

ALGORITHM DESIGN - APPROXIMATION

Approximation Schemes

- Polynomial Time Approximation Scheme (PTAS)
- Design and Example.

APPROXIMATION RATIO VS. TIME COMPLEXITY

- Given an **NP**-complete problem, is there a way to trade-off approximation ratio for time complexity?
 - In practice, even 2 may not be a “good” approximation ratio.
 - In particular, can we find algorithms with an approximation ratio of $1+\epsilon$ where ϵ can be made arbitrarily small?
 - Note that ϵ must be positive:
 - $\epsilon = 0$ would imply an exact solution for a **NP** hard problem.

APPROXIMATION SCHEMES

○ Polynomial Time Approximation Scheme (PTAS):

- An **NP**-complete problem π is said to admit a PTAS if there exists an algorithm A such that
- for any input instance x of π , and for any $r > 1$, $A(x, r)$ returns an r -approximate solution i.e.

- $m(x, A(x, r)) \leq r * m^*(x)$ (assuming minimization)

○ Typically:

- As r decreases (and approaches 1) the time complexity of the algorithm increases.
- Of course, when $r=1$ we expect the algorithm to be have exponential time complexity (unless **P=NP**)

PROBLEM - PARTITION

○ Problem Definition:

- Given a multi-set X of rational numbers, partition the numbers into two sets Y_1 and Y_2 , so that sums of their values are as close as possible.

○ Formal Problem Definition:

- $I = \{ X \mid X \text{ is a multi-set of numbers in } \mathbb{Z}^+ \}$
- $F(X) = (Y_1, Y_2)$ a partition of X
- $m(X, (Y_1, Y_2)) = \max(w(Y_1), w(Y_2))$,
 - where for any set Y , $w(Y) = \sum_{a \in Y} a$
- goal: min

PARTITION – RELATIVE APPROXIMATION

○ Lower Bound (on optimal measure)

- Lemma : $m^*(X) \geq w(X) / 2$
- Proof:
 - For $(Y1, Y2) \in F(X)$ $m(X, (Y1, Y2)) \geq w(X)/2$
 - (by definition of m and partition)

○ Corollary:

- Consider the trivial solution $(X, \{\})$ for any input instance X :
 - $m(X, (X, \{\})) = w(X) \leq 2 * m^*(x)$
- i.e. we only need to consider r -approximation algorithms for values of $r < 2$

PARTITION – R-APPROXIMATION ALGORITHM

○ Greedy_PART(X)

1. Sort items in X in decreasing order to get x_1, x_2, \dots, x_n
2. $Y_1 = Y_2 = \{ \}$
3. for $j = 1$ to n do
 if $((\sum_{a \in Y_1} a) \geq (\sum_{a \in Y_2} a))$
 then $Y_2 = Y_2 \cup \{ x_j \}$
 else $Y_1 = Y_1 \cup \{ x_j \}$
4. return (Y_1, Y_2)

○ Claim:

- Greedy_PART(X) does not yield an optimal solution

○ (Counter)Example:

- Consider the input instance

$\{ 50, 49, 41, 39, 19 \}$

- Optimal solution:

$(\{ 50, 49 \}, \{ 41, 39, 19 \})$

- Greedy_PART(X) will return

$(\{ 50, 39, 19 \}, \{ 49, 41 \})$

PARTITION – R-APPROXIMATION ALGORITHM - ANALYSIS

○ Lemma 1:

- If Greedy_PART(X) returns $(Y1, Y2)$ such that $|Y1| = 1$ and $w(Y1) \geq w(Y2)$ then $(Y1, Y2)$ is the optimal solution for X.
- Proof: Trivial

○ Lemma 2:

- Let Greedy_PART(X) returns $(Y1, Y2)$ such that $|Y1| \geq 2$ and $w(Y1) \geq w(Y2)$. If the last item added to Y1 is x_h and the rank of x_h is k then $x_h \leq w(X)/k$
- Proof:
 - If rank of x_h was k then at least k items each $\geq x_h$ have been added
 - i.e. X includes at least k items each $\geq x_h$.
 - i.e. $w(X) \geq k * x_h$
 - i.e. $x_h \leq w(X)/k$

PARTITION – R-APPROXIMATION ALGORITHM – ANALYSIS [2]

