

# Expander Graphs and Their Applications (XI)

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## Review of the Previous Lecture

# MAX- $d$ SAT

Let  $d \in \mathbb{N}$ .

MAX- $d$ SAT

*Input:* An  $\alpha \in d\text{CNF}$ .

*Solution:* An assignment  $\mathcal{V}$  for  $\alpha$ .

*Cost:* the number of clauses in  $\alpha$  that  $\mathcal{V}$  satisfies.

*Goal:* max.

## Definition

For every  $d\text{CNF}$ -formula  $\alpha$  and an assignment  $\mathcal{V}$  for  $\alpha$ ,

$\text{sat}(\alpha, \mathcal{V})$  := the number of clauses in  $\alpha$  that  $\mathcal{V}$  satisfies

and

$\text{optsat}(\alpha)$  :=  $\max_{\mathcal{V}} \text{sat}(\alpha, \mathcal{V})$ .

# Hardness of Approximating MAX- $d$ SAT

## Definition

Let  $\mathbb{A}$  be an algorithm that on every  $d$ CNF-formula  $\alpha$  always output an assignment for  $\alpha$  which we denote by  $\mathcal{V}_\alpha$ . The performance ratio of  $\mathbb{A}$  is

$$\max_{\alpha \in d\text{CNF}} \frac{\text{optsat}(\alpha)}{\text{sat}(\alpha, \mathcal{V}_\alpha)}.$$

## Theorem

*For some constants  $d \in \mathbb{N}$  and  $r > 1$ , there is **no** polynomial time approximation algorithm for MAX- $d$ SAT with performance ratio  $r$ , unless  $P = NP$ .*

## Proof Idea.

We reduce an NP-complete  $Q$  with  $(c \cdot \log n, d)$ -restricted verifier to  $\text{MAX-}d\text{SAT}$ .

Given an  $x \in \Sigma^*$ , we will construct a  $d\text{CNF}$  formula  $\alpha_x$  satisfying

$x \in Q \Rightarrow \alpha_x$  is satisfiable,

$x \notin Q \Rightarrow$  for any assignment  $\mathcal{V}$ ,  
at least  $2^{-d-1}$ -fraction of clauses are not satisfied.

$\text{unsat}(\alpha)$

Let  $\alpha \in d\text{CNF}$ , and we define

$\text{unsat}(\alpha)$  := the smallest fraction of  
*unsatisfied* clauses of  $\alpha$  under any assignment.

$\alpha$  is satisfiable  $\iff \text{unsat}(\alpha) = 0$ ;  
 $\alpha$  is not satisfiable  $\iff \text{unsat}(\alpha) \geq \frac{1}{m}$   
*where  $m$  is the number of clauses in  $\alpha$ .*

Let  $\epsilon > 0$ .

GAP- $d$ SAT $_{\epsilon}$

*Input:* A propositional formula  $\alpha \in d\text{CNF}$ , such that *either*  
 $\text{unsat}(\alpha) = 0$  or  $\text{unsat}(\alpha) \geq \epsilon$ .

*Problem:* Decide whether  $\alpha$  is satisfiable.

GAP- $d$ SAT $_{\epsilon}$  is a *promise problem*.

# Hardness of $\text{GAP-}d\text{SAT}$

## Theorem

There exist some  $d \in \mathbb{N}$  and  $\epsilon > 0$  such that  $\text{GAP-}d\text{SAT}_\epsilon$  is NP-hard.

## Proof.

Recall in the previous proof, for some NP-hard problem  $Q$ , appropriate  $d \in \mathbb{N}$  and  $\epsilon > 0$ , we implicitly define an algorithm  $\mathbb{A}$ , such that for each instance  $x \in \Sigma^*$ , the algorithm  $\mathbb{A}$  produces a  $d\text{CNF}$ -formula  $\alpha_x$  satisfying:

- $x \in Q \Rightarrow \alpha_x$  is satisfiable, i.e.,  $\text{unsat}(\alpha_x) = 0$ ,
- $x \notin Q \Rightarrow$  for any assignment  $\mathcal{V}$ , at least  $\epsilon$ -fraction of clauses are not satisfied, i.e.,  $\text{unsat}(\alpha_x) \geq \epsilon$ .

