

Assignment 1 - Homework Exercises on Approximation Algorithms

Pattarawat Chormai, Duy Pham, Maciej Wikowski

September 13, 2015

A.I-1

We will show that the approximation ratio of the *GreedySchedulingAlgorithm* is at least $2 - \frac{1}{m}$ by showing an example as follow.

Let's consider this setting:

- 3 machines: M1, M2, M3
- 7 jobs: 1, 1, 1, 2, 1, 1, 2

GreedySchedulingAlgorithm will come up with this scheduling:

- M1: 1 2 2
- M2: 1 1
- M3: 1 1

Thus $ALG = makespan = 1 + 2 + 2 = 5$

We know that:

$$OPT \geq Average_{load} \tag{1}$$

Where $Average_{load} = \frac{1}{m} \sum_{i=1}^n j_i = \frac{9}{3} = 3$

Here we can find a solution with $makespan = 3$. That is:

- M1: 1 2
- M2: 1 2

- M3: 1 1 1

Therefore, $OPT = 3$

Thus, the approximation ratio is:

$$\rho = \frac{ALG}{OPT} = \frac{5}{3} \quad (2)$$

According to the theorem, the estimated ratio is:

$$\rho_{estimated} = 2 - \frac{1}{m} = 2 - \frac{1}{3} = \frac{5}{3} \quad (3)$$

From 2 and 3, we have $\rho_{estimated} = \rho$. Therefore, this bound is tight.

A.I-2

From the question, we know that

$$\begin{aligned} m &= 10 \\ \sum_{j=1}^n t_j &\geq 1000 \\ t_j &\in [1, 20] ; \text{ for all } i \leq j \leq n \end{aligned}$$

Let T'_{i^*} denote the load of M_i before t_j^* , last job, is assigned to the machine. Thus T_i^* , which represents makespan of the assignment, equals to

$$T_i^* = T'_{i^*} + t_j^*$$

Because T'_{i^*} is the minimum load among all machines, so that we can derive

$$T'_{i^*} \leq \frac{1}{m} \sum_{i=1}^m T'_i = \sum_{j=1}^{j^*} t_j \leq \frac{1}{m} \left[\sum_{j=1}^n t_j - t_j^* \right] \leq LB$$

Then we can derive

$$\begin{aligned}
T_i^* &= T_{i^*}' + t_j^* \\
&\leq \frac{1}{m} \left[\sum_{j=1}^n t_j - t_j^* \right] + t_j^* \\
&\leq \frac{1}{m} \sum_{j=1}^n t_j + \left(1 - \frac{1}{m}\right) t_j^* \\
&\leq 100 + \left(1 - \frac{1}{10}\right) 20 \\
&\leq 118
\end{aligned}$$

According *Algorithm Greedy Scheduling* and the question, we know

$$\begin{aligned}
\max \left(\frac{1}{m} \sum_{j=1}^n t_j, \max_{1 \leq j \leq n} (t_j) \right) &\leq LB \leq OPT \\
\max_{1 \leq j \leq n} (t_j) &= 20
\end{aligned}$$

Then we can derive

$$\max \left(\frac{1}{m} \sum_{j=1}^n t_j, 20 \right) \leq LB$$

Since $\frac{1}{m} \sum_{j=1}^n t_j \geq 1000$, thus

$$\begin{aligned}
100 &\leq LB \\
&\leq OPT
\end{aligned}$$

Therefore, approximation-ratio(ρ) equals to

$$\begin{aligned}
T_i^* &\leq \rho OPT \\
\frac{118}{100} &\leq \rho \\
1.18 &\leq \rho
\end{aligned}$$

For this particular setting, *Algorithm Greedy Scheduling* is 1.18 approximation algorithm.

AI-3-i)

Assume we have the optimal solution, which has n squares

$$n \leq LB \leq OPT$$

Lemma 1. *Each unit square in the grid can overlap at most 4 cells. Let n_s be the number of square in the integer grid solution. Thus*

$$n_s \leq 4n \leq 4OPT$$

A.I-3-ii

We propose the algorithm as follow.

