

Approximation Algorithms

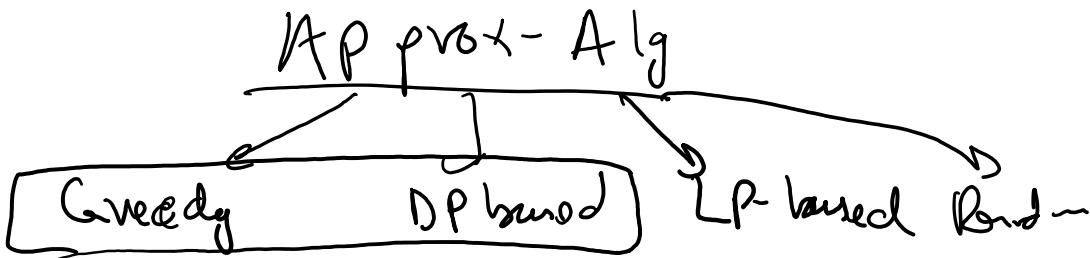
Tuesday, April 26, 2022 1:24 PM

decision problem $\leq_p \hookrightarrow$ optimization problem
(NP-hard)

A - Approximation Alg for Problem A

A Poly-time algorithm that outputs
a feasible solution ALG to A. Let
 OPT be the optimum solution to A.

$$\left(\frac{VAL(ALG)}{VAL(OPT)}, \frac{VAL(OPT)}{VAL(ALG)} \right) \leq \alpha$$



Greedy Approximation Algs:

Scheduling:
1) Set of n jobs $S[n]$
(efficiently divide jobs)
2) each job i has a processing time P_i .
3) m machines.

$$P_1, P_2, \dots, P_m, \dots, P_n$$

Goal: \rightarrow Find a partition of $[n]$ into $\{x_1, x_2, \dots, x_m\}$ that minimizes max-span.

$$\max_{i \in [n]} \sum_{j \in x_i} p_j \rightarrow \text{max-span}$$

total time for completion by machine 1.

$$\min_{x = \{x_1, x_2, \dots, x_m\}} \max_{i \in [n]} \sum_{j \in x_i} p_j$$

claim: \rightarrow SCHEDULING is NP-hard.

PARTITION \leq_p SCHEDULING.

I of partition
 $S = \{ \underbrace{\dots}_{\text{2 machines}} \}$

Appx-Alg:

1) for $i = 1$ to n

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 assign job i to machine with
 minimum load / processing time
 currently

- δ possible partition ✓
- Poly-time ✓

Claim: \rightarrow Algorithm is a 2-approximation algorithm.

let x_1, x_2, \dots, x_n be the partition
output by the alg

1) $x_1^*, x_2^*, \dots, x_n^*$ = optimum-pa

$$\max_{j \in [m]} \sum_{i \in X_j} p_i \leq 2 \max_{j \in [m]} \underbrace{\sum_{i \in X_j^*} p_i}_{2 \text{ max-span}}$$

intuitive - lower-bound

$$1) \text{max-span} \geq \frac{1}{m} \cdot \sum_{i \in [n]} p_i$$

$$2) \text{max-span} \geq \max_{i \in [n]} p_i$$

$\rightarrow \text{max } \sum_i r_i$ // $\max_{i \in [n]} r_i$

At any point in time, if the alg. assigns job a to machine b

$$X_b^{t+1} = X_b^t \cup \{a\}$$

$$\sum_{i \in X_b^t} p_i \leq \frac{1}{m} \sum_{i \in [n]} p_i \leq \frac{\max_{i \in [n]} p_i}{m}$$

$$\Rightarrow p_a \leq \max_{i \in [n]} p_i \leq \frac{\max_{i \in [n]} p_i}{m}$$

$$\sum_{i \in X_b^m} p_i \leq 2 \cdot \max_{i \in [n]} p_i \leq 2 \cdot \max_{i \in [n]} p_i \cdot m$$

3) - Appi-alg

- 1). Sort the jobs in decreasing order of processing.
- 2) Run the prev-alg.

claim: \Rightarrow but $\{x_1, x_2, \dots, x_n\}$ = partition

claim: \Rightarrow but $\langle x_1, x_2, \dots, x_n \rangle$ = partition returned by the alg.

$$\max_{j \in [m]} \sum_{i: x_i \in j} p_i \leq \frac{3}{2} \text{-max-span.}$$

OBS: $\Rightarrow n < m$, then x_1, x_2, \dots, x_n is optimum.

CASE: $\Rightarrow n > m$

$$\begin{aligned} \text{max-span} &\geq 2 \frac{p_{m+1}}{m} \\ &\geq 2 p_j \quad \begin{matrix} \hookrightarrow \\ j \geq m+1 \end{matrix} \end{aligned}$$

Alg assigns job a to machine b .

$$+ \sum_{i \in x_b} p_i \leq \frac{1}{m} \cdot \sum_{i \in [n]} p_i \leq \text{max-span}$$

$$p_a \leq \frac{1}{m} \text{max-span}$$

$$\sum_{i: x_i \in b} p_i \leq \frac{3}{2} \text{-max-span}$$

\leftarrow 'u' $\in \Sigma$ -max-star
 $i \in X^M$

SET-COVER

Given: $\Rightarrow D S_1, S_2, \dots, S_m \subseteq U = \{1, \dots, n\}$

2) Each set S_i has a weight
or cost w_i .