○ **Lemma 3:** If Greedy_PART(X) returns $(Y1, Y2)$

then $m(X, (Y1, Y2)) \leq (3/2) * m^*(X)$

○ **Proof:** Assume without loss of generality:

○ $w(Y1) \geq w(Y2)$ and x_h is the last item added to $Y1$

• Then: $w(Y1) - x_h \leq w(Y2)$

○ i.e. $2 * w(Y1) - x_h \leq w(X)$

○ i.e. $2 * w(Y1) \leq x_h + w(X)$ [E1]

1. If $|Y1| = 1$, then $m(X, (Y1, Y2)) / m^*(X) = 1$ by Lemma 1.

2. If $|Y1| \geq 2$ then

○ $m(X, (Y1, Y2)) / m^*(X) \leq w(Y1) / m^*(X)$

○ $\leq 2 * w(Y1) / w(X)$ [by lower bound]

○ $\leq 1 + x_h / w(X)$ [by E1]

○ $\leq 1 + 1/2$ [by Lemma 2]



Tighter result: $1 + 1/k$, where k is rank of last item added to larger set.

PARTITION – R-APPROXIMATION ALGORITHM – ANALYSIS [3]

- Lemma 4:

- Greedy_PART is a polynomial time $3/2$ – approximate algorithm for PARTITION.

- Proof:

- $m(X, \text{Greedy_PART}(X)) / m^*(X) \leq 3/2$ for any X by Lemma 3
- Time complexity of Greedy_PART is $O(N \cdot A(N) + N \cdot \log N)$
 - where $A(N)$ is the time taken to add an element to a set of size N

PTAS – EXAMPLE – PARTITION

- Assume EXACT_PART is an exact algorithm for PARTITION
 - For instance, it may obtain all subsets $Y1$ of the given input set X and find the one with minimum measure among all $(Y1, X - Y1)$
- PTAS_PART(X, r)
 1. If $r \geq 2$ then return $(X, \{ \})$
 2. Sort items in X in decreasing order to get x_1, x_2, \dots, x_n
 3. $k = \text{fpart}(r, n)$ // fpart returns an integer in the range $1..n$
 4. $(Y1, Y2) = \text{EXACT_PART}(\{x_1, \dots, x_k\});$
 5. for $j = k+1$ to n do
 - if $((\sum_{a \in Y1} a) \geq (\sum_{a \in Y2} a))$ then $Y2 = Y2 \cup \{x_j\}$
 - else $Y1 = Y1 \cup \{x_j\}$
 6. return $(Y1, Y2)$

PTAS – EXAMPLE – PARTITION

[2]

○ Lemma 5: If PTAS_PART(X) returns (Y1,Y2)

then $m(X, (Y1,Y2)) \leq (1 + (1/(1 + fpart(r,n)))) * m^*(X)$

○ Proof:

- Assume without loss of generality:

- $w(Y1) \geq w(Y2)$ and x_h is the last item added to $Y1$

- Then:

1. If x_h was added to $|Y1|$ by EXACT_PART then the solution is optimal.

2. Otherwise: $x_h * (fpart(r,n) + 1) \leq w(X)$ [E3]

1. Modify Lemma 3 as follows:

In the last step (application of Lemma 2), use E3.

2. Then $m(X, (Y1,Y2)) / m^*(X) \leq 1 + (1/(1 + fpart(r,n)))$

PTAS – EXAMPLE – PARTITION

[10]

- **Theorem:** PTAS_PART is a Polynomial Time Approximation Scheme for PARTITION
- **Proof:**
 - By lemma 5,
 - $m(X, \text{PTAS_PART}(X)) \leq (1 + (1/(1 + f_{\text{part}}(r)))) * m^*(X)$
 - Then let $1 + (1/(1 + f_{\text{part}}(r))) = r$
 - i.e. define $f_{\text{part}}(r) = (1/(r-1)) - 1$
 - i.e. PTAS_PART(X,r) with $f_{\text{part}}(r) = (1/(r-1)) - 1$ will be an r-approximate algorithm for PARTITION.
 - Time Complexity of PTAS_PART:
 - $O(n \log n)$ for sorting + $O(2^{f(r)})$ for EXACT_PART + $O(n * A(n))$ for greedy part
 - i.e. $O(n \log n + 2^{f(r)})$ i.e. $O(n \log n + 2^{O(1/(r-1))})$