
Complexity and Advanced Algorithms

Spring 2021

Approximation Algorithms – A Brief Introduction

Approximation Algorithms

- Suppose a problem is known to be NP-Complete.
- No hope to solve in polynomial time unless $P = NP$.
- But, several practical problems fall in this category.
- Need some solution to this issue.
- This is where approximation algorithms help.

Approximation Algorithms

- For a problem P , let A be an approximation algorithm.
- Suppose that the problem P is a minimization problem.
- Then, the performance of Algorithm A for P is measured as its approximation ratio defined as follows.
- For an instance I of P , let $OPT(I)$ denote the best possible solution.
Handwritten notes: "value of the optimal" with an arrow pointing to $OPT(I)$; "negative of" with an arrow pointing to $A(I)$.
- Let $A(I)$ denote the solution produced by the algorithm A .
- The approximation ratio of algorithm A is $\max_I |A(I)|/|OPT(I)|$.
- Notice that the ratio is always at least 1.

$$OPT(I) \leq A(I)$$

length(Tour)

Alg

return Tour

Approximation Algorithms

- For a problem P , let A be an approximation algorithm.
- Suppose that the problem P is a maximization problem.
- Then, the performance of Algorithm A for P is measured as its approximation ratio defined as follows.
- For an instance I of P , let $OPT(I)$ denote the best possible solution.
- Let $A(I)$ denote the solution produced by the algorithm A .
- The approximation ratio of algorithm A is $\max_I |OPT(I)|/|A(I)|$.
- Notice that the ratio is always at least 1.

$$OPT(I) \geq A(I)$$

Approximation Algorithms

- We have seen an example of this definition earlier.
- In the context of MAXSAT.
- Today, we will study two more problems and approximation algorithms for them.

max $f(I)$

s.t. ,

best possible $I : f(I) = 20$

for any alg. $f(A(I)) \leq 20$

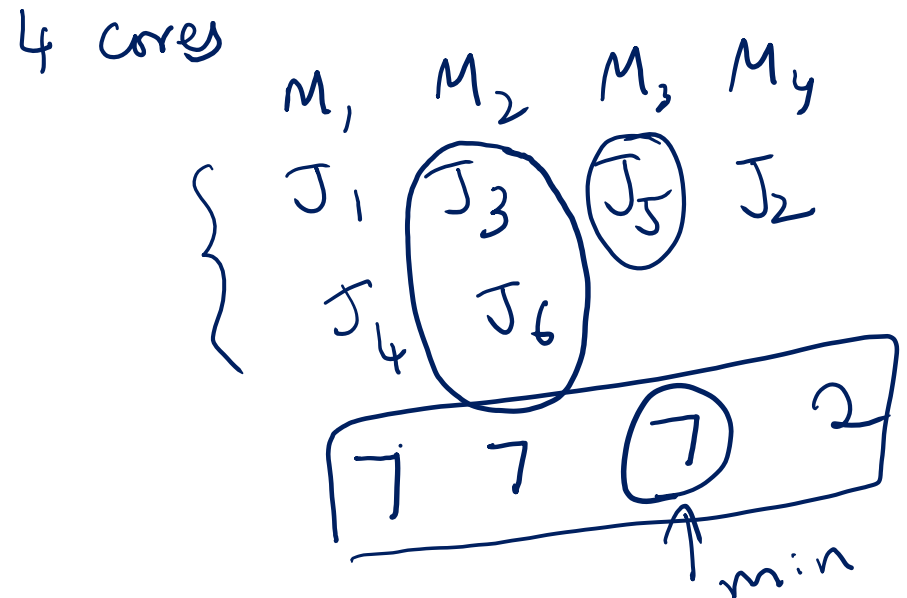
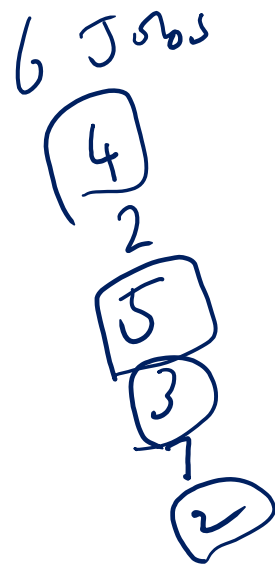
for any input, if $\frac{|OPT(I)|}{|A(I)|} \leq \frac{3}{2}$

$|A(I)| \geq \left(\frac{2}{3}\right) |OPT(I)|$

\Rightarrow $\frac{2}{3}$ for MAXSAT

Load Balancing

- Consider m machines M_1, M_2, \dots, M_m , that are identical in all respects.
 - Like the m cores of your multicore computer.
- We have n jobs, J_1, J_2, \dots, J_n to be processed and any job can be processed by any machine.
- Our goal is to minimize the time spent by any machine.



