used the fact that g is a contractive embedding, and the notation b_u refers to the node in B which is closest to u .

All that remains is to give a lower bound on $d(\varphi(u), \varphi(v))$. We consider two separate cases:

$$G_1 = \{(u, v) \in G' : \max\{\rho_u(\hat{\epsilon}), \rho_v(\hat{\epsilon})\} \geq d(u, v)/4\} \quad (7)$$

$$G_2 = G' \setminus G_1. \quad (8)$$

1.2 Lower Bound on $d(\varphi(u), \varphi(v))$ for G_1

Since for all edges in G_1 either $\rho_u(\hat{\epsilon})$ or $\rho_v(\hat{\epsilon})$ is $\Omega(d(u, v))$, the h_j coordinates will provide the desired lower bound. The σ_j random variables help ensure that in some non-negligible fraction of the dimensions we get $\rho_u(\hat{\epsilon}) + \rho_v(\hat{\epsilon})$. More concretely, consider an edge $(u, v) \in G_1$, and without loss of generality assume $\rho_u(\hat{\epsilon}) \geq \rho_v(\hat{\epsilon})$, so that $\rho_u(\hat{\epsilon}) \geq d(u, v)/4$.

Let $\mathcal{E}_j(u, v)$ be the event that $\sigma_j(v) = 0$ and $\sigma_j(v) = 1$. This event happens with probability $\frac{1}{4}$. Let $A(u, v) = \sum_{j=1}^t \mathbf{1}_{\mathcal{E}_j(u, v)}$. Then $\mathbb{E}[A(u, v)] = t/4$, and by Chernoff's bound we have:

$$\Pr \left[A(u, v) \leq \frac{\mathbb{E}[A(u, v)]}{2} \right] \leq e^{-t/50} \leq \hat{\epsilon}. \quad (9)$$

We now define D_3 , the third set to be removed:

$$D_3 = \{(u, v) \in G_1 \mid A(u, v) \leq t/8\} \quad (10)$$

Note that there is nothing inherently difficult about the edges in D_3 – we were just unlucky with the coin tosses. By Markov's inequality, $|D_3| \leq 2\hat{\epsilon}n^2$ with probability $\geq 1/2$.

We now get the desired lower bound for $(u, v) \in G_1 \setminus D_3$:

$$\begin{aligned} \|\varphi(u) - \varphi(v)\|_p^p &\geq \sum_{j=1}^t |h_j(u) - h_j(v)|^p \\ &= \sum_{j=1}^t |\sigma_j(u)\rho_u(\hat{\epsilon})t^{-1/p} - \sigma_j(v)\rho_v(\hat{\epsilon})t^{-1/p}|^p \\ &\geq \frac{t}{8} \left(\rho_u(\hat{\epsilon})t^{-1/p} \right)^p \geq \frac{1}{8} \left(\frac{1}{4}d(u, v) \right)^p. \end{aligned} \quad (11)$$

1.3 Lower Bound on $d(\varphi(u), \varphi(v))$ for G_2

Recall that b_u refers to the beacon in B which is closest to u . Since $G_2 \subseteq \overline{D_2}$, $d(u, B) < \rho_u(\hat{\epsilon})$ and $d(v, B) < \rho_v(\hat{\epsilon})$. Since $G_2 \subseteq \overline{G_1}$, $d(u, v)/4 > \rho_u(\hat{\epsilon})$ and $d(u, v)/4 > \rho_v(\hat{\epsilon})$. Together these imply that $d(u, B) < d(u, v)/4$ and $d(v, B) < d(u, v)/4$. Applying the triangle inequality,

$$\begin{aligned} d(b_u, b_v) &\geq d(u, v) - d(u, b_u) - d(v, b_v) \\ &> d(u, v) - d(u, v)/4 - d(u, v)/4 = d(u, v)/2. \end{aligned} \quad (12)$$

We thus have the following lower bound for $(u, v) \in G_2$:

$$\begin{aligned} \|\varphi(u) - \varphi(v)\|_p^p &\geq \|f(u) - f(v)\|_p^p \\ &= \|g(b_u) - g(b_v)\|_p^p \\ &\geq \frac{1}{\alpha(\frac{t}{\hat{\epsilon}})} \cdot d(b_u, b_v) \\ &\geq \frac{d(u, v)}{2\alpha(\frac{t}{\hat{\epsilon}})}. \end{aligned} \quad (13)$$