Goal: \Rightarrow Find $C = \{S_1, S_2, \dots, S_r\}$

s.t. $\bigcup_{i \in C} S_i = U$ and

$\sum_{S_i \in C} w_i$ is minimum.

Thm: \Rightarrow SET-COVER is NP-hard.

$\text{VERTEX-COVER} \leq_p \text{SET-COVER}$

Appx-Alg

while all elements in U are uncovered
choose a set S_i with smallest

choose a set S_i with smallest value of $\frac{w_i}{|S_i \cap R|}$ value

$R = \text{set of uncovered elements in } U$

Alg runs in poly time.

$$\underline{H(d)} = \frac{1}{d} + \frac{1}{d+1} + \dots + 1$$

harmonic
number

$$H(d) \in O(\lg d)$$

let C be the cover returned by the algorithm &

$OPT = \{S_1, S_2, \dots, S_r\}$ be the optimum-set-cover

$$VAL(ALG) = \sum_{S_i \in C} w_i = \sum_{e \in [n]} \underline{e}$$

A D algorithm will run in poly time

If element $e \in U$ was covered by set S_i by the ALG

$$c_e = \frac{w_i}{|S_i \cap R|}$$

\downarrow Set of elements at that time

$$OPT = \{S_1, S_2, \dots, S_V\}$$

$$val(OPT) = \sum_{S_i \in OPT} w_i$$

$w_i \cdot H(e)$

$$VAL(ALG) = \sum_{e \in U} c_e \leq \sum_{S_i \in OPT} \sum_{e \in S_i} c_e$$

$H/e \cdot val(OPT)$

Claim: $\sum_{e \in S_i} c_e \leq H(|S_i|) \cdot \frac{w_i}{|S_i|}$

$$S_i = \{S_1, S_2, \dots, S_d\}$$

order in which elements in S_i are covered by alg.

we cover a set S_i

$$\frac{\frac{w_i}{|S_i \cap R|}}{\tau} \leq \frac{w_i}{d-j+1}$$

All g covered S_j by set S_k

$$c_{S_j} = \frac{w_k}{|S_k \cap R|} \leq \frac{w_i}{|S_i \cap R|} \leq \frac{w_i}{d-j+1}$$

$$\sum_{j \in [d]} c_{S_j} \leq w_i \sum_{j \in [d]} \frac{1}{d-j+1}$$

\downarrow

$$= w_i \left(\frac{1}{d} + \frac{1}{d-1} + \dots + \frac{1}{1} \right)$$
$$= w_i H(d)$$

