

# Complexity Theory

Jan Křetínský

Technical University of Munich  
Summer 2019

June 4, 2019

## Lecture 21

$$\text{NP} \subseteq \text{PCP}[\text{poly}(n), 1]$$

## Recap: Two views of the PCP theorem

prob. checkable proofs

hardness of approximation

---

PCP verifier  $V$

$\leftrightarrow$  CSP instance

proof  $\pi$

$\leftrightarrow$  variable assignment

$|\pi|$

$\leftrightarrow$  number of variables in CSP

number of random bits

$\leftrightarrow$   $\log m$ , where  
 $m$  is number of clauses

number of queries

$\leftrightarrow$  arity of constraints

# Goal and plan

## Goal

- proof a weaker PCP theorem
- learn interesting encoding/decoding schemes useful in such proofs

## Plan

- proof
  - an NP-complete language: Quadeq
  - Walsh-Hadamard encodings
  - a PCP[poly, 1] system for Quadeq
- summary: PCP and hardness of approximation

# Weak PCP

## Theorem

$$\text{NP} \subseteq \text{PCP}[\text{poly}, 1]$$

**Proof:** It suffices to come up with a PCP system for **one NP**-complete language, where the verifier

- uses **polynomially many** random bits (exponentially long proofs)
- makes a **constant** number of queries to that proof

**Plan:**

- an **NP**-complete language: **Quadeq**
- Walsh-Hadamard encodings
- a **PCP**[*poly*, 1] system for **Quadeq**

# Disclaimer

All arithmetic today will be **modulo 2**, that is, over the field  $\{0, 1\}$ !

- $1 + 1 = 0$
- $x^2 = x$
- $x + y = x - y$

# Quadeq

- **satisfiable** quadratic equations over  $\{0, 1\}$
- $n$  variables/ $m$  equations
- no purely linear terms
- **NP**-complete (exercise!)

## Example (Running example)

$$\begin{aligned}xy + xz &= 1 \\y^2 + yz + z^2 &= 1 \\x^2 + yx + z^2 &= 0\end{aligned}$$

# Quadeq

- **satisfiable** quadratic equations over  $\{0, 1\}$
- $n$  variables/ $m$  equations
- no purely linear terms
- **NP**-complete (exercise!)

## Example (Running example)

$$\begin{aligned}xy + xz &= 1 \\y^2 + yz + z^2 &= 1 \\x^2 + yx + z^2 &= 0\end{aligned}$$

**Solution:**  $x = 1, y = 0, z = 1$

as a vector:  $\mathbf{s} = (1 \ 0 \ 1)$



## Be smart, use vector notation

$$xy + xz = 1$$

$$y^2 + yz + z^2 = 1$$

$$x^2 + yx + z^2 = 0$$

$$\mathbf{s} = (1 \ 0 \ 1)$$

## Be smart, use vector notation

$$xy + xz = 1$$

$$y^2 + yz + z^2 = 1$$

$$x^2 + yx + z^2 = 0$$

$$\mathbf{s} = (1 \ 0 \ 1)$$

**vector notation:** for a given  $m \times n^2$  matrix  $A$  and  $m$  vector  $\mathbf{b}$  find solution  $\mathbf{u} = (x \ y \ z)$  such that

$$A(\mathbf{u} \otimes \mathbf{u}) = \mathbf{b}$$

$\mathbf{u} \otimes \mathbf{u}$	$x^2$	$xy$	$xz$	$yx$	$y^2$	$yz$	$zx$	$zy$	$z^2$	
$\mathbf{s} \otimes \mathbf{s}$	1	0	1	0	0	0	1	0	1	$\mathbf{b}$
$A$	0	1	1	0	0	0	0	0	0	1
	0	0	0	0	1	1	0	0	1	1
	1	0	0	1	0	0	0	0	1	0

# Overview

- **Quadeq** is the language of **satisfiable** systems of quadratic equations over  $\{0, 1\}$
- natural PCP system expects a solution **u** and checks whether it is valid
- but this yields **superconstant** number of queries!
- how can we encode a solution such that a constant number of queries suffices?