To finish the proof, note that D_2 and D_3 are independent (D_3 is a function of the σ_j random variables). Thus, in total from D_1, D_2 , and D_3 we have removed at most $5\hat{\epsilon}n^2$ nodes with probability at least $\frac{1}{4}$. Let $G = E \setminus (D_1 \cup D_2 \cup D_3)$ be the set of edges whose distortion we bounded in the previous two sections. We conclude that, with probability $\geq \frac{1}{4}$,

$$|G| \geq \binom{n}{2} - 5\hat{\epsilon}n^2 \geq \binom{n}{2} - \frac{\epsilon n^2}{4} \geq (1 - \epsilon) \binom{n}{2}. \quad (14)$$

□

2 Bad news: Lower bounds for embedding with slack

2.1 Why distortion $\Omega(\log \frac{1}{\epsilon})$ is necessary

Theorem 3. *There exists a finite metric (X, d) on arbitrary many nodes that requires distortion $\Omega(\frac{1}{p} \log \frac{1}{\epsilon})$ for embedding with ϵ -slack into ℓ_p , where $p \geq 1$.*

Proof. Suppose that for every finite metric, there exists an embedding into ℓ_p with ϵ -slack and distortion D . We will show that $D = \Omega(\frac{1}{p} \log \frac{1}{\epsilon})$.

Suppose without loss of generality that $0 < \epsilon \leq 1/4$. Let k be $1/(2\sqrt{\epsilon})$. Let $G = (V, E)$ be a constant-degree expander on k nodes, and (V, dist_G) be the corresponding shortest-path metric. For each node v in V , create a new path on $n/k = 2\sqrt{\epsilon}n$ nodes, and attach one of its ends to v . This way we get a graph on n nodes. Let δ such that $\delta \cdot D \leq 1/3$ be the length of each attached path. The new weighted graph $H = (V', E')$ induces the shortest-path metric (V', dist_H) .

There exists an embedding ϕ of (V', dist_H) into ℓ_p with ϵ -slack and distortion D . We can assume without loss of generality that

$$\text{dist}_H(v, w) \leq \|\phi(v) - \phi(w)\|_p \leq D \cdot \text{dist}_H(v, w) \quad (15)$$

for all but an ϵ -fraction of pairs v and w . Let I be the set of the pairs for which the above inequality does not hold. We have $|I| \leq \epsilon n^2/2$.

We will show that the expander metric (V, dist_G) embeds into ℓ_p with distortion $3D$ in the standard sense (i.e. with 0-slack). Remove from H nodes that belong to at least $\sqrt{\epsilon}n$ pairs in I . There are at most $\sqrt{\epsilon}n$ such nodes, which means that at least $\sqrt{\epsilon}n$ nodes survive in each path that was attached to the expander. For each node v of the expander choose a node v^* in the path attached to v that survived, and define an embedding ψ of (V, dist_G) into ℓ_p as

$$\psi(v) \stackrel{\text{def}}{=} \phi(v^*). \quad (16)$$

Let v and w be two different nodes in V . An easy counting argument shows that there exists a node $u \in V'$ that belongs to the path that was attached to v , and such that neither (v^*, u) nor

(w^*, u) belongs to I , the set of “bad” pairs. We have

$$\begin{aligned}
\|\psi(v) - \psi(w)\|_p &= \|\phi(v^*) - \phi(w^*)\|_p \\
&\leq \|\phi(v^*) - \phi(u)\|_p + \|\phi(u) - \phi(w^*)\|_p \\
&\leq D \cdot \text{dist}_H(v^*, u) + D \cdot \text{dist}_H(u, w^*) \\
&\leq D \cdot \text{dist}_H(v, w) + 3D\delta \\
&\leq D \cdot \text{dist}_G(v, w) + 1 \\
&\leq D \cdot \text{dist}_G(v, w) + \text{dist}_G(v, w) \\
&\leq 2D \cdot \text{dist}_G(v, w),
\end{aligned} \tag{17}$$

