

Approximation Algorithms

- Approximation Algorithms for Optimization Problems: Types
- Absolute Approximation Algorithms
- Inapproximability by Absolute Approximate Algorithms
- Relative Approximation Algorithm
- Inapproximability by Relative Approximate Algorithms
- Polynomial Time Approximation Schemes
- Fully Polynomial Time Approximation Schemes

Negative Results for Absolute Approximation

Absolute approximation algorithms are the most desired

- ▷ For large objective values, small additive error is negligible

Generally absolute approximation algorithms exists for problems where the optimal value lie in a small range

- The hardness of such problems is determining the exact value of the optimum solution within this range
- An absolute approximate algorithm finds solution within a small range and uses the fact that the range is small to get a tight guarantee

Not many hard problems have an absolute approximation algorithm

Typically such impossibility of absolute approximation (inapproximability) results use the [scaling method](#)

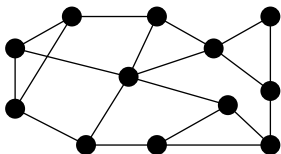
Negative Result by Scaling

Broad idea of scaling

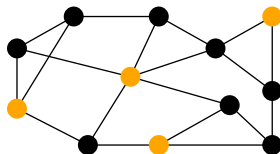
- 1 Scale up certain parameters associated with the instance
- 2 Then show that if there is an absolute approximate algorithm for the scaled up instance, then the solution can be rescaled to get an optimum solution for the original instance
- 3 Conclude that this is an efficient algorithm to solve the NP-HARD optimization problem, which by our assumption of $P \neq NP$ is not possible

Maximum Independent Set Problem

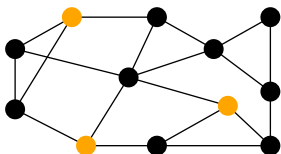
An **independent set** in G is subset of vertices no two of which are adjacent



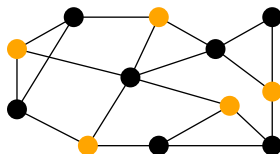
A graph on 12 vertices



An independent set of size 4



An independent set of size 3



An independent set of size 5 (max)

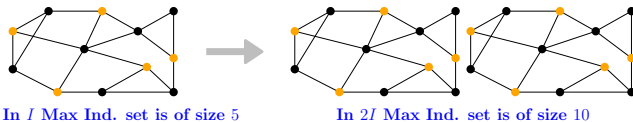
The **MAX-IND-SET(G)** problem (MIS): Find a max independent set in G ?

MIS: Impossibility of absolute approximation

$P \neq NP \implies$ there is no poly-time k -absolute approximation algorithm for MIS

Proof: Suppose there is a k -absolute approximation algorithm \mathcal{A}

Scale the original instance I by a factor of 2 (call this instance $2I$)



Note: $f(\text{OPT}(2I)) = 2f(\text{OPT}(I))$

Run \mathcal{A} on $2I$ to get an ind-set of size $\geq f(\text{OPT}(2I)) - k = 2f(\text{OPT}(I)) - k$

▷ This gives an independent set in I of size $\geq f(\text{OPT}(I)) - k/2$

We got a better algorithm — a $k/2$ -absolute approximate algorithm

Repeat the scaling trick until the approximation guarantee drops below 1

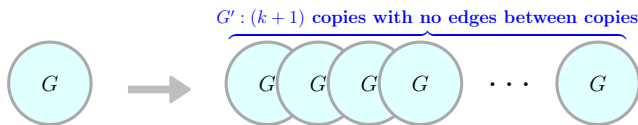
Using integrality of optimal solution we get an optimal solution

MIS: Impossibility of absolute approximation

$P \neq NP \implies$ there is no poly-time k -absolute approximation algorithm for MIS

Proof: Suppose there is a k -absolute approximation algorithm \mathcal{A}

Scale the original instance G by a factor of $(k + 1)$ (call this instance G')



$$\text{Note: } f(\text{OPT}(G')) = (k + 1)f(\text{OPT}(G))$$

\mathcal{A} on G' gives ind-set of size $\geq f(\text{OPT}(G')) - k = (k + 1)f(\text{OPT}(G)) - k$

We get an ind-set in G of size $\geq f(\text{OPT}(G)) - k/k+1 \geq f(\text{OPT}(G))$

Hence we get a maximum independent set in G (of size $f(\text{OPT}(G))$)

Thus, we solved MIS problem in poly-time and proved $P = NP$

The KNAPSACK Problem

Input:

- Items: $U = \{a_1, \dots, a_n\}$ ▷ Fixed order
- Weights: $w : U \rightarrow \mathbb{Z}^+$ ▷ (w_1, \dots, w_n)
- Values: $v : U \rightarrow \mathbb{R}^+$ ▷ (v_1, \dots, v_n)
- Capacity: $C \in \mathbb{Z}^+$

Output:

- A subset $S \subset U$
- Capacity constraint:

$$\sum_{a_i \in S} w_i \leq C$$

- Objective: Maximize

$$\sum_{a_i \in S} v_i$$

KNAPSACK: Impossibility of Absolute Approximation

If $P \neq NP$, then there is no polynomial time k -absolute approximation algorithm for the KNAPSACK problem

Proof: Suppose there is a k -absolute approximation algorithm \mathcal{A}

Consider an instance $I = [U, w, v, C]$, with $v : U \rightarrow \mathbb{Z}^+$

Make an instance $I' = [U, w, v', C]$, $v'(u) = 2k \cdot v(u)$

▷ Note: $f(\text{OPT}(I')) = (2k)f(\text{OPT}(I))$

Run \mathcal{A} on I' to get a $S \subseteq U$ of total capacity $\leq C$ and total value $\geq f(\text{OPT}(I')) - k = 2kf(\text{OPT}(I)) - k$

S is also a solution of I of value (by v) $2kf(\text{OPT}(I)) - k / 2k = f(\text{OPT}(I)) - \frac{1}{2}$

By integrality, S is an optimal solution to I , contradicting $P \neq NP$