

(6) LP Techniques

1) $\text{prec}(\sum w_j C_j)$ and $1|\text{prec}, \{r_j, p_m\}|\sum w_j C_j$

$1|\text{prec}|\sum w_j C_j$ is NP-hard (and hence so is $1|\text{prec}, \{r_j, p_m\}|\sum w_j C_j$). In fact, even the non-weighted problem $1|\text{prec}|C_j$ is NP-hard.

Our goal is to write an LP relaxation for $1|\text{prec}|\sum w_j C_j$, with C_j variables denoting the completion times of jobs, and work our way towards an approximation algorithm.

Notice that the basic set of constraints need not make a valid schedule.

i.e.

$$\min \sum w_j C_j$$

s.t.

$$C_k \geq C_j + p_k \quad \text{if } j \rightarrow k$$

$$C_j \geq p_j$$

(J = set of all jobs,

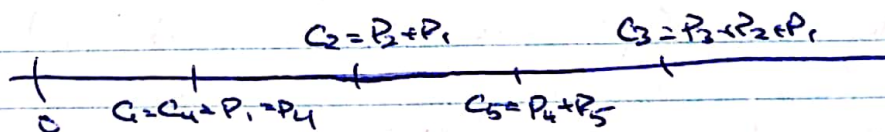
$$|J| = n)$$

if $j \rightarrow k$

$$\forall j \in J$$

But the following is a feasible solution:

$$1 \rightarrow 2 \rightarrow 3, \quad 4 \rightarrow 5$$



which is clearly not a valid schedule.

Continued

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Consider:

(LP)

$$\min \sum_{i \in J} w_i c_i$$

s.t.

$$C_k \geq C_j + P_k \quad \text{if } j \rightarrow k \quad (1)$$

$$C_j \geq P_j \quad \forall j \in J \quad (2)$$

$$\sum_{i \in A} P_i C_i \geq \frac{1}{2} P(A)^2 + \frac{1}{2} P^2(A) \quad \forall A \subseteq J \quad (3)$$

where:

$$P(A) = \sum_{i \in A} P_i \quad \text{and} \quad P^2(A) = \sum_{i \in A} P_i^2$$

Lemma 6.1:

Let OPT be the optimal value of $\sum_{i \in J} w_i c_i$ and OPT_{LP} be the optimal value of (LP). Then,

$$OPT_{LP} \leq OPT.$$

[Proof]

We want to show that any feasible schedule yields a feasible solution to (LP), where $C_i = C_i^s \quad \forall i$

Clearly the C_i^s will satisfy (1) and (2).

Now, consider any $A \subseteq J$.

→ continued.

{Proof} (cont.)

We will say $k \leq j$ if k is a schedule before j in S .

By def'n:

$$C_j^S = C_j = \sum_{k \in A: k \leq j} p_k \geq \sum_{k \in A: k \leq j} p_k$$

Then,

$$\sum_{j \in A} p_j C_j \geq \sum_{j \in A} \sum_{\substack{k \in A \\ k \leq j}} p_j p_k$$

$$= \sum_{j \in A} p_j^2 + \sum_{\substack{j, k \in A \\ k < j}} p_j p_k$$

$$= \frac{1}{2} \sum_{j \in A} p_j^2 + \frac{1}{2} \left(\sum_{j, k \in A} p_j p_k \right)^2$$

$$= \frac{1}{2} P^2(A) + \frac{1}{2} (P(A))^2$$

□

Using the LP, we can consider the following algorithm:

Algorithm A:

(1) Solve (LP) to obtain solution $\{C_j^*\}$

(Note: These C_j^* 's need not correspond to job completion times in a feasible schedule)

(2) Schedule jobs in increasing order of C_j^*

(Note: Schedule returned will be feasible, by design of the (LP): If $j \rightarrow k$, then $C_j^* \leq C_k^*$)

→ Theorem 6.1

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Theorem 6.1:

A_1 is a 2-approx. alg. for $\text{IPred} \sum w_j C_j$
 [Proof]

Let jobs be ordered $1, \dots, n$ so that $C_1^* \leq \dots \leq C_n^*$ and let S be the schedule outputted by A_1 .