Algorithm 1 Finding minimum row square cover

Require: Set of Points P

Ensure: Min Square Cover min

Operation:

```
set currentCoveringPosition = 0
QuickSortAscending( $S$ )
for all Point  $p$  in  $P$  do
  if  $p.x \leq \textit{currentCoveringPosition}$  then
    create square  $s = (p.x, 1, p.x + 1, 0)$ ;
    add  $s$  to  $S$ 
    set currentCoveringPosition =  $p.x + 1$ 
  end if
  set  $min = \textit{sizeof} S$ 
  return  $min$ 
end for
```

This algorithm is correct because:

- Every point in p will be covered by a square
- There are no intersections between the squares because we traverse in one direction

This algorithm consists of 2 parts: QuickSort and Traversing the Point to create squares. Let t be the run time of this algorithm, $t_{quicksort}$ be the time for quick-sort, and t_{assign} be the time for creating the squares. We have:

$$t = t_{quicksort} + t_{assign} \leq n \log n + n = O(n \log n) \quad (4)$$

Thus the runtime of this algorithm is $O(n \log n)$.

A.I.3.iii

The idea of our algorithm is that, we put all the points in to a coordinate system, then we divide the coordinate system into a set of unit rows (i.e. rows with height 1). For each row, we use algorithm 1 to find the minimum size square-cover. The global min-square-cover is the sum of all row-square-cover.

Algorithm 2 Finding global minimum square cover

Input: Set of points P

Output: Min Square Cover min

Operation:

```

currentMin = 0;
for all Row  $r$  in the space do
    currentMin += FindRowMinSquare()
end for
set  $min = currentMin$ 
return min

```

Theorem. FindingGlobalMinimumSC is 2 – approximation

Proof. Suppose we have the optimal solution with n squares. Then $OPT \geq n$, thus $LB = n$. We know that each unit square in the coordinate system can overlap with at most 2 unit rows.

Let min be our algorithm solution, then:

$$min \leq 2n = 2.LB \leq 2.OPT \quad (5)$$

Therefore, this algorithm is 2 – approximation.

□

AII.1

(i)

We prove this statement by contradiction.

- Suppose that $V \setminus C$ is not an independent set of G . Then there exists a pair of vertices (u, v) in $V \setminus C$ which are connected by an edge $e \in E$. Thus, both u and v are not in C . Therefore, C is not the vertex cover of G anymore.
- Suppose C is not the vertex cover of G , then there exists a pair of vertices (u, v) that are connected by an edge $e \in E$ but are not in C . Thus, $u \in (V \setminus C)$ and $v \in (V \setminus C)$. Therefore, $(V \setminus C)$ is not the vertex cover of G anymore.

From the reasoning above, we can state that: C is the vertex cover of G if and only if $V \setminus C$ is an independent set of G .

(ii)

We prove that *ApproxMaxIndependentSet* is not a 2-approximation algorithm by showing a counter example. That is, consider a complete graph, for example, a graph $G = (V, E)$ where $V = x_1, x_2$ and $E = (x_1, x_2)$.

Applying the *ApproxMinVertexCover*(G), we get $C = x_1, x_2$ (picking both vertices from the edge).

Now we take the approx max independent set $ALG = V \setminus C = \emptyset$.

The optimal solution now is $OPT = 1$ (picking x_1 or x_2).

The approximation ratio is $\rho = \frac{OPT}{ALG} = \infty \neq 2$.

So the approximation ratio is not 2.

AII.2

(i)

Considering the input we can define the best possible scenario for the solution as being a CNF formula where a single element picked from the first clause can be found in all other clauses. The algorithm then would remove all other

clauses. This leads us to the assumption that the optimal value OPT in our case is equal to $OPT = 1$

Since no assigned method is specified as to which element of the clause we should choose, we assume a random pick. This leads us to the worst possible choice scenario of a variable existing in all the clauses of the CNF and algorithm always picking the wrong one for the search.

$$(x_1 \vee x_4 \vee x_5) \wedge (x_3 \vee x_1 \vee x_6) \wedge (x_2 \vee x_1 \vee x_7)$$

If the algorithm picked in the first clause x_4 as the variable to look for and x_3 too look for in the second clause it would return a value of one element in each clause until the last clause leaving us with an $ALG = \frac{n}{3} - 1$. Since all elements are different but they share one element in all clauses we pick 1 element in each clause meaning $\frac{1}{3}$ of all used elements with exclusion of the first clause. This would however vary greatly from the most optimal solution of $OPT = 1$ as described previously.

That leads to the approximation ratio of :

$$1 \leq \rho \leq \frac{n}{3} - 1$$