

Approximation Preserving Reductions

Summary

- Memo
- AP-reducibility
 - L-reduction technique
- Complete problems
- Examples: MAXIMUM CLIQUE, MAXIMUM INDEPENDENT SET, MAXIMUM 2-SAT, MAXIMUM NAE 3-SAT, MAXIMUM SAT(B)

Memo: Approximation classes inclusions

- It holds that

$$\text{PTAS} \subseteq \text{APX} \subseteq \text{log-APX} \subseteq \text{poly-APX} \subseteq \text{exp-APX} \subseteq \text{NPO}$$

where

- $\text{log-APX} = \{\text{polynomial-time } O(\log n)\text{-approximate problem}\}$
- $\text{poly-APX} = \{\text{polynomial-time } O(n^k)\text{-approximate problem, for any } k > 0\}$
- $\text{exp-APX} = \{\text{polynomial-time } O(2^{n^k})\text{-approximate problem, for any } k > 0\}$

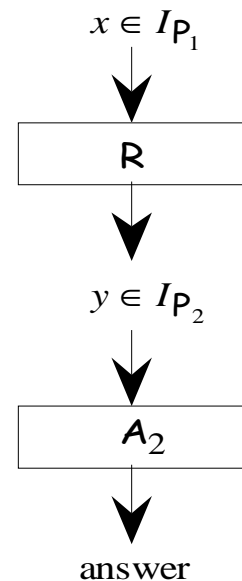
Memo: Approximation classes inclusions

- Polynomial bound on $m()$ implies that any NPO problem is $h2^{nk}$ -approximable for some h and k ...
so every problem is in exp-APX?
- NO! There are problems for which it is even hard (NP-hard) to decide if any feasible solution exists.
 - Example: MIN $\{0,1\}$ -linear programming
 - Given an integer matrix A and an integer vector b , deciding whether a binary vector x exists such that $Ax \geq b$ is NP-hard
- If $P \neq NP$, then
$$PTAS \subset APX \subset \log\text{-APX} \subset \text{poly-APX} \subset \text{exp-APX} \subset \text{NPO}$$

