

# Approximation algorithms

## Hardness of approximation (chapter 29)

NTUA, June 2009

# Reductions, gaps and hardness factors

Main technical core is the PCP theorem.

Example:

We can map a Boolean formula  $\varphi$  to a graph  $G=(V,E)$  such that:

- If  $\varphi$  is satisfiable,  $G$  has a vertex cover of size  $\leq \frac{2}{3} |V|$
- If  $\varphi$  is not satisfiable, smallest v.c. is of size  $> a \frac{2}{3} |V|$

Consequence:

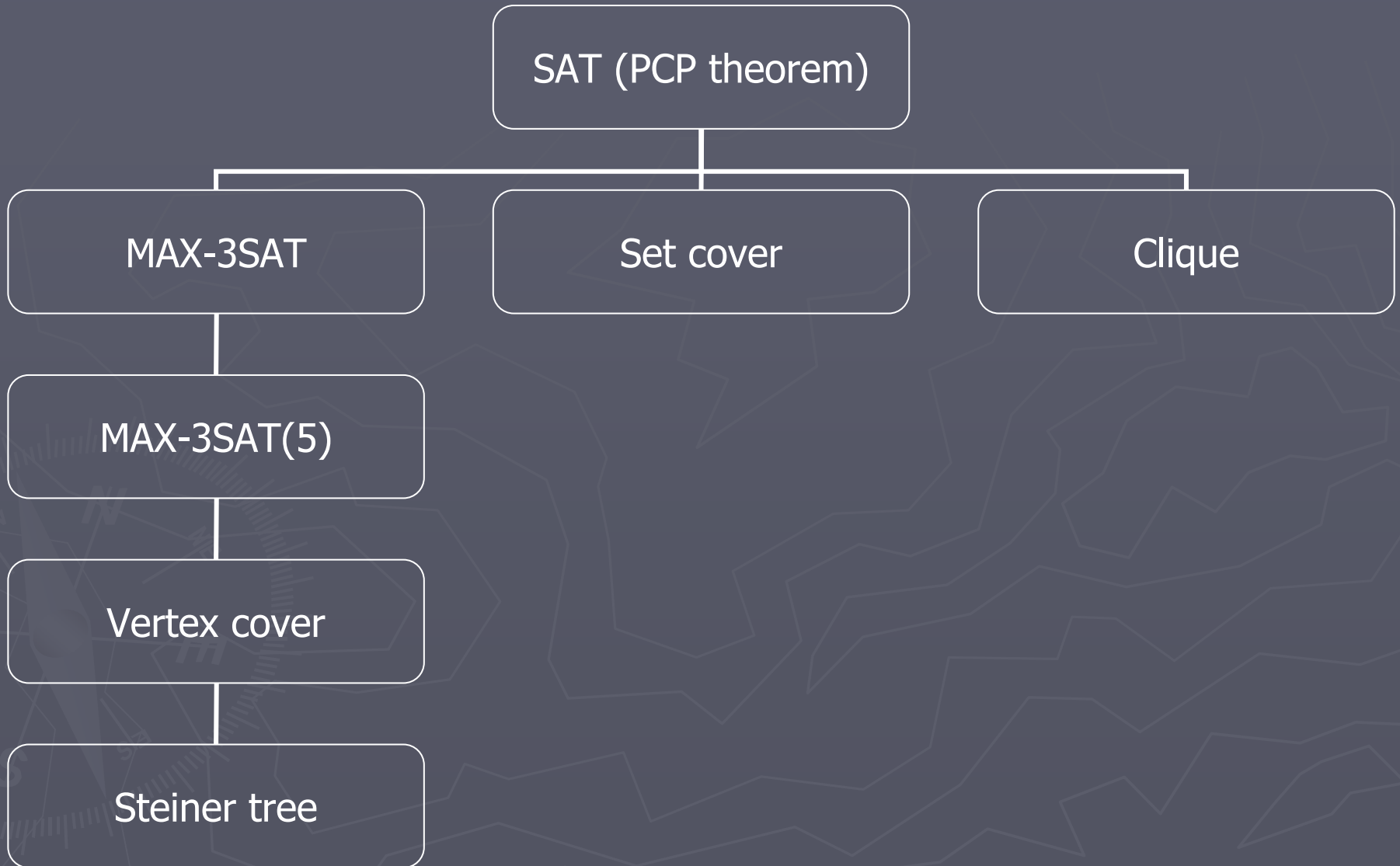
There is no polynomial time algorithm that achieves an approximation guarantee of  $a$  (unless  $P=NP$ ).