That is,  $\mathbb{A}$  gives a reduction from  $Q$  to  $\text{GAP-}d\text{SAT}_\epsilon$ . □



# Hardness of GAP-3SAT

## Theorem

*There exists an  $\epsilon > 0$  such that  $\text{GAP-3SAT}_\epsilon$  is NP-hard.*

# The Reduction from $d$ SAT to 3SAT

## Lemma

Let  $d > 3$  and  $r \geq 1$ . there is a polynomial time algorithm  $\mathbb{R}$  such that, given an  $\alpha \in d\text{CNF}$ , the algorithm  $\mathbb{R}$  produces an  $\alpha^* \in 3\text{CNF}$  such that

- (i)  $\alpha$  is satisfiable  $\Rightarrow \alpha^*$  is satisfiable.
- (ii) For any assignment, at least  $\epsilon$ -fraction of clauses in  $\alpha$  is not satisfied  $\implies$  for any assignment, at least  $\frac{\epsilon}{d-2}$ -fraction of clauses in  $\alpha^*$  is not satisfied.

**Proof.** Let  $\alpha \in d\text{CNF}$ , i.e., for an **index set**  $I$ ,

$$\alpha = \bigwedge_{i \in I} \beta_i$$

where each

$$\beta_i = \lambda_{i,1} \vee \lambda_{i,2} \vee \dots \vee \lambda_{i,t_i}$$

for some literals  $\lambda_{i,1}, \lambda_{i,2}, \dots, \lambda_{i,t_i}$  and  $t_i \leq d$ .

## Proof. (cont'd)

For every  $i \in I$  with  $t_i > 3$ , let  $Y_{i,1}, \dots, Y_{i,t_i-3}$  be some new variables. And let  $\gamma_i$  be the following formula

$$(\lambda_{i,1} \vee \lambda_{i,2} \vee Y_{i,1}) \wedge (\neg Y_{i,1} \vee \lambda_{i,3} \vee Y_{i,2}) \wedge \cdots \wedge (\neg Y_{i,t_i-3} \vee \lambda_{i,m-1} \vee \lambda_{i,t_i}).$$

Otherwise let  $\gamma_i := \beta_i$ . Then we take

$$\alpha^* := \bigwedge_{i \in I} \gamma_i$$

as the required 3CNF formula.

(i) is easy. We show (ii).

Let  $\mathcal{V}^*$  be an assignment for  $\alpha^*$ . And let  $\mathcal{V}$  be *the restriction of  $\mathcal{V}^*$  to the variables of  $\alpha$* .

Assume  $\alpha$  contains  $m$  clauses. By assumption, under  $\mathcal{V}$ , at least

$$[\epsilon \cdot m]$$

clauses are *not* satisfied.

## Proof. (cont'd)

Let  $\beta_i = \lambda_{i,1} \vee \lambda_{i,2} \vee \dots \vee \lambda_{i,t_i}$  be an unsatisfied clause. Thus, in

$$\gamma_i = (\lambda_{i,1} \vee \lambda_{i,2} \vee Y_{i,1}) \wedge (\neg Y_{i,1} \vee \lambda_{i,3} \vee Y_{i,2}) \wedge \dots \wedge (\neg Y_{i,t_i-3} \vee \lambda_{i,m-1} \vee \lambda_{i,t_i}).$$

at least one clause is not satisfied under  $\mathcal{V}^*$ .

Hence we have at least  $\lceil \epsilon \cdot m \rceil$  unsatisfied clauses in  $\alpha^*$ .

$\alpha^*$  contains at most

$$(d-2) \cdot m$$

clauses. Thus the fraction of clauses that are not satisfied under  $\mathcal{V}^*$  is at least

$$\frac{\lceil \epsilon \cdot m \rceil}{(d-2) \cdot m} \geq \frac{\epsilon}{d-2}.$$

□

## Hardness of MAX-3SAT

### Theorem

There is a constant  $r > 1$  such that there is no polynomial time approximation algorithm for MAX-3SAT with performance ratio  $r$ , unless  $P = NP$ .