Load Balancing

- We define the following quantities.
- Let $A(i)$ be the jobs assigned to machine M_i .
- Job j has a time requirement t_j , for $j = 1$ to n .
- The makespan of machine M_i is $T_i := \sum_{j \in A(i)} t_j$.
- The makespan of an assignment $T = \max_i T_i$.
- The goal of the problem is to find an assignment that minimizes the makespan.

Load Balancing

- One of the popular algorithms is to use a greedy technique.
- We can assume that all the jobs are given apriori.
 - A bit unlike the real world setting where user jobs are fired any time.
- Assign the next job to the machine that is presently least loaded.
 - Note that all jobs are given at the beginning.
 - Assignment is done before any machine starts its processing.

Load Balancing

• GreedyAssign(Jobs J, m)

• begin

$J_1 \quad J_2 \quad J_3 \quad \dots \quad J_n$
4 2 3 5

• Set $A(i)$ = empty for all i between 1 to m .

• Set T_i = 0 for all i between 1 to m

• For $j = 1$ to n do

$M_1 \quad M_2 \quad \dots \quad M_m$

• Find an index i such that machine M_i has minimum T_i

• $A(i) = A(i) \cup \{j\}$

• $T_i = T_i + t_j$ ↖ load

• Endfor

• End

$\left[\begin{array}{l} j: \text{jobs} \\ i: \text{machines} \end{array} \right]$

~~load~~

M_1	M_2
4	2
5	3
9	5

Load Balancing

- Illustrate how this algorithm works on the following set of jobs and four machines.

- Runtimes of jobs = 2, 3, 2, 2, 4, 1, 2, 1.
 Some machine makespan ≥ 4
max makespan ≥ 4

- Find the best possible makespan and the makespan produced by the greedy algorithm.

- For $j = 1$ to n do
 - Find an index i such that machine M_i has minimum T_i
 - $A(i) = A(i) \cup \{j\}$
 - $T_i = T_i + t_j$
- Endfor

greedy : 6
best : 5

Load Balancing

- We start with two observations that help us prove the approximation ratio of the greedy algorithm.

- Let T^* denote the best possible makespan. (ex: $T^* = 5$)

- Observation 1:** $T^* \geq (1/m) \sum_{j=1}^n t_j$.
 $T^* > \frac{15}{4} = 15$ (ex.)
- Comes from the fact that there is a total work of $\sum_{j=1}^n t_j$ and some machine will have to work for at least a $1/m$ fraction of the total.

- Observation 2:** $T^* \geq \max_j t_j$

J_1	J_2	J_3	J_4	J_5	$m = 3$
1	1	100	1	1	
$\sum_{j=1}^n t_j = 100 + 4$					$T^* \geq \frac{100 + 4}{3}$

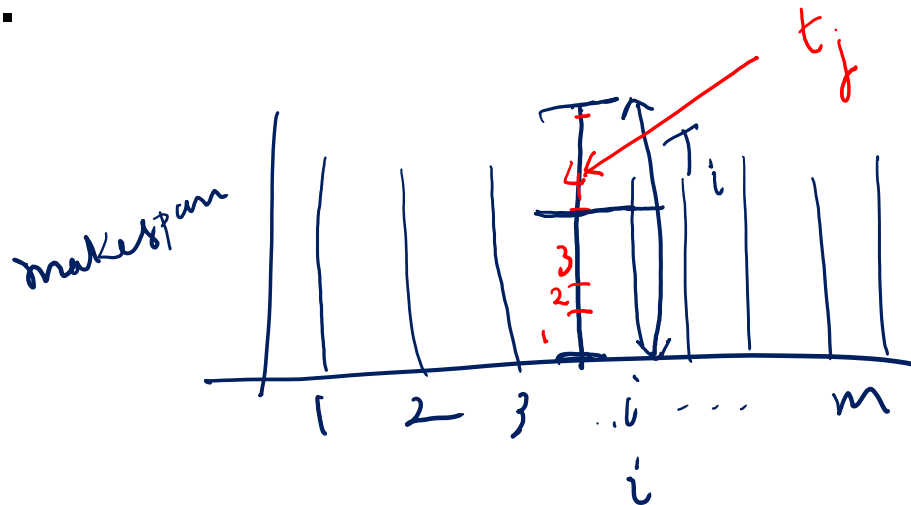
- The longest job will be on some machine which will work for at least that much time.