# Reductions, gaps and hardness factors

## Definitions

- Let  $\Pi$  be a minimization problem. A gap-introducing reduction comes with parameters  $f$  and  $a$ . In polynomial time maps an instance  $\varphi$  of SAT to an instance  $x$  of  $\Pi$  such that:
  - ▶ If  $\varphi$  is satisfiable,  $OPT(x) \leq f(x)$
  - ▶ If  $\varphi$  is not satisfiable,  $OPT(x) > a(|x|)f(x)$
- Let  $\Pi_1$  be a minimization problem and  $\Pi_2$  be a maximization problem. A gap-preserving reduction comes with parameters  $f_1$ ,  $a$ ,  $f_2$  and  $b$ . In polynomial time maps an instance  $x$  of  $\Pi_1$  to an instance  $y$  of  $\Pi_2$  such that:
  - ▶  $OPT(x) \leq f_1(x) \wedge OPT(y) \geq f_2(y)$
  - ▶  $OPT(x) > a(|x|)f_1(x) \wedge OPT(y) < b(|y|)f_2(y)$

# Figuring results



# The PCP theorem

## The class $NP$

Suppose that there is a verifier checking incoming proofs for prospective strings.

If  $x \in L$  then there exists a proof that makes verifier accept.

If  $x \notin L$  then no proof makes verifier accept.

## The class $PCP(\log n, 1)$

The verifier can read  $O(\log n)$  random bits but only  $O(1)$  bits from the proof.

If  $x \in L$  then there exists a proof that forces verifier to accept with probability 1.

If  $x \notin L$  then on every proof verifier accepts with probability  $< 1/2$ .

# The PCP theorem

1.  $PCP(\log n, 1) \equiv NP$

Proof:

Simulate the PCP verifier for each random string of length  $O(\log n)$ .  
These are polynomially many.  
Accept if and only if all simulations accept, otherwise reject.

2.  $NP \equiv PCP(\log n)$

Proof:

difficult (and omitted!)

PCP theorem:  $NP = PCP(\log n, 1)$ .

# Hardness of MAX-3SAT

Max  $k$ -function SAT:

- $n$  variables
- $m$  functions of  $k$  of the  $n$  variables

Find truth assignment that maximizes the number of satisfied functions

Lemma: There is a constant  $k$  for which there is a gap-introducing reduction from SAT to  $max\ k$ -function SAT transforming  $\varphi$  to an instance  $I$  such that

- If  $\varphi$  is satisfiable,  $OPT(I)=m$
- If  $\varphi$  is not satisfiable,  $OPT(I)<(1/2)m$ .

# Hardness of MAX-3SAT

## Proof:

Let  $V$  be a  $PCP(\log n, 1)$  verifier for SAT.

For each random string  $r$  of length  $c \log n$ ,  $V$  reads  $q$  bits of the proof (a total of at most  $qn^c$  bits).

Introduce one variable for each of these bits

For fixed  $\varphi$  and  $r$  the verifier's answer depends only on the  $q$  bits that will read on the proof tape

For each  $r$  (and fixed  $\varphi$ ) we introduce a function  $f_r$  which is a function of  $q$  variables (there are  $n^c$  such functions).

- If  $\varphi$  is satisfiable there is a proof that makes verifier accept with probability 1 and thus for all the random strings  $r$ ,  $f_r$  is satisfied.
- If  $\varphi$  is not satisfiable then the acceptance probability is  $< 1/2$  which means that  $< 1/2$  of the random strings lead to acceptance. So  $< 1/2$  of the functions are satisfied.