and also

$$\begin{aligned}
\|\psi(v) - \psi(w)\|_p &= \|\phi(v^*) - \phi(w^*)\|_p \\
&\geq \|\phi(u) - \phi(w^*)\|_p - \|\phi(u) - \phi(v^*)\|_p \\
&\geq \text{dist}_H(u, w^*) - D \cdot \text{dist}_H(u, v^*) \\
&\geq \text{dist}_H(v, w) - D\delta \\
&\geq \text{dist}_G(v, w) - 1/3 \\
&\geq \text{dist}_G(v, w) - 1/3 \cdot \text{dist}_G(v, w) \\
&\geq 2/3 \cdot \text{dist}_G(v, w),
\end{aligned} \tag{18}$$

which implies that ψ is an embedding of the expander metric into ℓ_p with distortion $3D$. It is known that to embed a bounded-degree expander metric on k nodes into ℓ_p we need distortion $\Omega(\frac{1}{p} \log k)$ ([3], see Lecture 3 for a proof of distortion $\Omega(\log k)$ for ℓ_2). This implies that $D = \Omega(\frac{1}{p} \log \frac{1}{\epsilon})$. \square

2.2 Contracting embeddings with slack

Suppose we wanted to construct a *contracting embedding* ϕ with ϵ -slack of a finite metric into ℓ_p , that, is a contracting embedding such that ϕ contracts by at most D on all but an ϵ -fraction of the pairs. What D can we hope for? It turns out that for bounded-degree expanders this still implies distortion $\Omega(\log n)$, that is, we do not gain anything over the standard notion of an embedding.

Theorem 4. *A contracting embedding of a bounded-degree expander on n nodes into ℓ_p requires distortion $\Omega(\frac{1}{p} \log n)$ even with $1/2$ -slack.*

We will only prove it for $p = 2$, using our knowledge from Lecture 3.

Theorem 5. *Let ϵ be a fixed constant in $(0, 1)$. A contracting embedding of a constant-degree expander on n nodes into ℓ_2 with ϵ -slack requires distortion $\Omega(\log n)$.*

Proof. Let $G = (V, E)$ be an r -regular expander, where r is a constant, and let ϕ be a contracting embedding with ϵ -slack and distortion D .

We know that $\mu_2(G) = \Theta(1)$, and in Lecture 3 in the proof of Theorem 4 we showed that for any embedding ϕ of G into ℓ_2 , we have

$$\sum_{(v,w) \in E} \|\phi(v) - \phi(w)\|^2 \geq \frac{\mu_2(G)}{n} \sum_{(v,w) \in \binom{V}{2}} \|\phi(v) - \phi(w)\|^2. \tag{19}$$

Note that because the embedding is contracting, we have

$$\sum_{(v,w) \in E} \|\phi(v) - \phi(w)\|^2 \leq \frac{r}{2}n, \quad (20)$$

and it follows from the last two equations that

$$\sum_{(v,w) \in \binom{V}{2}} \|\phi(v) - \phi(w)\|^2 = O(n^2). \quad (21)$$

On the other hand, for all but an ϵ -fraction of pairs v and w , it holds

$$\|\phi(v) - \phi(w)\|^2 \geq \frac{\text{dist}_G^2(v, w)}{D^2}. \quad (22)$$

For all but an $o(1)$ -fraction of pairs v and w , we have $\text{dist}_G(v, w) = \Theta(\log n)$, and therefore, even if the embedding completely contracts the distance on some ϵ -fraction of pairs, we still have

$$\sum_{(v,w) \in \binom{V}{2}} \|\phi(v) - \phi(w)\|^2 = \Omega\left(\frac{n^2 \log^2 n}{D^2}\right), \quad (23)$$

which means that $D = \Omega(\log n)$. □

References

- [1] Ittai Abraham, Yair Bartal, T.-H. Hubert Chan, Kedar Dhamdhere, Anupam Gupta, Jon M. Kleinberg, Ofer Neiman, and Aleksandrs Slivkins. Metric embeddings with relaxed guarantees. In *FOCS*, pages 83–100. IEEE Computer Society, 2005.
- [2] T.-H. Hubert Chan, Kedar Dhamdhere, Anupam Gupta, Jon M. Kleinberg, and Aleksandrs Slivkins. Metric embeddings with relaxed guarantees. Full version of the FOCS paper. Available at http://www.cs.brown.edu/people/slivkins/focs05_full.pdf.
- [3] Jiří Matoušek. On embedding expanders into ℓ_p spaces. *Israel J. Math.*, 102:189–197, 1997.