## Bourgain's Theorem via Padded Decompositions<sup>1</sup>

• Bourgain's Theorem. In the last lecture, we saw how the generalized/non-uniform sparsest cut can be solved if we could find metric embeddings of a general metric into  $\mathcal{L}_1$  with low distortion. In particular, the following theorem of Bourgain (stylized to capture distortion with respect to S) immediately implies a  $O(\log k)$ -approximation for the general sparsest cut problem.

**Theorem 1** (Bourgain's Theorem, the Terminal Version). Given any metric space (V,d) and a set  $S \subseteq V$  of size at most k, there is a mapping  $\psi: V \to \mathbb{R}^{O(\log^2 k)}$  such that with high probability, we have that for any pair of vertices u and v,  $||\psi(u)-\psi(v)||_1 \le d(u,v)$  and for any pair  $u,v \in S$ ,  $d(u,v) \le O(\log k)||\psi(u)-\psi(v)||_1$ .

• In this note we give a sketch of a proof. In particular, we focus on the case of S=V, that is the case of all pairs. Furthermore, we only prove an "expectation" result rather than a "with high probability" result. More precisely, we describe a randomized algorithm which produces a  $\phi:V\to\mathbb{R}^h$  such that for any two points u and v we have  $\|\phi(u)-\phi(v)\|_1\geq d(u,v)$  but  $\exp[\|\phi(u)-\phi(v)\|_1]\leq O(\log n)\cdot d(u,v)$ . The "with high probability" statement can be obtained by "repeating, averaging, and concatenating" and applying standard deviation inequalities like the Chernoff bound. We leave this as an exercise. The  $\psi$  in the theorem is obtained by defining  $\psi(u):=\frac{\phi(u)}{C\log n}$  for a sufficiently large C.

We describe a proof which uses the random permutation idea that we saw in the randomized multicut algorithm. The key definition is that of **padded decompositions**.

**Definition 1.** Given a metric d over V, a  $(\beta, \Delta)$ -padded decomposition of (V, d) is a distribution over partitions  $\Pi := (V_1, \dots, V_T)$  with the following two properties

- a. The (weak) diameter of each  $V_i \in \Pi$  is at most  $\Delta$ .
- b. For any vertex u and radius r,  $\mathbf{Pr}_{\Pi}[B(u,r) \text{ is shattered by } \Pi] \leq \beta(u) \cdot \frac{4r}{\Lambda}$

Here  $\beta: V \to \mathbb{R}_{\geq 0}$  is a function mapping a non-negative real to u, and could depend on  $\Delta$ . The weak diameter of a subset S is  $\max_{u,v \in S} d(u,v)$ . The set  $B(u,r) := \{v: d(u,v) \leq r\}$  is the ball of radius r around u, and it is shattered by a partition  $\Pi$  if at least two parts  $\Pi_i$  and  $\Pi_j$  have non-trivial intersection with the ball. Finally, a padded decomposition distribution is said to be efficient if it can be efficiently sampled from.

• Padded Decompositions and Embedding into  $\ell_1$ . We now describe how padded decompositions imply embeddings in a fairly natural way. Let  $D := \max_{u,v \in V} d(u,v)$ . Our (randomized) mapping  $\phi$  will be a concatenation of these  $\lceil \log_2 D \rceil$  different  $\phi_t$ 's.

<sup>&</sup>lt;sup>1</sup>Lecture notes by Deeparnab Chakrabarty. Last modified: 18th Mar, 2022

These have not gone through scrutiny and may contain errors. If you find any, or have any other comments, please email me at deeparnab@dartmouth.edu. Highly appreciated!