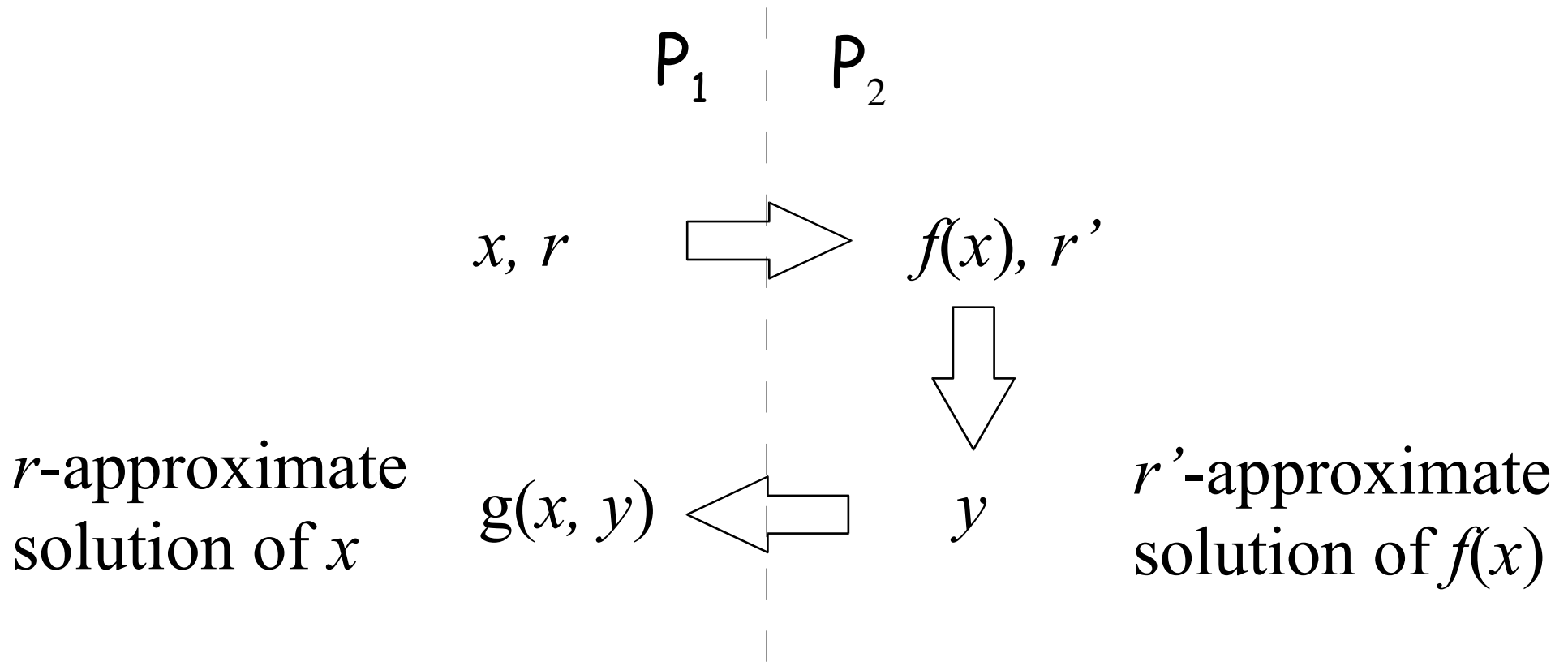
Memo: Karp reducibility

- A decision problem P_1 is *Karp reducible* to a decision problem P_2 (in short, $P_1 \leq P_2$) if there exists a polynomial-time computable function R such that, for any x , x is a YES-instance of P_1 if and only if $R(x)$ is a YES-instance of P_2

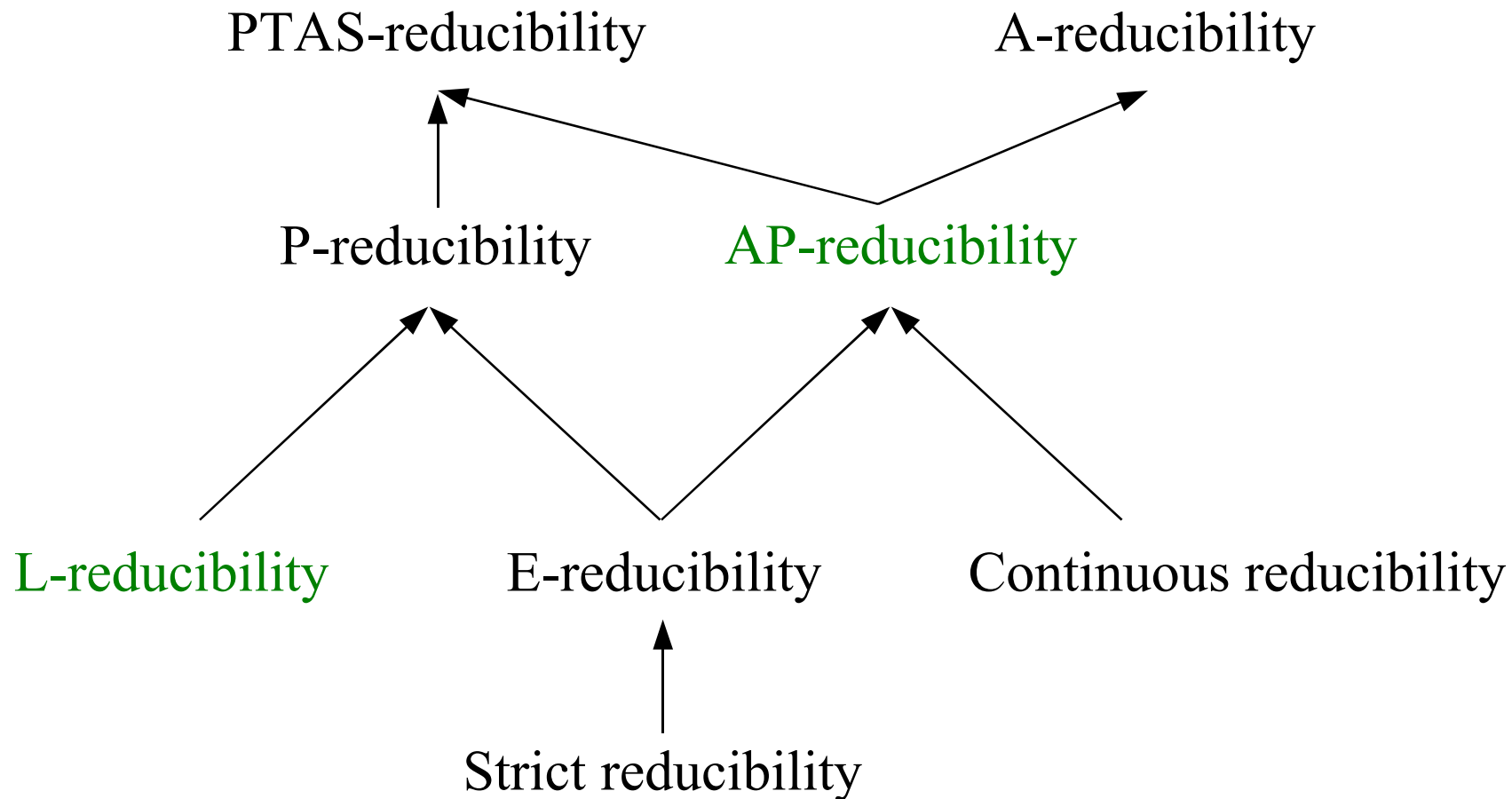
- If $P_1 \leq P_2$ and P_2 is in P , then P_1 is in P