# Hardness of MAX-3SAT

Theorem:

There is a constant  $\varepsilon_M > 0$  for which there is a gap-introducing reduction from *SAT* to *max-3SAT* that transforms a boolean formula  $\varphi$  to  $\psi$  such that:

- If  $\varphi$  is satisfiable,  $OPT(\psi) = m$
- If  $\varphi$  is not satisfiable,  $OPT(\psi) < (1 - \varepsilon_M)m$ .

Proof:

Using previous lemma we can transform the *SAT* formula to an instance of *max k-function SAT*.

Each  $f_r$  can be written as a *SAT* formula  $\psi_r$   
 $\psi$  is the conjunction of these  $\psi_r$ 's.

- If  $\varphi$  is satisfiable then there is a proof that satisfies all the clauses of each  $\psi_r$
- If  $\varphi$  is not satisfiable then for every proof every  $\psi_r$  must have one clause unsatisfied and so  $> (1/2)n^c$  clauses of  $\psi$  unsatisfied.

# MAX-3SAT with bounded occurrence

## Theorem

There is a gap-preserving reduction from *MAX-3SAT* to *MAX-3SAT(29)* that transforms  $\varphi$  to  $\psi$  such that

- If  $OPT(\varphi) = m$ , then  $OPT(\psi) = m'$
- If  $OPT(\varphi) < (1 - \epsilon_M)m$ , then  $OPT(\psi) = (1 - \epsilon_b)m'$

## Proof

For each variable  $x$  of  $\varphi$  that occurs  $k$  times we introduce a new set of  $k$  variables  $x_1, \dots, x_k$  and substitute each occurrence of  $x$  with one of these variables.

Additionally we construct a  $14$ -regular expander  $G$  on  $k$  vertices. We add to the formula the clauses  $(x_i \vee \bar{x}_j)$  and  $(\bar{x}_i \vee x_j)$  for each edge  $(x_i, x_j)$  of  $G$ .

We do this for all the variables of the formula and the resulting formula is  $\psi$ .  
Every optimal assignment for  $\psi$  must assign the same value to "same" variables

- If  $\varphi$  is satisfiable so is  $\psi$
- $OPT(\varphi) < (1 - \epsilon_M)m$  implies  $> \epsilon_M m$  clauses unsatisfied. Using the underlined remark  $\psi$  has  $> \epsilon_M m$  clauses unsatisfied.

# Hardness of Vertex Cover

## Theorem

There is a gap-preserving reduction from  $\max 3SAT(29)$  to  $VC(30)$  that transforms a boolean formula  $\varphi$  to a graph  $G=(V,E)$  such that

- ▶ If  $OPT(\varphi)=m$ , then  $OPT(G) \leq \frac{2}{3}|V|$
- ▶ If  $OPT(\varphi) < (1-\epsilon_b)m$ , then  $OPT(G) > (1 + e_u) \frac{2}{3}|V|$

## Proof

The same reduction for showing NP-completeness.  $G$  has  $3m$  vertices.

Maximum independent set =  $OPT(\varphi)$ :

- ▶ For an optimal truth assignment pick for each satisfied clause a literal that is satisfied. The corresponding vertices form an independent set
- ▶ For a maximum independent set  $I$  satisfy the corresponding literals. The "extension" of this assignment satisfies at least  $|I|$  clauses.

The complement of a max independent set is a minimum vertex cover

- If  $OPT(\varphi)=m$  then  $OPT(G)=2m$
- If  $OPT(\varphi) < (1-\epsilon_b)m$ , then  $OPT(G) > (2 + e_b)m = (1 + \frac{e_b}{2}) \frac{2}{3}|V|$

# Hardness of Steiner tree

## Theorem

There is a gap-preserving reduction from  $VC(30)$  to the Steiner tree problem that transforms an instance of  $G$  of  $VC(30)$  to an instance  $H=(R,S,cost)$  satisfying:

$$OPT(G) \leq \frac{2}{3}|V| \wedge OPT(H) \leq |R| + \frac{2}{3}|S| - 1$$

$$OPT(G) > \frac{2}{3}(1 + e_u)|V| \wedge OPT(H) > (1 + e_s)(|R| + \frac{2}{3}|S| - 1)$$

## Proof

### Vertices of $H$

- ▶ Required:  $r_e$ , one for each edge of  $G$
- ▶ Steiner:  $s_u$ , one for each vertex of  $G$

### Edge costs of $H$

- ▶ between *Steiner* vertices  $cost=1$
- ▶ between *Required* vertices  $cost=2$
- ▶ between *Required* vertex and "incident" *Steiner* vertex  $cost=1$
- ▶ between all other pairs  $cost=2$

$$VC(G) = c \quad \text{ST}(H) = |R| + c - 1$$

# Hardness of Steiner tree

Proof (ctd.)