ex: $T^* \geq 100$

≈ 34.6

Load Balancing

- Let us now turn our attention to the greedy algorithm.
- Consider the machine that has the largest makespan according to the assignment produced by the greedy algorithm.
- Let this be the machine M_i , and its makespan be T_i .
- Let t_j be the last job assigned to M_i .
- Why did we assign t_j to M_i ?
 - As M_i has the least load just before assigning t_j to M_i .



$$A(I) = 6$$
$$OPT(I) = 5$$

$$\frac{|A(I)|}{|\text{OPT}(I)|} \leftarrow ub$$

- $\sum_{j=1}^n t_j \geq m(T_i - t_j)$, or

- sum of job durations \rightarrow

makepan

The diagram shows a vertical axis with indices 1, 2, 3, 4. A horizontal axis is labeled m_i . A vertical line segment is labeled T_i . A horizontal line segment is labeled t_j . The distance between the horizontal line and the horizontal axis is labeled $T_i - t_j$.

Load Balancing

- Further, notice that $t_j \leq T^*$.
- Use Observation 2.

$$t_j \leq \max_k t_k \leq T^* \quad \uparrow \text{Obs(2)}$$

- Therefore, $T_i = T_i - t_j + t_j = (T_i - t_j) + (t_j) \leq T^* + T^* = 2T^*$.

- Hence, $T_i \leq 2T^*$.

$$\frac{|A(I)|}{|\text{OPT}(I)|} = \frac{T_i}{T^*} \leq \frac{2T^*}{T^*} \leq 2$$

Load Balancing

- What would be one way to improve the solution produced by the greedy algorithm?
- Sort the jobs in descending order according to their runtime and assign them just as earlier.
- Let us call this as SortedGreedyAssignment.
- We will now analyze this approach.

Load Balancing

- Convince yourself that also SortedGreedyAssignment does not produce the best possible makespan.

Load Balancing

- The proof is not much different from the earlier proof.
- We will obtain a better bound for t_j , the last job assigned to the machine M_i that eventually has the largest makespan, T_i .
- Consider the case where there are fewer than m jobs.
- Then, the best possible assignment is to give each job to a different machine.
 - Our algorithm also does the same.
 - So, we indeed produce the best possible makespan.



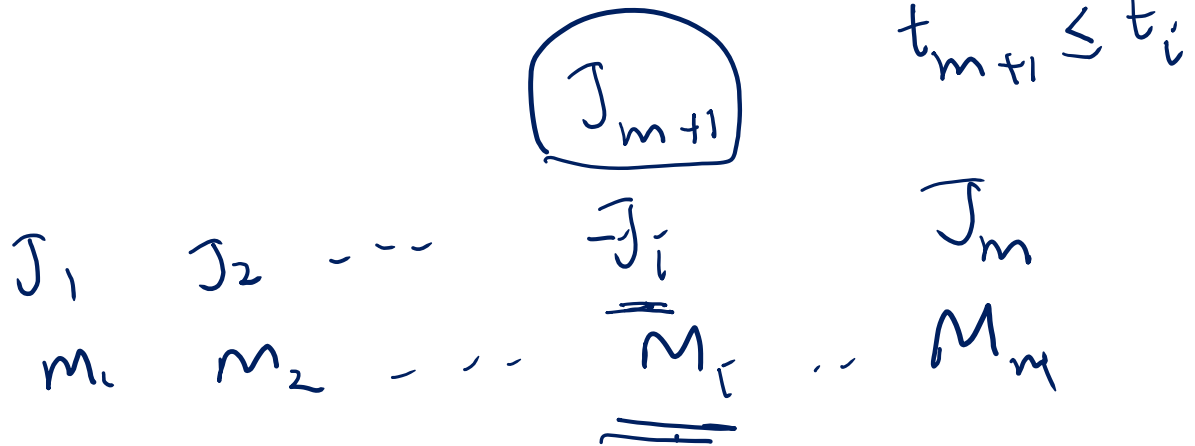
Load Balancing

- Let us consider the case that n is more than m .
- More jobs than machines.

- In any assignment, the job t_{m+1} in sorted order, is given to some machine that already has one additional job that is at least as long as t_{m+1} .

- So, $T^* \geq 2t_{m+1}$.

Best possible
makespan



Load Balancing

- Further, note that for machine M that has the largest makespan in our algorithm, if t_j is the last job assigned to M_i , then $t_j \leq t_{m+1}$.
- From the previous, $t_{m+1} \leq 1/2 T^*$.
- Now, $T_i = T_i - t_j + t_j \leq T^* + 1/2 T^* = 3/2 T^*$.

$$t_j \leq t_{m+1}$$

as $j > m+1$ &
duration in
sorted
order

by our
algorithm

observed

makespan