**Proof.** By the previous theorem, there is some  $\epsilon > 0$  such that  $\text{GAP-3SAT}_\epsilon$  is NP-hard. Without loss of generality, we assume  $\epsilon < 1$ . Choose any

$$r < \frac{1}{1 - \epsilon}.$$

Assume an algorithm  $\mathbb{A}$  approximates MAX-3SAT with performance ratio  $r$ . Let  $\alpha \in 3\text{CNF}$  with  $m$  clauses.

(i) If  $\text{unsat}(\alpha) = 0$ , then  $\mathbb{A}$  outputs an assignment that satisfies at least

$$\left\lceil \frac{m}{r} \right\rceil > (1 - \epsilon) \cdot m \text{ clauses.}$$

(ii) If  $\text{unsat}(\alpha) \geq \epsilon$ , then any assignment (including the one  $\mathbb{A}$  outputs) can satisfy less than

$$m - \lceil \epsilon \cdot m \rceil \leq m - \epsilon \cdot m \text{ clauses.}$$



# PCP by the Hardness of GAP- $d$ SAT

## Theorem

*If there exists some  $1 \geq \epsilon > 0$ , such that  $\text{GAP-3SAT}_\epsilon$  is NP-hard, then  $\text{NP} = \text{PCP}(\log n, 1)$ .*

## Proof.

Let  $Q \subseteq \Sigma^*$  be a problem in NP. Then there is a polynomial time algorithm  $\mathbb{R}$  such that for every  $x \in \Sigma^*$ , the algorithm  $\mathbb{A}$  computes an  $\alpha_x \in 3\text{CNF}$  such that

$$x \in Q \implies \text{unsat}(\alpha_x) = 0;$$

$$x \notin Q \implies \text{unsat}(\alpha_x) \geq \epsilon.$$

Without loss of generality, we assume  $\alpha_x$  *contains at most*  $|x|^c$  *many clauses*, for some  $c \in \mathbb{N}$ .

Choose any  $d \in \mathbb{N}$  with

$$\left(1 - \frac{\epsilon}{2}\right)^d \leq \frac{1}{2},$$

e.g.,  $d := \lceil 2/\epsilon \rceil$ .

## Proof. (cont'd)

Consider the following verifier  $V(x, \tau, \pi)$  with  $|\tau| = c \cdot d \cdot \log |x|$ :

1. We view  $\tau$  as *a sequence of  $d$  many  $c \cdot \log |x|$ -bit binaries*, which correspond to numbers  $t_1, \dots, t_d$ , each between 0 and  $|x|^c - 1$ .
2. Simulate  $\mathbb{R}$  on  $x$  to compute  $\alpha_x$ .
3.  $m \leftarrow$  the number of clauses in  $\alpha_x$ .
4. For all  $1 \leq i \leq d$ ,  $p_i \leftarrow (t_i \bmod m) + 1$ .
5. Check whether  $\pi$ , as an assignment, satisfies the  $p_1$ -th,  $\dots$ ,  $p_d$ -th clauses. If so, accept. Otherwise reject.

Clearly  $V$  is a  $(\log n, 1)$ -restricted verifier.



## Proof. (cont'd)

Now we have to check  $V$  correctly decides  $Q$ .

If  $x \in Q$ , then  $\text{unsat}(\alpha_x) = 0$ .

There is an assignment  $\pi$  satisfy  $\alpha_x$ . Thus no matter what  $\tau$  we have,  $V(x, \tau, \pi)$  always accepts.

If  $x \notin Q$ , then  $\text{unsat}(\alpha_x) \geq \epsilon$ , i.e., no matter what assignment  $\pi$  we have, at least  $\epsilon$ -fraction of clauses are not satisfied.