We want to show: $C_j^S \leq 2C_j^* \quad \forall j=1, \dots, n$.
 So that:

$$\sum w_j C_j^S \leq 2 \sum w_j C_j^* \\ \Rightarrow \text{OPT}_S \leq 2 \cdot \text{OPT}_{\text{IPred} \sum w_j C_j}$$

Fix job j . Let $A = \{1, \dots, j\}$ and consider
 (C) for job-set A :

$$(i) \sum_{k \in A} P_k C_k^* \geq \frac{1}{2} P(A)^2 + \frac{1}{2} P^2(A) \geq \frac{1}{2} P(A)^2$$

$$(ii) C_j^* \cdot \sum_{k \in A} P_k \geq \sum_{k \in A} P_k C_k^* \quad (\text{Since } C_k^* \leq C_j^*)$$

$$\Rightarrow C_j^* \cdot \sum_{k \in A} P_k \geq \frac{1}{2} P(A)^2$$

$$\Rightarrow C_j^* \geq \frac{1}{2} P(A)^2 = \frac{1}{2} \sum_{k \in j} P_k = \frac{1}{2} C_j^S$$

□.

Q: Is A efficient? In particular, is solving
 (LP) efficient?

- Notice that if $|J|=n$, then there are 2^n
 subsets $A \subseteq J$ and hence 2^n constraints.
 However, in fact:

Theorem 6.2:

(LP) can be solved in polytime
 [Proof] See Theorem 6.8

$I(\text{prec}, r_j | \sum w_j C_j)$ and $I(\text{prec}, r_j, \text{prefn} | \sum w_j C_j)$

Claim 6.3:

We may assume wlog that if $j \rightarrow k$, then $r_k \geq r_j + p_j$

[Proof]

The earliest time k can start is $r_j + p_j$, so if $r_k < r_j + p_j$, we can always advance the release date of k to $r_k \leftarrow \max(r_k, r_j + p_j)$ without affecting the feasibility of the schedule. \square

As with $I(\text{prec} | \sum w_j C_j)$, we will write an LP-relaxation for our problem.

(LP)

$$\min \sum w_j C_j$$

s.t.

$$C_k \geq C_j + p_k \quad \forall j \rightarrow k \quad (1)$$

$$C_j \geq r_j + p_j \quad \forall j \quad (2)$$

$$\sum_{j \in A} p_j C_j \geq p(A) r(A) + \frac{1}{2} p(A)^2 + \frac{1}{2} p^2(A) \quad \forall A \subseteq J \quad (3)$$

where:

$$r(A) := \min_{j \in A} r_j \quad \text{for all } A \subseteq J.$$

Note:

It is easy to see:

$$\begin{aligned} \text{OPT}_{LP} &\leq \text{OPT}_{I(\text{prec}, r_j, \text{prefn} | \sum w_j C_j)} \\ &\leq \text{OPT}_{I(\text{prec}, r_j | \sum w_j C_j)} \end{aligned}$$

\hookrightarrow Algo.

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Algorithm A_2 (for $I(\text{prec}, r_j, \text{pntn}(\sum w_j C_j))$)

- 1) Solve (LP') to obtain $\{C_j^*\}$
- 2) Schedule preemptively in increasing C_j^* order.

i.e. At each point of time, schedule the available job with smallest C_j^* value, preempting as necessary)

Theorem 6.4:

A_2 produces a feasible schedule

[Proof]

We want to ensure that precedence constraints are respected.

Suppose $j \rightarrow k$. By Claim 3, if k is available at time t , either:

- j is also available at time t , or
- j is completed by time t

And also, $C_j^* \leq C_k^*$ by our ordering, unless j has completed, we will not process k . \square

Theorem 6.5:

A_2 is a 2-approx. for $I(\text{prec}, r_j, \text{pntn}(\sum w_j C_j))$.

[Proof]

Let S be our schedule outputted by A_2 , and order jobs so that:

$$C_1^S \leq C_2^S \leq \dots \leq C_n^S$$

We want to show that: $C_j^S \leq 2 \cdot C_j^* \quad \forall j$

\hookrightarrow can't

[Proof] (cont)

Consider job j and let t be the earliest time before C_j^s s.t.:

- 1) There is no idle time in $[t, C_j^s]$, and
- 2) m/c is only processing jobs k s.t. $k \leq j$ in $[t, C_j^s]$

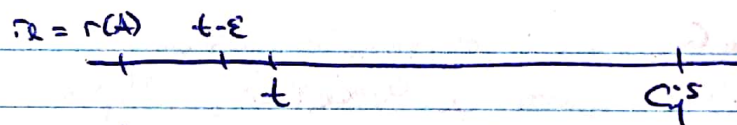
Let $A = \{ \text{jobs } k: k \text{ scheduled in } [t, C_j^s] \}$

By defn: $\forall k \in A, C_k^* \leq C_j^*$

We claim: $t = r(A)$.

