

COMPLEXITY – OPTIMIZATION PROBLEMS

Approximation Algorithms

- Relative Approximation
- Design Techniques
 - Greedy Method
 - Sequencing
 - Randomization

RELATIVE APPROXIMATION

- Given an optimization problem π , for any input instance x and for any feasible solution y , the performance ratio of y is defined as:
 - $R(x,y) = \max(m(x,y)/m^*(x), m^*(x)/m(x,y))$
- Given an optimization problem π , an algorithm A is said to be an r -approximate algorithm if there exists a constant r such that
 - i.e. for any x in I_π $R(x,A(x)) \leq r$
- Example:
 - Greedy_Vertex_Cover is a 2-approximation algorithm

PROBLEM – BIN PACKING

○ Problem Definition: BIN PACKING

- Given a set of N items each with values S_1, S_2, \dots, S_n , distribute them into equal-sized bins such that the number of bins required is minimum.

○ Application:

- Memory allocation problem (e.g.):
 - Memory is available as fixed-size blocks (i.e. bins)
 - Memory requests come in different sizes

○ Note:

- As the bins are equal-sized, we can assume that they are of unit size and scale the values S_1, S_2, \dots, S_n accordingly.

PROBLEM — BIN PACKING

[2]

Formal Problem Definition: BIN PACKING

- $I = \{ S \mid S \text{ is a finite multi-set of } n \text{ rational numbers in } (0,1] \}$
- $F(S) = \text{a partition } \{ B_1, B_2, \dots, B_k \} \text{ of } S$
s.t. $\sum_{a \in B_j} a \leq 1$ for each j
- $m(S, \{B_1, B_2, \dots, B_k\}) = k$
- goal = min

Definition: Partition

- $\{ B_1, B_2, \dots, B_k \}$ is said to be a partition of S if
- $\bigcup B_j = S$ and $\bigcap B_j = \{\}$

RELATIVE APPROXIMATION – EXAMPLE – BIN PACKING

- Lower Bound (on the optimal solution):
 - Given input instance S , let $A = \sum_{a \in S} a$
 - Claim:
 - $m^*(x) \geq \text{ceil}(A)$
 - Proof:
 - Trivial (Perfect packing)
- Algorithm Next_Fit (S)
 1. $i=0$;
 2. for each item a in S :
 - if $((a + \sum_{b \in B_i} b) \leq 1)$ { assign a to B_i }
 - else { assign a to B_{i+1} ; $i = i+1$; }

RELATIVE APPROXIMATION – EXAMPLE – BIN PACKING [2]

○ Comments:

- The technique used by NEXT_FIT is referred to as *Sequencing*.
- NEXT_FIT is an online algorithm.

○ Theorem:

- NEXT_FIT is a polynomial time 2-approximate algorithm for BIN PACKING.
- Proof:
 - For each pair of consecutive bins
 - the sum of values assigned to these two bins is > 1
 - i.e. $\# \text{ bins used} / 2 \leq A$ where $A = \sum_{a \in S} a$
 - $m_{\text{NEXT_FIT}}(S) \leq 2 * \text{ceil}(A) \leq 2 * m^*(S)$

RELATIVE APPROXIMATION – EXAMPLE – BIN PACKING [3]

○ Claim:

- The approximation ratio of 2 is asymptotically tight for NEXT_FIT
- Proof:
 - Given any integer n , consider the instance with $4n$ items
$$S = \{ 1/2, 1/2n, 1/2, 1/2n, \dots, 1/2, 1/2n \}$$
 - Optimal solution would require: $n+1$ bins
 - NEXT_FIT would require: $2*n$ bins

○ Question:

- What is the weakness of NEXT_FIT ?

RELATIVE APPROXIMATION – BIN PACKING - FIRST FIT

○ Algorithm FIRST_FIT (S):

1. $j=0$; // $j+1$ is the number of bins; initially there is 1 bin
2. for each item a in S {
 1. $i = 0$;
 2. while ($i \leq j$) {
 1. if $((a + \sum_{b \in B_i} b) \leq 1)$ { assign a to B_i ; break; }
 2. else { $i = i+1$; }}
 3. if ($i > j$) { $j=j+1$; assign a to B_j ; }}

Exercise:

Compare the running time of NEXT_FIT w. that of FIRST_FIT.

RELATIVE APPROXIMATION – BIN PACKING – FIRST FIT [2]

- Claim (w/o Proof):

- $m_{\text{FIRST_FIT}}(S) \leq 1.7 * m^*(S) + 2$

- Question: When is this better than NEXT_FIT?