# Overview

- **Quadeq** is the language of **satisfiable** systems of quadratic equations over  $\{0, 1\}$
  - natural PCP system expects a solution **u** and checks whether it is valid
  - but this yields **superconstant** number of queries!
  - how can we encode a solution such that a constant number of queries suffices?
  - **use longer proofs!**
- 
- an **NP**-complete language: **Quadeq** ✓
  - Walsh-Hadamard encodings
  - a **PCP**[*poly*, 1] system for **Quadeq**

# PCP for Quadeq

Input:  $m \times n^2$  matrix  $A$ ,  $m$  vector  $\mathbf{b}$

Verifier

Proof  $\pi$

---

1. check that  $f, g$  are linear functions
2. check that  $g = WH(\mathbf{u} \otimes \mathbf{u})$  where  $f = WH(\mathbf{u})$
3. check that  $g$  encodes a satisfying assignment

- $\pi \in \{0, 1\}^{2^n + 2^{n^2}}$
- $\pi$  is a pair of linear functions  $\langle f, g \rangle$ , i.e. strings from  $\{0, 1\}^{2^n}$  and  $\{0, 1\}^{2^{n^2}}$ , resp.
- if  $\mathbf{u}$  satisfies  $A(\mathbf{u} \otimes \mathbf{u}) = \mathbf{b}$  then  $f = WH(\mathbf{u})$  and  $g = WH(\mathbf{u} \otimes \mathbf{u})$  are Walsh-Hadamard encodings

# Walsh-Hadamard encoding

## Definition (WH)

Let  $\mathbf{u} \in \{0, 1\}^n$  be a vector. The **Walsh-Hadamard** encoding of  $\mathbf{u}$  written  $WH(\mathbf{u})$  is the **truth table** of the linear function  $f : \{0, 1\}^n \rightarrow \{0, 1\}$  with  $f(\mathbf{x}) = \mathbf{u} \odot \mathbf{x}$  where  $(u_1 \dots u_n) \odot (x_1 \dots x_n) = \sum_{i=1}^n u_i x_i$ .

# Walsh-Hadamard encoding

## Definition (WH)

Let  $\mathbf{u} \in \{0, 1\}^n$  be a vector. The **Walsh-Hadamard** encoding of  $\mathbf{u}$  written  $WH(\mathbf{u})$  is the **truth table** of the linear function  $f : \{0, 1\}^n \rightarrow \{0, 1\}$  with  $f(\mathbf{x}) = \mathbf{u} \odot \mathbf{x}$  where  $(u_1 \dots u_n) \odot (x_1 \dots x_n) = \sum_{i=1}^n u_i x_i$ .

## Example

The solution to our running example is  $\mathbf{s} = (1 \ 0 \ 1)$ . We have

$$WH(\mathbf{s}) = (0 \ 1 \ 0 \ 1 \ 1 \ 0 \ 1 \ 0)$$

Note:  $|WH(\mathbf{u})| = 2^{|\mathbf{u}|}$

# Properties (without proof)

## Random subsum principle

- if  $\mathbf{u} \neq \mathbf{v}$  then for 1/2 of the choices of  $\mathbf{x}$  we have  $\mathbf{u} \odot \mathbf{x} \neq \mathbf{v} \odot \mathbf{x}$
- if  $\mathbf{u} \neq \mathbf{v}$  then  $WH(\mathbf{u})$  and  $WH(\mathbf{v})$  differ on at least half their bits



# Properties (without proof)

## Random subsum principle

- if  $\mathbf{u} \neq \mathbf{v}$  then for 1/2 of the choices of  $\mathbf{x}$  we have  $\mathbf{u} \odot \mathbf{x} \neq \mathbf{v} \odot \mathbf{x}$
- if  $\mathbf{u} \neq \mathbf{v}$  then  $WH(\mathbf{u})$  and  $WH(\mathbf{v})$  differ on at least half their bits

## Local linearity testing

- we say that  $f, g : \{0, 1\}^n \rightarrow \{0, 1\}$  are  $\rho$ -close if

$$\Pr_{\mathbf{x} \in_R \{0,1\}^n} [f(\mathbf{x}) = g(\mathbf{x})] \geq \rho$$

- if there exists a  $\rho > 1/2$  s.t.

$$\Pr_{\mathbf{x}, \mathbf{y} \in_R \{0,1\}^n} [f(\mathbf{x} + \mathbf{y}) = f(\mathbf{x}) + f(\mathbf{y})] \geq \rho$$

then  $f$  is  $\rho$ -close to a linear function

# PCP for Quadeq

Input:  $m \times n^2$  matrix  $A$ ,  $m$  vector  $\mathbf{b}$

Verifier

Proof  $\pi$

---

1. check that  $f, g$  are linear functions
2. check that  $g = WH(\mathbf{u} \otimes \mathbf{u})$  where  $f = WH(\mathbf{u})$
3. check that  $g$  encodes a satisfying assignment