1: **procedure** Randomized Embedding(V, d):

- 2: **for** t = 0 to  $\lceil \log_2 D \rceil$  **do**:  $\triangleright D := \max_{u,v \in V} d(u,v)$ .
- 3: Sample  $\Pi_t := (V_1, \dots, V_{d_t})$  from a  $(\beta_t, 2^t)$ -padded decomposition distribution.  $\triangleright \beta_t$ 's will be defined later
- 4: Define  $\phi_t(u)$  as a  $d_t$ -dimensional vector corresponding to the different parts:

$$\phi_t(u)[i] = \begin{cases} 2^t & \text{if } u \in V_i \\ 0 & \text{otherwise} \end{cases}$$

- 5:  $\triangleright$  If u and v are in different parts of  $\Pi_t$ , then  $\|\phi_t(u) \phi_t(v)\|_1 = 2^{t+1}$ , else it is 0
- 6: Let  $\phi$  be a *concatenation* of these  $\lceil \log_2 D \rceil$  different  $\phi_t$  vectors. So, the dimension of  $\phi$  is  $\sum_t d_t$ .

**Claim 1.** For any two points u and v and any t,  $\mathbf{Exp}[||\phi_t(u) - \phi_t(v)||_1] \le \beta_t(u) \cdot 8d(u, v)$ . Furthermore, if  $t < \log_2 d(u, v)$ , then  $||\phi_t(u) - \phi_t(v)||_1 = 2^{t+1}$  with probability 1.

*Proof.* u and v are in different parts of  $\Pi_t$  is equivalent to the event that the ball B(u,d(u,v)) is shattered by  $\Pi_t$ . By the definition of padded decompositions, the probability of this is at most  $4\beta_t(u)d(u,v)/2^t$ . Therefore,  $\mathbf{Exp}[||\phi_t(u)-\phi_t(v)||_1] \leq \frac{4\beta_t(u)}{2^t} \cdot 2^{t+1}$ , and thus the first assertion of the claim follows. Furthermore, if  $t < \log_2 d(u,v)$  implying  $d(u,v) > 2^t$ , then from the fact that the diameter of every part is  $\leq 2^t$  one gets that u and v cannot be in the same part. And so,  $||\phi_t(u)-\phi_t(v)||_1 = 2^{t+1}$  with probability 1.

By the second assertion in Claim 1, we get

For any 
$$u, v, ||\phi(u) - \phi(v)||_1 \ge \sum_{t=0}^{\lfloor \log_2 d(u, v) \rfloor} 2^{t+1} \ge d(u, v)$$
 (1)

By the first assertion in Claim 1, we get

For any 
$$u, v$$
,  $\mathbf{Exp}[||\phi(u) - \phi(v)||_1] \le 8d(u, v) \sum_{t=0}^{\log_2 D} \beta_t(u)$  (2)

In sum, we get an embedding of d into  $\ell_1$  with distortion depending on the  $\beta$ -parameter of the padded decomposition. In the next bullet point, we show how to obtain a padded decomposition with the following parameters.

**Theorem 2.** For any metric space (V, d) and parameter t, there exists a  $(\beta_t, 2^t)$  padded decompisation with

$$\beta_t(u) \le 2 \ln \left( \frac{|B(u, 2^t)|}{|B(u, 2^{t-3})|} \right)$$

If we substitute this in (2), we get

For any 
$$u, v$$
,  $\mathbf{Exp}[||\phi(u) - \phi(v)||_1] \le 8d(u, v) \sum_{t=0}^{\log_2 D} \ln\left(\frac{|B(u, 2^t)|}{|B(u, 2^{t-3})|}\right)$ 

Now note that the summations telescope to  $\leq 24 \ln n \cdot d(u,v)$ . And this completes the proof sketch of Theorem 1.

- **Padded Decomposition Distributions.** We now describe a randomized algorithm which generates samples from a  $(\beta_t(u), 2^t)$ -padded decomposition with  $\beta_t(u) \leq 2 \ln \left( \frac{|B(u, 2^t)|}{|B(u, 2^{t-3})|} \right)$ .
  - 1: **procedure** PADDED DECOMPOSTION(t): ▶ Return a padded decomposition as asserted in Theorem 2
  - 2: Sample a random permutation  $\sigma$  of the points in V.
  - 3: Sample  $R \in [2^{t-2}, 2^{t-1}]$  uniformly at random.
  - 4: Define  $V_i := \{v : d(i, v) \leq R\} \setminus \bigcup_{j \leq \sigma^i} V_j$ .

