

## Lecture 20

Lecturer: Madhu Sudan

Scribe: Amit J. Deshpande

## 1 Overview

In this lecture, we will see some complexity results for coding problems - known hardness results and some open questions.

- Hardness of the Nearest Codeword Problem (NCP)
- Approximation variants
- Decoding with preprocessing
- Decoding with relatively near codeword
- Minimum distance problem

## 2 Nearest Codeword Problem

The problem of finding out the nearest codeword (or maximum likelihood decoding) to a given received vector has been of crucial importance in the theory of error-correcting codes. Since in the general case, where the code is described by an encoding circuit, the problem of finding a message corresponding to a given codeword is already hard so we might as well restrict our attention only to the linear codes. So given the code by its generator matrix and a received vector, find out a codeword nearest to it. We will formalize this as follows -

**Definition 1 (Nearest Codeword Problem - NCP)** *Given a code with generator matrix  $G$  and received vector  $r$ , find  $x$  that minimizes  $\Delta(xG, r)$ .*

How hard is it to solve NCP ? We will show that NCP is hard even for the special case when  $r = \bar{1}$ . This is done by a reduction from Max Cut (which is a well-known NP-hard problem).

**Definition 2 (Max Cut Problem)** *Given graph  $H = (V, E)$  find  $S \subseteq V$  such that, the number of edges between  $S$  and  $\bar{S}$  is maximum.*

The reduction goes as follows - Let  $G$  be the incidence matrix of a graph  $H = (V, E)$  with  $|V| = k$  and  $|E| = n$ . So our message  $x$  corresponds to the subset of  $V$  specified by the 1's in it and codewords correspond to those edges  $e$  which give 1 after multiplication by  $x$ . i.e. both the 1's in  $e$  cannot be in  $S$  or  $\bar{S}$ . So  $e$  must be a crossing edge. Thus the codewords correspond to cuts and finding max cut is equivalent to finding the maximum weight codeword (meaning, nearest to  $\bar{1}$ ).

### 3 Approximation variants of NCP

There are three important variants of the approximation problems. For a given instance  $(G, r)$  we will define  $\tau = \min_x \{\Delta(xG, r)\}$ , and  $\alpha > 1$  be our approximation parameter.

**Definition 3 (Search question):** Find  $x'$  such that  $\tau \leq \Delta(x'G, r) \leq \alpha \cdot \tau$ .

**Definition 4 (Estimation question):** Estimate  $t$  such that  $\tau \leq t \leq \alpha \cdot \tau$ .

**Definition 5 (Gap decision problem):** (“promise” problem) Given  $(G, r, t)$  with the promise that  $\tau \notin [t, \alpha t]$  decide if  $\tau \leq t$  or not.

And it’s easy to observe that a solution to search problem gives a solution to estimation problem, and a solution to estimation problem gives a solution to Gap decision problem. Also as  $\alpha$  becomes closer and closer to 1 the problems get harder. Analogous definitions can be made for the maximization versions of these problems.

### 4 Hardness of approximating NCP

A critical question would be - is it hard even to find an approximately nearest codeword ?

We know that Max Cut is hard to approximate to within some  $\alpha < 1$ . So we can use this fact to show the hardness for *NCP*. Elementary probability (first moment method) gives that every graph has at least a cut of size  $|E|/2$ , where  $|E|$  is the number of edges. And the reduction that we used for showing the NCP is NP-hard says that finding a Max Cut of size  $x$  corresponds to getting a codeword of weight  $x$ . i.e. a codeword within distance  $n - x$  from  $\bar{1}$ . But since we know that  $x \geq n/2$ . This alongwith an  $\beta$ -approximation to NCP within  $n - x$ ,  $n - x \leq n - x' \leq \beta(n - x)$ , gives that  $\frac{1}{(2-\beta)}x \leq x' \leq x$ . And thus a  $\alpha = 1/(2-\beta)$ -approximation to Max Cut. And  $\alpha \rightarrow 1$  as  $\beta \rightarrow 1$ . But we already know that Max Cut cannot be approximated within  $\alpha < 1$  for some  $\alpha$ , which implies the corresponding hardness result for NCP as -

**Theorem 6** *NCP is hard to approximate to within some  $\beta > 1$ .*

Moreover, we can prove something stronger as this problem has a self-improving property.

**Theorem 7**  *$\beta$ -approximation to NCP is hard implies that  $\beta^2$ -approximation is also hard. And using this repetitively we get, any constant approximation to NCP is hard.*

**Proof** The proof involves a clever construction - given  $G$  generator matrix of a code of length  $n$ , we can construct a “product”  $G^{(2)}$  generator matrix of a code of length  $n^2$  such that  $G$  has a codeword of weight  $n - w$  iff  $G^{(2)}$  has a codeword of weight  $n^2 - w^2$ .