Reducibility and NPO problems



A taxonomy



AP-reducibility (\leq_{AP})

- \mathcal{P}_1 is AP-reducible to \mathcal{P}_2 ($\mathcal{P}_1 \leq_{\text{AP}} \mathcal{P}_2$) if two functions f and g and a constant $c \geq 1$ exist such that:
 - For any instance x of \mathcal{P}_1 and for any $r > 1$, $f(x, r)$ is an instance of \mathcal{P}_2
 - For any instance x of \mathcal{P}_1 , for any $r > 1$, and for any solution y of $f(x, r)$, $g(x, y, r)$ is a solution of x
 - For any fixed $r > 1$, f and g are computable in polynomial time
 - For any instance x of \mathcal{P}_1 , for any $r > 1$, and for any solution y of $f(x, r)$, if $R_{\mathcal{P}_2}(f(x, r), y) \leq r$, then $R_{\mathcal{P}_1}(x, g(x, y, r)) \leq 1 + c(r - 1)$

Basic properties

- **Theorem:** If $P_1 \leq_{AP} P_2$ and $P_2 \in APX$, then $P_1 \in APX$

- If A is an r -approximation algorithm for P_2 then

$$A_{P_1}(x) = g(x, A(f(x, r)), r)$$

is a $(1+c(r-1))$ -approximation algorithm for P_1

Basic properties

- **Theorem:** If $P_1 \leq_{AP} P_2$ and $P_2 \in PTAS$, then $P_1 \in PTAS$

- If A is a polynomial-time approximation scheme for P_2 then

$$A_{P_1}(x, r) = g(x, A(f(x, r'), r'), r')$$

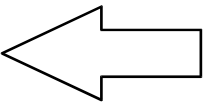
is a polynomial-time approximation scheme for P_1 , where

$$r' = 1 + (r-1)/c$$

L-reducibility

- P_1 is L-reducible to P_2 ($P_1 \leq_L P_2$) if two functions f and g and two constants a and b exist such that:
 - For any instance x of P_1 , $f(x)$ is an instance of P_2
 - For any instance x of P_1 , and for any solution y of $f(x)$, $g(x, y)$ is a solution of x
 - f and g are computable in polynomial time
 - For any instance x of P_1 , $m^*(f(x)) \leq a m^*(x)$
 - For any instance x of P_1 and for any solution y of $f(x)$,
 $|m^*(x) - m(x, g(x, y))| \leq b |m^*(f(x)) - m(f(x), y)|$

Basic property of L-reductions

- **Theorem:** If $P_1 \leq_L P_2$ and $P_2 \in \text{PTAS}$, then $P_1 \in \text{PTAS}$
 - Relative error in P_1 is bounded by $a b$ times the relative error in P_2
- However, in general, **it is not true** that if $P_1 \leq_L P_2$ and $P_2 \in \text{APX}$, then $P_1 \in \text{APX}$  **NO!**
 - The problem is that the relation between r and r' may be non-invertible

Basic property of L-reductions

- **Lemma:** Let P_1 and P_2 be two NPO problems such that is $P_1 \leq_L P_2$.