Clearly $t \geq r(A)$, since some job $k \in A$ is scheduled at time t .

Suppose $t > r(A)$, and suppose let s.t. $r = r(A)$.



At time $t - \epsilon$, j is available since $r \leq t - \epsilon$ and j has not completed since j is scheduled in $[t, C_j^s]$. So, at time $t - \epsilon$, the m/c cannot be idle, and either:

- has scheduled j or some $j' \leq j$. but then we can move j to an earlier time
- has scheduled some $j' > j$, contradicting our schedule rule.

→ cont

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[Proof] (cont.)

So, $t = rCH$. Applying (3) in (LP'), we get:

$$\begin{aligned} \bullet \sum_{k \in A} p_k C_k^* &\geq pCH rCH + \frac{1}{2} pCH^2 + \frac{1}{2} p^2 CH \\ &\geq pCH rCH + \frac{1}{2} pCH^2 \end{aligned}$$

$$\bullet C_j^* pCH \geq \sum_{k \in A} p_k C_k^*$$

(Since $C_k^* \leq C_j^*$)

$$\Rightarrow C_j^* pCH \geq \frac{pCH rCH}{\cancel{pCH}} + \frac{1}{2} pCH^2$$

$$\Rightarrow C_j^* \geq rCH + \frac{1}{2} pCH$$

So,

$$C_j^S \leq t + pCH = rCH + pCH \leq 2C_j^* \quad \square$$

Proposition 6.6:

$A_2 = \text{CONVERT}$ gives a 4-approx $|I_{\text{prec}, r_j}| \sum w_j c_j$

Algorithm A_3 :

1) Solve (LP') to obtain $\{C_j^*\}$

2) Non-preemptively schedule in increasing C_j^* order

Theorem 6.7:

A_3 is a 3-approx. algo. for $|I_{\text{prec}, r_j}| \sum w_j c_j$.

→ Proof.

[Proof]

Let S be the schedule produced by A_3 and order jobs: $C_1^* \leq \dots \leq C_n^*$ (So, $C_1^S \leq \dots \leq C_n^S$)

Let $A = \{1, \dots, i\}$, then by (3) from (LP'), we get:

$$P(A) C_i^* \geq \sum_{k \in A} P_k C_k^* \geq \frac{1}{2} P(A)^2$$

$$\Rightarrow P(A) \leq 2 C_i^*$$

So,

$$C_i^S \leq \max_{k \leq i} r_k + \sum_{k \leq i} P_k$$

$$\leq C_i^* + 2 C_i^* \quad (\text{since } r_k \leq C_k^* \leq C_i^*)$$
$$= 3 C_i^* \quad \forall i$$

Which gives:

$$\sum w_i C_i^S \leq 3 \cdot \sum w_i C_i^*$$

$$= 3 \text{OPT}_{LP'}$$

$$\leq 3 \cdot \text{OPT}_{LP} \sum w_i$$

□

Ellipsoid Method:

First polynomial algorithm to solve LPs

Def'n 6.1: (Separation Oracle)

For an LP with feasible region:

$$K := \{x \in \mathbb{R}^n : a_i^T x \leq b_i \quad \forall i=1, \dots, m\}$$

Given input $y \in \mathbb{R}^n$, the separation oracle is a procedure that correctly determines if $y \in K$, or finds a constraint $a_i^T x \leq b_i$ of K that is violated by y .

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Theorem 6.8 (Ellipsoid Method)

Consider an LP:

(P)

$$\min c^T x$$

s.t.

$$a_i^T x \leq b_i \quad \forall i=1, \dots, m$$

$$x \in \mathbb{R}^n$$

Let $S := \max(\text{size of } c, \max_{i=1, \dots, m} (\text{size of } (a_i, b_i)))$

Given a separation oracle A for (P), we can solve (P) using $\text{poly}(n, S)$ calls to A and $\text{poly}(n, S)$ of other operations.

Corollary 6.8:

Given polytime separation oracle, for $\{x \in \mathbb{R}^n : a_i^T x \leq b_i, i=1, \dots, m\}$, we can solve (P) in $\text{poly}(n, S)$ time.