- Comments:

- FIRST_FIT is also an online algorithm

- Algorithm FIRST_FIT_DEC(S) // An offline algorithm

1. Sort S in non-increasing order
2. FIRST_FIT(S)

- How does FIRST_FIT_DEC perform?

RELATIVE APPROXIMATION – BIN PACKING – FIRST FIT [3]

○ Claim (w.o. proof):

- $m_{\text{FIRST_FIT_DEC}}(S) \leq (11 * m^*(S))/9 + 4$

○ Claim:

- The bound given above is tight.

○ Proof:

- For any +ve integer n , define an instance x_n of $5*n$ items:

- n items of size $1/2 + \epsilon$
- n items of size $1/4 + 2*\epsilon$
- n items of size $1/4 + \epsilon$
- $2*n$ items of size $1/4 - 2*\epsilon$

Optimal Packing: ($\lceil 3 * n/2 \rceil$ bins)

- n bins each w. one of a, c, and d
- $\lceil n/2 \rceil$ bins each w. two of b and d

$$m_{\text{FIRST_FIT_DEC}}(x_n) = 11*n/6$$
$$m^*(x) = 3*n/2$$

FIRST_FIT_DEC Packing:

- n bins each w. one of a and b
- $\lceil n/3 \rceil$ bins each w. three of c
- $\lceil n/2 \rceil$ bins each w. four of d

RELATIVE APPROXIMATION – BIN PACKING – FIRST FIT [4]

- Question:

- What is the weakness of FIRST_FIT (and that of FIRST_FIT_DEC)?

- BEST_FIT attempts to reduce fragmentation

- Algorithm BEST_FIT_DEC(S) // An offline algorithm

1. Sort S in non-increasing order
2. BEST_FIT(S)

RELATIVE APPROXIMATION – BIN PACKING – BEST FIT

Algorithm BEST_FIT (S):

1. $j=0$;
2. for each item a in S {
 1. $i = 0$; $f = -1$; $\text{minGap} = \text{MAX_INT}$;
 2. while ($i \leq j$) {
 1. $w = (a + \sum_{b \in B_i} b)$
 2. if ($w \leq 1$) { if ($\text{minGap} > 1-w$) { $\text{minGap} = 1-w$; $f=i$; }}
 3. $i = i+1$;}}
if ($f=-1$) { $j=j+1$; assign a to B_j ; } else { assign a to B_f ; }
}

Compare the running time of BEST_FIT_DEC with that of FIRST_FIT_DEC.

RELATIVE APPROXIMATION – BIN PACKING – BEST FIT [2]

○ Exercise:

- Compare the running time of BEST_FIT_DEC with that of FIRST_FIT_DEC.

○ Claim :

- For any instance S of BIN_PACKING,

$$m_{\text{FIRST_FIT_DEC}}(S) \geq m_{\text{BEST_FIT_DEC}}(S)$$

- Exercise: Prove this claim!

RELATIVE APPROXIMATION – BIN PACKING – BEST FIT [3]

○ Claim :

- There exists instances S of BIN_PACKING such that $m_{\text{FIRST_FIT_DEC}}(S) > m_{\text{BEST_FIT_DEC}}(S) = m^*(S)$:

- Construct such instances!

○ Claim:

- The approximation ratio of BEST_FIT_DEC is the same as that of FIRST_FIT_DEC.
 - The tight example works for this as well!
 - Combine this with the claim on the measures of the two solutions (*see previous slide*)

APPROXIMATION BY RANDOMIZATION : EXAMPLE - MAXIMUM SATISFIABILITY

○ Definition:

- Instances: $I = \{ C \mid C \text{ is a set of disjunctive clauses on a set of variables } V \}$
- $SOL(C) = f$, s.t. $f : V \rightarrow \{ 0,1 \}$
- $m(C,f) = \# \text{ clauses satisfied by } f$
- Goal = max

○ Algorithm Random_Satisfiability (RS)

- for each v in V ,
 - independently set $f(v) = 1$ with probability $\frac{1}{2}$
- return f ;

APPROXIMATION BY RANDOMIZATION : EXAMPLE - MAXIMUM SATISFIABILITY

- Algorithm Random_Satisfiability (RS)
 - for each v in V ,
 - independently set $f(v) = 1$ with probability $\frac{1}{2}$
 - return f ;
- Measure of the Optimal Solution:
 - The expected measure of a solution is given by
$$E[m_{RS}(C)] = (1 - (1/2^k))^* |C|$$

where each clause in C has at least k literals.
- Exercises:
 - Prove the above claim.
 - Find an upper bound for $m^*(x) / E[m_{RS}(C)]$
 - A trivial upper bound for the optimal measure is $|C|$