$$\boxed{\sum_{e \in S_i} c_e = H(|S_i|) \cdot w_i}$$

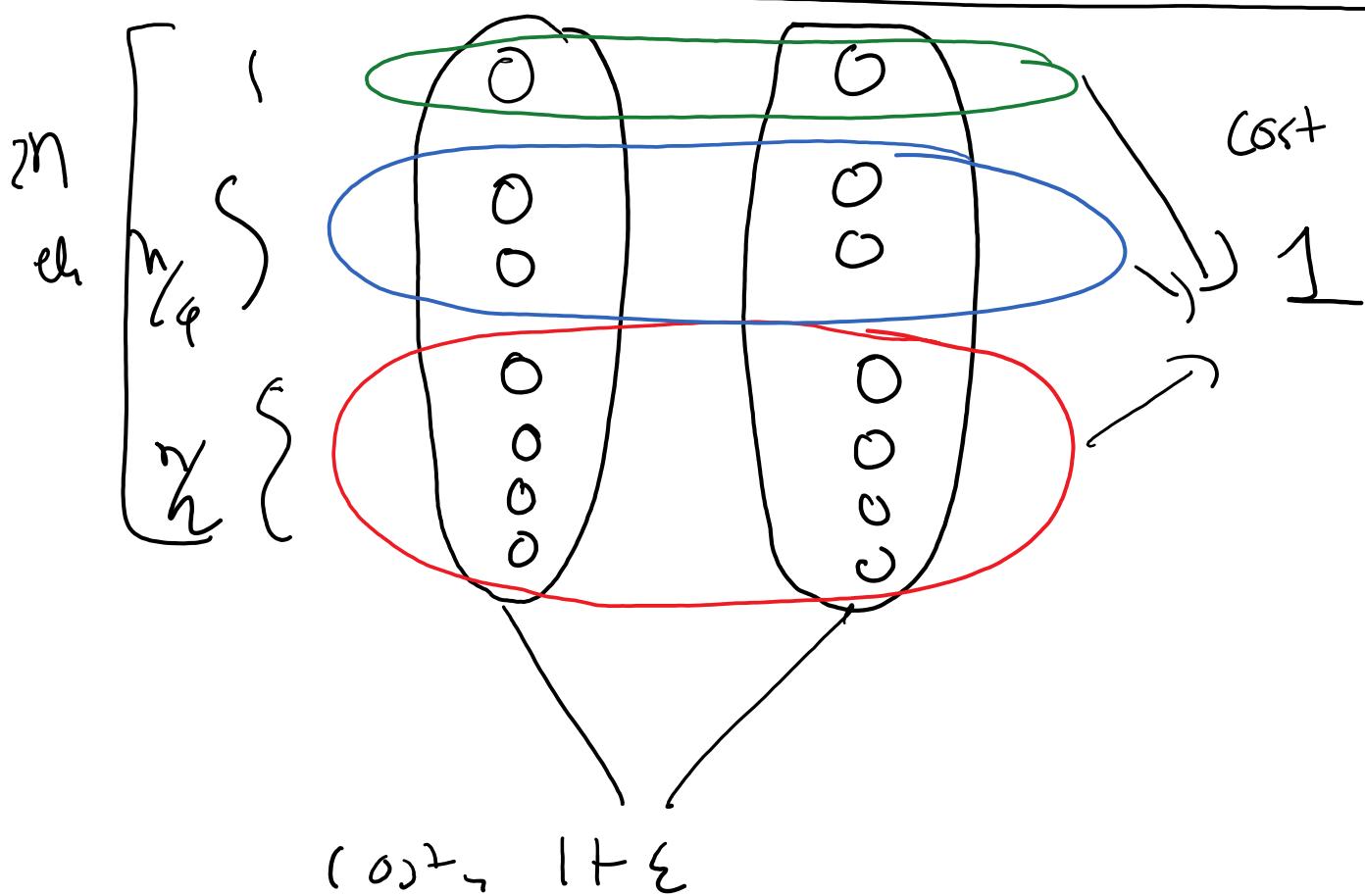
$$\lambda = \max_{i \in [m]} |S_i|$$

S. r.

$$\sum_{e \in S_i} c_e \leq h(l) \cdot w_i$$

$$\Rightarrow \sum_{e \in S} c_e \leq \sum_{e \in [n]} c_e$$

Thm: \Rightarrow Algorithm is a $H(l)$ -approximate-algorithm



LOWER BOUND of $\lg(n)$

DYNAMIC PROGRAMMING BY RD APPROX

KNAPSACK

Given: \rightarrow D set of n elements

- 1) each element i has value v_i & weight w_i
- 2) knapsack bag of size W .

Goal: Find $S \subseteq [n]$ s.t.

$$\sum_{s \in S} w_s \leq W \text{ and}$$

$$\sum_{s \in S} v_s \text{ is maximum}$$

PARTITION \leq_p KNAPSACK

Recall: \rightarrow Solve knapsack exactly
In time $O(n^2 V)$ \rightarrow PP-alg

In time $O(n^2 V) \rightarrow \underline{\text{PP-alg}}$

value $V = \max_{1 \leq i \leq n} v_i$

Thm.

For any $\epsilon > 0$, in time $O(n^{3\epsilon})$, we can find $S \subseteq [n]$ that fits in the knapsack f

$$\sum_{S \in S} v_S \geq (1 + \epsilon) \sum_{S \in S^*} v_S$$

value $S^* = \underline{\text{Optimum Subt.}}$

Alg

-- w_i about v . $\mathcal{O}(n^3 \varepsilon)$

$$1 \quad V''_{xi} = \lceil \frac{v_i}{b} \rceil \in \underline{\mathcal{O}(n/\varepsilon)}$$

4) Solve knapsack on I'' & f
return solution.

S is the set returned by AIG

S^* is the set -- -- -- by OPT.

$$\begin{aligned} \sum_{i \in S} v_i &\leq \sum_{i \in S^*} \lceil \frac{v_i}{b} \rceil \cdot b \\ &\leq \sum_{i \in S} \lceil \frac{v_i}{b} \rceil b \\ &\leq \sum_{i \in S} v_i + \frac{nb}{2} \end{aligned}$$

Recall $b = \frac{\varepsilon}{2n} \cdot V \Rightarrow nb = \frac{\varepsilon V}{2}$

- $S, S^* \perp \varepsilon \text{ (D)}$

$$= \sum_{i \in S} v_i + \varepsilon_2 \sum_{i \notin S} v_i$$

$$\sum_{i \in S} v_i \leq \sum_{i \in S} v_i + \varepsilon_2 \sum_{i \notin S} v_i$$

$$\Rightarrow (1 - \varepsilon_2) \cdot \sum_{i \in S} v_i \leq \sum_{i \in S} v_i$$

$$\Rightarrow \sum_{i \in S} v_i \leq (1 - \varepsilon_2)^{-1} \cdot \sum_{i \in S} v_i$$

$$\leq (1 + \varepsilon) \sum_{i \in S} v_i$$

for all
 $\varepsilon \leq 1$

Running time: $\underline{\underline{O(n^2 \sqrt{ })}}$