- $\pi \in \{0, 1\}^{2^n + 2^{n^2}}$
- $\pi$  is a pair of linear functions  $\langle f, g \rangle$ , i.e. strings from  $\{0, 1\}^{2^n}$  and  $\{0, 1\}^{2^{n^2}}$ , resp.
- if  $\mathbf{u}$  satisfies  $A(\mathbf{u} \otimes \mathbf{u}) = \mathbf{b}$  then  $f = WH(\mathbf{u})$  and  $g = WH(\mathbf{u} \otimes \mathbf{u})$  are Walsh-Hadamard encodings

# Local linearity testing

- we test the **linearity condition** ( $f(x + y) = f(x) + f(y)$ ) independently  $1/\delta > 2$  times, and accept if **all tests pass**
- we accept a linear function with **probability 1**
- if  $f$  is **not  $1 - \delta$ -close to a linear function**
  - all tests are passed with probability **at most**  $(1 - \delta)^{(1/\delta)}$   
 $\Rightarrow$  such a function is rejected with probability at least  $1 - 1/e > 1/2$
- for instance, we could make a **0.999** linearity test using 1000 trials

# Local decoding

- it might happen, that we accept non-linear functions that are **very close** to linear functions
- in this case we treat them as if they were linear
- if we want to query  $f(\mathbf{x})$ 
  1. we choose  $\mathbf{x}' \in \{0, 1\}^n$  at random
  2. set  $\mathbf{x}'' = \mathbf{x} + \mathbf{x}'$
  3. let  $\mathbf{y}' = f(\mathbf{x}')$  and  $\mathbf{y}'' = f(\mathbf{x}'')$
  4. output  $\mathbf{y}' + \mathbf{y}''$
- this makes **two queries instead of one**
- and recovers the value of the closest linear function with high probability

# PCP for Quadeq

Input:  $m \times n^2$  matrix  $A$ ,  $m$  vector  $\mathbf{b}$

Verifier

Proof  $\pi$

---

1. check that  $f, g$  are linear functions ✓
2. check that  $g = WH(\mathbf{u} \otimes \mathbf{u})$  where  $f = WH(\mathbf{u})$
3. check that  $g$  encodes a satisfying assignment

- $\pi \in \{0, 1\}^{2^n + 2^{n^2}}$
- $\pi$  is a pair of linear functions  $\langle f, g \rangle$ , i.e. strings from  $\{0, 1\}^{2^n}$  and  $\{0, 1\}^{2^{n^2}}$ , resp.
- if  $\mathbf{u}$  satisfies  $A(\mathbf{u} \otimes \mathbf{u}) = \mathbf{b}$  then  $f = WH(\mathbf{u})$  and  $g = WH(\mathbf{u} \otimes \mathbf{u})$  are Walsh-Hadamard encodings

## Check WH encodings

Test 10 times for random  $\mathbf{r}, \mathbf{r}' \in \{0, 1\}^n$

$$f(\mathbf{r})f(\mathbf{r}') = g(\mathbf{r} \otimes \mathbf{r}')$$

## Check WH encodings

Test 10 times for random  $\mathbf{r}, \mathbf{r}' \in \{0, 1\}^n$

$$f(\mathbf{r})f(\mathbf{r}') = g(\mathbf{r} \otimes \mathbf{r}')$$

If the proof is correct we **always accept**:

$$\begin{aligned} f(\mathbf{r})f(\mathbf{r}') &= \left(\sum_{i \in [n]} u_i r_i\right) \left(\sum_{j \in [n]} u_j r'_j\right) \\ &= \sum_{i, j \in [n]} u_i u_j r_i r'_j \\ &= ((\mathbf{u} \otimes \mathbf{u}) \odot (\mathbf{r} \otimes \mathbf{r}')) \\ &= g(\mathbf{r} \otimes \mathbf{r}') \end{aligned}$$