- ▶ For a vertex cover of size  $c$  let  $S_c$  be the corresponding *Steiner* vertices of  $H$ .  $H$  has a *Steiner tree* with all edges of *cost 1* since each edge is incident to one vertex in  $G$ .

It's total cost is  $|R|+|S|-1=|R|+c-1$

- ▶ Let  $T$  be a Steiner tree of cost  $|R|+c-1$ .

Let  $(u,v)$  be an edge of *cost 2* in  $T$

- Suppose  $u$  is *Steiner*. Remove  $(u,v)$  and add an edge from  $v$  to a *Required* vertex to connect the components. So both  $u, v$  "become" *Required*

- Let  $e_u$  and  $e_v$  be the corresponding edges in  $G$ .

$G$  is connected so there is a path that includes both of them.

Remove  $(u,v)$  disconnecting the tree.

From the path there is a *Steiner* vertex that is connected to both the connected components

Throw in these edges .

$T$  is transformed to have all edges with unit cost having the same total cost.

Thus it has exactly  $c$  *Steiner* vertices. Their corresponding vertices in  $G$  form the required vertex cover of size  $c$

# Hardness of Clique

## Lemma

There is a gap introducing reduction from SAT to Clique transforming  $\phi$  of size  $n$  to a graph  $G$  of  $2^q n^b$  vertices such that

- If  $\phi$  is satisfiable,  $OPT(G) \geq n^b$
- If  $\phi$  is not satisfiable,  $OPT(G) < \frac{1}{2} n^b$

## Proof

Let  $F$  be a  $PCP(\log n, 1)$  verifier for SAT.

For each choice of  $r$  (length  $b \log n$ ), and each truth assignment  $\tau$ , to  $q$  variables we get a vertex, say  $u_{r,\tau}$  (total of  $2^q n^b$  vertices)

We connect vertices that

- ▶ have an  $r$  so that if it “leads” to  $\tau$ , then verifier accepts.
- ▶ their  $\tau$  may be part of the same proof.

If  $\phi$  is satisfiable let  $p(r)$  be the part of the (good) proof that  $r$  “points”. A clique of size  $n^b$ :

$$\{u_{r,p(r)} \mid r \text{ possible random choice}\}$$

If  $\phi$  is not satisfiable then for every proof probability of acceptance is  $< 1/2$ . So  $< (1/2)n^b$  random choices “lead” to acceptance and so  $|largest\ clique| < (1/2)n^b$ .

( If we have a clique  $C$  then there is a proof for all  $\tau$  of the “accepting” vertices. By this proof at least  $|C|$  random choices lead to acceptance. Thus probability at least  $|C|/n^b$ .)

# Generalizing the Verifier

We want something better.

Idea: Why don't we run the verifier more than once to obtain better results. We will need more random bits and read more bits!

The class  $PCP_{c,s}[r(s),q(n)]$

- The verifier can read  $O(r(s))$  random bits and  $O(q(n))$  bits from the proof.
- If  $x \in L$  then there exists a proof, forcing verifier to accept with probability  $\geq c$
- If  $x \notin L$  then on every proof verifier accepts with probability  $< s$ .

Ok. Simulate  $k$  times the verifier:

- reduce soundness (the  $s$ ) to  $< 1/2^k$
- but  $O(k \log n)$  random bits and
- $O(k)$  bits queried

$$NP = PCP_{1,1/n}[\log n, \log n]$$

Proof:

Let  $L \in PCP(\log n, 1)$  decided by verifier  $F$ . If we simulate the verifier  $O(\log n)$  times we have  $O(\log n)$  bits queried but we will need  $O(\log^2 n)$  random bits.

Use expanders.