A codewords of  $G^{(2)}$  is an  $n \times n$  matrix with columns labelled by a codeword of  $G$ . Each column is a codeword of  $G$  or its complement according to the label 0 or 1, respectively. To our surprise, this happens to be a linear code.

So if  $G$  has a code of weight  $n - w$  then we can consider the codeword in  $G^{(2)}$  that has  $\bar{1}$  in all the columns labelled by 1’s and the  $n - w$  weight code in  $G$  in all the columns labelled

by 0's. And the labelling also corresponds to the  $n - w$  weight codeword of  $G$ . This gives a codeword of in  $G^{(2)}$  of weight  $n^2 - w^2$ . And that's the maximum you can do to stuff your matrix with more and more 1's.

This clearly implies that if there is a  $\beta^2$ -approximation algorithm for the code  $G^{(2)}$  then it should give a  $\beta$ -approximation for code  $G$ . And thus  $\beta$ -approximation hardness for  $G$  translates into  $\beta^2$ -approximation hardness, too. ■

## 5 Criticism

There has been a lot of criticism on this which gives rise to the following problems -

- Code shouldn't be part of the input and we should be given a lot of preprocessing time to devise the decoding algorithm.
- How do these results relate to the error-correction property ? To make sense, we should be trying to correct less errors than the minimum distance of the code.
- The codes we saw here had a very low-density generator matrix as it was corresponding to the incidence matrix of a graph. But we want hardness results for better codes. e.g. Reed-Solomon codes, algebraic geometry codes, LDPC codes, Turbo codes (any of your favourite codes).

We will analyze some results that try to address these questions.

## 6 More hardness results addressing the criticism

### 6.1 Hardness of decoding a fixed family of codes [Bruck-Naor]

The first criticism regarding sparse generator matrix was addressed by Bruck-Naor [1] and the idea was to “inject” the generator of the code into received vector, while fixing the code. Let  $G$  be the incidence matrix of a graph. For every pair of vertices  $(u, v)$ , have twin-pair of columns. So such a code  $C$  has a generator matrix with  $2 \binom{k}{a}$  columns. Now suppose that we have code  $B$  and received vector  $r$  as an instance of NCP. Construct a new received vector as follows: if edge  $(u, v)$  is in  $G$  then duplicate the entry of  $r$  in the corresponding coordinate of  $r'$ , and otherwise put 0, 1.

Now note that,  $\Delta(xC, r') = N/2 - n - 2\Delta(xB, r)$  where  $N$  and  $n$  are the block lengths of  $C$  and  $B$ , respectively. So the minimum distances are related and we cannot compute NCP exactly for the code  $C$ .

This method also works when the generator matrix is  $a$ -sparse (in fact, more generally). Hardness of approximating in this setting is studied in Feige-Micciancio [2].

## 6.2 Decoding codes upto min distance [Dumer-Micciancio-Sudan]

This addressed the other criticism regarding hardness results for asymptotically good codes. Dumer-Micciancio-Sudan [3] show that we can “boost” the distance of the code without altering the problem too much. This was shown by showing a hardness result for a version of Gap Decision Problem for the minimum distance.

Suppose that finding the nearest codeword to code generated by  $A$  is hard to approximate (to within factor of 100, say). Then we specifically have  $A, r, d$  such that telling if  $\tau > d$  or  $\tau \leq d/100$  with high probability is hard. The trick is to attach to  $A$  a generator matrix  $B$  of a code of distance  $d$ , and getting an appropriate  $r'$ .

Dumer-Micciancio-Sudan [3] show that decoding codes of minimum distance  $d$  for upto less than  $d$  errors is NP-hard.

## 7 Open questions

All these still raise a few more open questions -

- Can you solve NCP is polytime for some asymptotically good family of codes ? Reed-Solomon ? or your favourite code ?
- Does there exist a single decoding algorithm decoding all codes upto half the minimum distance ?
- Does there exist an algorithm giving a lower bound for minimum distance which guarantees that if the relative distance is  $1 - \frac{1}{q} - \epsilon$  then the lower bound given by the algorithm is at least  $1 - \frac{1}{q} - \epsilon^2$  ?

## References

- [1] J.Bruck, M.Naor, *The hardness of decoding linear code with preprocessing*, IEEE Transaction on Information Theory, pp. 381-385, May 1990.
- [2] U.Feige, D.Micciancio, *The inapproximability of lattice and coding problems with preprocessing*, Proceedings of IEEE Conference on Computational Complexity, pp. 44-52, 2002.
- [3] I.Dumer, D.Micciancio, M.Sudan, *Hardness of approximating the minimum distance of a linear code*, IEEE Transaction on Information Theory, Jan 2003 (to appear).