## Check WH encodings

Test 10 times for random  $\mathbf{r}, \mathbf{r}' \in \{0, 1\}^n$

$$f(\mathbf{r})f(\mathbf{r}') = g(\mathbf{r} \otimes \mathbf{r}')$$

If the proof is correct we always accept:

$$\begin{aligned} f(\mathbf{r})f(\mathbf{r}') &= \left(\sum_{i \in [n]} u_i r_i\right) \left(\sum_{j \in [n]} u_j r'_j\right) \\ &= \sum_{i, j \in [n]} u_i u_j r_i r'_j \\ &= ((\mathbf{u} \otimes \mathbf{u}) \odot (\mathbf{r} \otimes \mathbf{r}')) \\ &= g(\mathbf{r} \otimes \mathbf{r}') \end{aligned}$$

If the proof is wrong we reject with probability at least 1/4 by applying the random subsum principle twice, because in essence we compute  $\mathbf{r}U\mathbf{r}'$  and  $\mathbf{r}W\mathbf{r}'$  for different matrices  $U$  and  $W$ .



# PCP for Quadeq

Input:  $m \times n^2$  matrix  $A$ ,  $m$  vector  $\mathbf{b}$

Verifier

Proof  $\pi$

---

1. check that  $f, g$  are linear functions ✓
2. check that  $g = WH(\mathbf{u} \otimes \mathbf{u})$  where  $f = WH(\mathbf{u})$  ✓
3. check that  $g$  encodes a satisfying assignment

- $\pi \in \{0, 1\}^{2^n + 2^{n^2}}$
- $\pi$  is a pair of linear functions  $\langle f, g \rangle$ , i.e. strings from  $\{0, 1\}^{2^n}$  and  $\{0, 1\}^{2^{n^2}}$ , resp.
- if  $\mathbf{u}$  satisfies  $A(\mathbf{u} \otimes \mathbf{u}) = \mathbf{b}$  then  $f = WH(\mathbf{u})$  and  $g = WH(\mathbf{u} \otimes \mathbf{u})$  are Walsh-Hadamard encodings

## Is the assignment satisfying?

- for each of  $m$  equations we can check  $g(\mathbf{z})$  at some place  $\mathbf{z}$  corresponding to the coefficients in matrix  $A$
- but this is **not constant queries!**

## Is the assignment satisfying?

- for each of  $m$  equations we can check  $g(\mathbf{z})$  at some place  $\mathbf{z}$  corresponding to the coefficients in matrix  $A$
- but this is **not constant queries!**
- instead multiply each equation **by a random bit** and take the sum of all equations
- if  $g$  encodes a solution, we will always have a solution to the sum
- otherwise, we have a solution with probability  $1/2$  only

## Is the system in $\text{PCP}[\text{poly}(n), 1]$ ?

1.  $\pi \in \{0, 1\}^{2^n + 2^{n^2}}$
2. check that  $f, g$  are linear functions
  - $2(1 - \delta) \cdot n$  random bits,  $2(1 - \delta)$  queries
3. check that  $g = WH(\mathbf{u} \otimes \mathbf{u})$  where  $f = WH(\mathbf{u})$ 
  - $20n$  random bits,  $20$  queries
4. check that  $g$  encodes a satisfying assignment
  - $m$  random bits (one per equation),  $1$  query

## Is the system in $\text{PCP}[\text{poly}(n), 1]$ ?

1.  $\pi \in \{0, 1\}^{2^n + 2^{n^2}}$
2. check that  $f, g$  are linear functions
  - $2(1 - \delta) \cdot n$  random bits,  $2(1 - \delta)$  queries
3. check that  $g = WH(\mathbf{u} \otimes \mathbf{u})$  where  $f = WH(\mathbf{u})$ 
  - $20n$  random bits,  $20$  queries
4. check that  $g$  encodes a satisfying assignment
  - $m$  random bits (one per equation),  $1$  query

Yes!

# Conclusion

## PCP and hardness of approximation

- computing approximate solutions to **NP**-hard problems is important
- the classical Cook-Levin reduction does not rule out **efficient approximations**
- many nontrivial approximation algorithms exist (2-app for metric **TSP**, knapsack, 2-app for vertex cover)
- **PCP theorem** shows hardness of approximating **max3SAT** to within any constant factor if **P**  $\neq$  **NP**
- we showed hardness of approximation for **Indset** as well
- this is equivalent to having a **probabilistically checkable proof system** with **logarithmic** randomness and **constant** queries
- PCP proofs involve intricate encoding schemes like Walsh-Hadamard

**Further Reading** *Luca Trevisan*, **Inapproximability of Combinatorial Optimization Problems**, available from

<http://www.cs.berkeley.edu/~luca/pubs/inapprox.pdf>

**Next and final topic:** Parallelism