- Construct an expander with  $n^b$  vertices labeled with  $O(b \log n)$  bits.
- Pick a vertex at random and take a random walk of length  $O(\log n)$ .
- Simulate the verifier  $O(\log n)$  times using as random bits the labels of the vertices of the path.
- Accept iff all simulations accept

If  $x \in L$  then all simulations will accept

If  $x \notin L$  then  $F$  accepts for  $< n^b/2$  random strings. Expanders ensure us that the probability that the path has only "accepting" vertices is  $< 1/n$ .



# (New) Hardness of Clique

## Lemma

There is a gap introducing reduction from SAT to Clique transforming  $\varphi$  of size  $n$  to a graph  $G$  of  $n^{b+q}$  vertices such that

- If  $\varphi$  is satisfiable,  $OPT(G) \geq n^b$
- If  $\varphi$  is not satisfiable,  $OPT(G) < n^{b-1}$

## Proof

Let  $F$  be a  $PCP_{1,1/n}(\log n, \log n)$  verifier for SAT.

For each choice of  $r$  (length  $b \log n$ ), and each truth assignment  $\tau$ , to  $q \log n$  variables we get a vertex, say  $u_{r,\tau}$  (total of  $n^{b+q}$  vertices)

We connect vertices that

- ▶ have an  $r$  that “leads” to  $\tau$  that “leads” to acceptance.
- ▶ their  $\tau$  may be part of the same proof.

If  $\varphi$  is satisfiable let  $p(r)$  be the part of the proof that  $r$  “points”. A clique of size  $n^b$   
 $\{u_{r,p(r)} \mid r \text{ possible random choice}\}$

If  $\varphi$  is not satisfiable then for every proof probability of acceptance is  $< 1/n$ . So  $< n^{b-1}$  random choices “lead” to acceptance and so  $|largest\ clique| < n^{b-1}$ .

# Another characterization for NP

The *two prover one round* model

- There are two proofs (provers, non communicating)
- $O(\log n)$  random bits can be used by the verifier and
- One position of each proof can be queried

The class  $2P1R_{c,s}(r(n))$

$L \in 2P1R_{c,s}(r(n))$  if there is a p.t. verifier that reads  $O(r(n))$  random bits and for every input  $x$

- ▶ If  $x \in L$ , there is a pair of proofs that makes verifier to accept with probability  $\geq c$
- ▶ If  $x \notin L$ , for every pair of proofs verifier accept with probability  $< s$

Theorem

$$NP = 2P1R_{1,1-e}(\log(n)) \quad (\text{for some constant } e > 0)$$

# $NP \stackrel{1,1-\epsilon}{\approx} P1R_{1,1-\epsilon}(\log n)$

## Proof

We can map a boolean formula  $\varphi$  to an instance  $\psi$  of  $Max3Sat(5)$  so that

- ▶ If  $\varphi$  is satisfiable,  $OPT(\psi)=m$
- ▶ If  $\varphi$  is not satisfiable,  $OPT(\psi)<(1-\epsilon)m$

The verifier:

- ▶ does the above reduction
  - ▶ “gets”
    - a proof containing a truth assignment for  $\psi$  and
    - another containing in each position the truth assignments for each clause (encoded)
  - ▶ uses  $O(\log n)$  random bits to pick
    - a clause  $C$
    - a variable  $x$  of the clause
  - ▶ asks
    - first proof for the value of  $x$
    - second proof for the values of the variables of  $C$  (including  $x$ )
  - ▶ accepts iff  $C$  is satisfied and the two assignments of  $x$  agree.
- 
- If  $\varphi$  is satisfiable so is  $\psi$  and so there is a pair of proofs forcing verifier to accept
  - If  $\varphi$  is not satisfiable then suppose  $\tau, z$  the two proofs
    - ▶ at least  $\epsilon m$  clauses unsatisfied by assignment  $\tau$ .
    - ▶  $C$  is unsatisfied with probability  $>\epsilon$  (under  $\tau$ ).
    - ▶ if that is the case and  $z$  satisfies  $C$  then  $\tau, z$  disagree at least at one assignment on the variables of  $C$ .
    - ▶  $V$  catches this with probability  $\epsilon/3$ .