It is clear that, by design, the diameter of every  $V_i$  is at most  $2R \le 2^t$ . What is more interesting is to prove that for any  $u \in V$  and any r, the probability B(u,r) is shattered is at most  $\beta_t(u) \cdot \frac{4r}{2^t}$ . Let B denote this ball B(u,r). First, observe that if  $r > 2^{t-3}$ , then  $\frac{4r}{2^t} > 1$  and so the shattering claim holds vacuously. Therefore, henceforth we assume  $r \le 2^{t-3}$ .

Let us consider a vertex i such that  $V_i$  is the first in  $\sigma$ -order to shatter B(u,r). For this to occur, we must have  $d(u,i)-r \leq R$  and  $R \leq d(u,i)+r$ : the former since  $V_i$  intersects B(u,r) and the latter since it doesn't contain all of it. Since  $R \in [2^{t-2}, 2^{t-1}]$ , we get that i must lie in the set  $X := B(u, 2^{t-1}+r) \setminus B(u, 2^{t-2}-r)$ . Furthermore, in the random permutation  $\sigma$ , i must appear before any vertex  $j \in B(u, 2^{t-2}-r)$  otherwise i won't be the *first* vertex to shatter the ball (either someone else would have shattered, or j would've gobbled the whole ball B(u,r).) Finally, note that if i can non-trivially intersect B, then any  $j \in X$  with  $d(j,B) \leq d(i,B)$  can non-trivially intersect B. Therefore, if i were the first in  $\sigma$  to shatter B, it better be that all  $j \in X$  with  $d(j,B) \leq d(i,B)$  come after i in  $\sigma$ .

$$\begin{aligned} \mathbf{Pr}[B(u,r) & \text{ shattered}] = & & & \mathbf{Pr}[\exists i \in X : V_i \text{ is the first in } \sigma \text{ to shatter } B(u,r)] \\ & \leq & & & & \sum_{i \in X} \mathbf{Pr}[V_i \text{ is the first in } \sigma \text{ to shatter } B(u,r)] \\ & \leq & & & & \sum_{i \in X} \mathbf{Pr}[R \in [d(u,i) \pm r] \text{ and } \mathcal{E}_i] \end{aligned}$$

where  $\mathcal{E}_i$  is the event that all vertices  $j \in B(u, 2^{t-1} + r) \leq_{\sigma} i$  satisfy (a)  $j \notin B(u, 2^{t-2} - r)$  and (b) d(j, B) > d(i, B). As explained above, if  $\mathcal{E}_i$  doesn't occur then i cannot be the first vertex to shatter B. Note that  $\mathcal{E}_i$  is independent of  $R \in [d(u, i) \pm r]$ . And therefore,

$$\mathbf{Pr}[B(u,r) \text{ shattered}] \leq \sum_{i \in X} \mathbf{Pr}[R \in [d(u,i) \pm r] \cdot \mathbf{Pr}[\mathcal{E}_i] \leq \frac{4r}{2^t} \cdot \sum_{i \in X} \mathbf{Pr}[\mathcal{E}_i]$$

If we sort the points in  $B(u, 2^{t-1} + r)$  in increasing order of distance from u, then  $\Pr[\mathcal{E}_i]$  is  $\frac{1}{i}$ , and i ranges precisely from  $|B(u, 2^{t-2} - r)|$  to  $|B(u, 2^{t+1} + r)|$  since that is where the points in X lie. This harmonic sum is indeed bounded by

$$\ln\left(\frac{|B(u,2^{t-1}+r)|}{|B(u,2^{t-2}-r)|}\right) \le \ln\left(\frac{|B(u,2^t)|}{|B(u,2^{t-3})|}\right)$$

since  $r \leq 2^{t-3}$ . This ends the sketch of the proof of Theorem 2.

## **Notes**

Bourgain's theorem on metric embeddings is from the paper [2]. The terminal version as stated in Theorem 1 is first stated in the paper [5] by Linial, London, and Rabinovich, and also in the paper [1] by Aumann and Rabani. The proof above is inspired from the paper [4] by Fakcharoenphol, Rao, and Talwar, which itself is inspired from the paper [3] by Calinescu, Karloff, and Rabani.

## References

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