If $P_1 \in \text{APX}$, then $P_1 \leq_{\text{AP}} P_2$

Inapproximability of independent set

- **Theorem:** $\text{MAX CLIQUE} \leq_{\text{AP}} \text{MAX INDEPENDENT SET}$
 - $G=(V, E)$, $G^c=(V, V^2-E)$ is the complement graph
 - $f(G) = G^c$
 - $g(G, U)=U$
 - $c=1$
 - Each clique in G is an independent set in G^c
- **Corollary:** $\text{MAX INDEPENDENT SET} \notin \text{APX}$

Complete problems

- AP-reduction is transitive
- AP-reduction induces a partial order among problems in the same approximation classes
- Given a class C of NPO problems, a problem P is **C -hard** (with respect to the AP-reduction) if, for any $P' \in C$, $P' \leq_{AP} P$. If $P \in C$, then P is **C -complete** (with respect to the AP-reduction).
 - For any class $C \not\subseteq APX$ ($\not\subseteq PTAS$), if P is C -complete, then $P \notin APX$ ($\notin PTAS$), unless $P=NP$

MAXIMUM WEIGHTED SAT

- INSTANCE: CNF Boolean formula φ with variables x_1, x_2, \dots, x_n , with non negative weights w_1, w_2, \dots, w_n
- SOLUTION: A satisfied truth-assignment f to φ
- MEASURE: $\max(1, \sum w_i f(x_i))$, where $f(x_i)=\text{true}$ is calculated as $f(x_i)=1$ and $f(x_i)=\text{false}$ as $f(x_i)=0$

NPO-complete problems

- Finding a feasible solution is as hard as SAT
 - MAX WEIGHTED SAT \notin exp-APX
- MAX WEIGHTED SAT is NPO-complete
 - MAX WEIGHTED SAT \leq_{AP} MAX WEIGHTED 3-SAT
 - MAX WEIGHTED 3-SAT \leq_{AP} MIN WEIGHTED 3-SAT
 - MIN WEIGHTED 3-SAT \leq_{AP} MIN {0,1}-LINEAR PROGRAMMING
 - MIN {0,1}-LINEAR PROGRAMMING is NPO-complete

APX-complete problems

- The PCP theorem permits to show that MAX 3-SAT is complete for the class of maximization problems in APX
- For any minimization problem P in APX, a maximization problem P' in APX exists such that $P \leq_{\text{AP}} P'$
- MAX 3-SAT is APX-complete

Inapproximability of 2-satisfiability

- **Theorem:** $\text{MAX 3-SAT} \leq_L \text{MAX 2-SAT}$
 - f transforms each clause $(x \text{ or } y \text{ or } z)$ into the following set of 10 clauses where i is a new variable:
 - $(x), (y), (z), (i), (\text{not } x \text{ or not } y), (\text{not } x \text{ or not } z), (\text{not } y \text{ or not } z), (x \text{ or not } i), (y \text{ or not } i), (z \text{ or not } i)$
 - $g(C, t) = \text{restriction of } t \text{ to original variables}$
 - $a=13, b=1$
 - $m^*(f(x)) = 6|C| + m^*(x) \leq 12m^*(x) + m^*(x) = 13m^*(x)$
 - $m^*(f(x)) - m(f(x), t) \leq m^*(x) - m(x, g(C, t))$
- **Corollary:** MAX 2-SAT is APX-complete

MAXIMUM NOT-ALL-EQUAL SAT

- INSTANCE: CNF Boolean formula, that is, set C of clauses over set of variables V
- SOLUTION: A truth-assignment f to V
- MEASURE: Number of clauses that contain both a false and a true literal

Inapproximability of NAE 3-SAT

- **Theorem:** $\text{MAX 2-SAT} \leq_L \text{MAX NAE 3-SAT}$
 - f transforms each clause x **or** y into new clause x **or** y **or** z where z is a new global variable
 - $g(C,t)$ =restriction of t to original variables
 - $a=1, b=1$
 - z may be assumed false
 - each new clause is not-all-equal satisfied iff the original clause is satisfied
- **Corollary:** MAX NAE 3-SAT is APX-complete

Other inapproximability results

- **Theorem:** MIN VERTEX COVER is APX-complete
 - Reduction from MAX 3-SAT(3)

- **Theorem:** MAX CUT is APX-complete
 - Reduction from MAX NAE 3-SAT

- **Theorem:** MIN GRAPH COLORING \notin APX
 - Reduction from variation of independent set

The NPO world if $P \neq NP$

NPO	MINIMUM TSP MAXIMUM INDEPENDENT SET MAXIMUM CLIQUE MINIMUM GRAPH COLORING
APX	MINIMUM BIN PACKING MAXIMUM SATISFIABILITY MINIMUM VERTEX COVER MAXIMUM CUT
PTAS	MINIMUM PARTITION
PO	MINIMUM PATH