We pick a number  $t$  with  $0 \leq t < |x|^c$  in random, and let

$$p := (t \bmod m) + 1$$

where  $m$  is the number of clauses in  $\alpha_x$ . And let

$$s := \left\lfloor \frac{|x|^c}{m} \right\rfloor.$$

$s \geq 1$  by our assumption, and

$$s \cdot m \leq |x|^c \leq (s + 1) \cdot m.$$

## Proof. (cont'd)

It follows that when  $t$  ranges from 0 to  $|x|^c - 1$ , for at least

$$\lceil \epsilon \cdot m \rceil \cdot s$$

many corresponding  $p$ s, the  $p$ -th clause is *not* satisfied.

$$\frac{\lceil \epsilon \cdot m \rceil \cdot s}{|x|^c} \geq \frac{s \cdot m}{(s+1) \cdot m} \cdot \epsilon \geq \frac{\epsilon}{2} \quad \text{by } s \geq 1.$$

Thus, the probability that the  $p$ -th clause is satisfied is less than

$$1 - \frac{\epsilon}{2}.$$

So the probability that  $V$  accepts is less than

$$\left(1 - \frac{\epsilon}{2}\right)^d,$$

hence less than  $1/2$  by  $(1 - \epsilon/2)^d \leq 1/2$ .

□

## Constraint Satisfaction Problems

# Constraints

## Definition

Let  $V = \{v_1, \dots, v_n\}$  be a set of variables, and let  $\Sigma$  be a finite alphabet. A  $q$ -ary constraint  $c = (C, i_1, \dots, i_q)$  consists of

- ▶ a  $q$ -tuple of indices  $i_1, \dots, i_q \in [n]$ ,
- ▶ and a subset  $C \subseteq \Sigma^q$  of “acceptable” values.

An assignment  $\mathcal{A} : V \rightarrow \Sigma$  *satisfies*  $c$  if

$$(\mathcal{A}(v_{i_1}), \dots, \mathcal{A}(v_{i_q})) \in C.$$

The formula  $X_3 \vee \neg X_7 \vee X_9$  can be viewed as a constraint  $(C, 3, 7, 9)$  with

$$\{(1, 0, 0), (1, 0, 1), (1, 1, 0), (1, 1, 1), (0, 0, 0), (0, 0, 1), (0, 1, 1)\}.$$

# Constraint Satisfaction Problems (CSP)

CSP

*Input:* A system (i.e., set)  $\mathcal{C}$  of constraints over a set of variables  $V$  and a finite alphabet  $\Sigma$ .

*Problem:* Is there an assignment  $\mathcal{A} : V \rightarrow \Sigma$  that satisfies every constraint?

## 3SAT as CSP

### Lemma

3SAT is reducible to CSP.

### Proof.

Let  $\Sigma := \{0, 1\}$ . Given an instance  $\alpha$  of 3SAT

$$\bigwedge_{i \in I} \lambda_{i,1} \vee \lambda_{i,2} \vee \lambda_{i,3},$$

let  $V$  be the set of propositional variables in  $\alpha$ . For each  $i \in I$ , assume  $\lambda_{i,j}$  is a literal over the variable  $X_{i,j}$  for  $j = 1, 2, 3$ . Then we introduce a constraint  $c_i := (C_i, i_1, i_2, i_3)$  where

$$C_i := \{(b_1, b_2, b_3) \in \Sigma^3 \mid X_{i_1} \mapsto b_1, X_{i_2} \mapsto b_2, X_{i_3} \mapsto b_3 \\ \text{satisfies } \lambda_{i,1} \vee \lambda_{i,2} \vee \lambda_{i,3}\}.$$

Then

$\alpha$  is satisfiable  $\iff$  there is an assignment satisfying  $\{C_i \mid i \in I\}$ .



## 3COLORABILITY as CSP

### Lemma

3COLORABILITY *is reducible to* CSP.

### Proof.

Let  $\Sigma := \{0, 1, 2\}$ . Given a graph  $G = (V, E)$ , we identify  $V$  with the set of variables. For each edge  $e \in E$  with end vertices  $v_i$  and  $v_j$ , we introduce a constraint

$$c_e := (C_e, i, j)$$

where

$$C_i := \{(a, b) \in \Sigma^2 \mid i \neq j\}.$$

Then

$G$  is 3-colorable  $\iff$  there is an assignment satisfying  $\{C_e \mid e \in E\}$ .

□