# Single machine models: Maximum Lateness <u>Problem 1|| $L_{max}$ </u>:

• Earliest due date first (EDD) is optimal for  $1||L_{max}$ (Jackson's EDD rule) -1-

• Proof: special case of Lawler's algorithm

Problem  $1|r_j|C_{max}$ :

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Problem  $1|r_j|C_{max}$ :

•  $1|r_j|C_{max} \propto 1||L_{max}|$ 

-define  $d_j := K - r_j$ , with constant  $K > \max r_j$ 

– reversing the optimal schedule of this  $1||L_{max}$ -problem gives an optimal schedule for the  $1|r_j|C_{max}$ -problem

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# Single machine models: Maximum Lateness <u>Problem 1 $prec | L_{max}$ </u>:

• if  $d_j < d_k$  whenever  $j \rightarrow k$ , any EDD schedule respects the precedence constraints, i.e. in this case EDD is optimal

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 $\bullet$  defining  $d_j:=\min\{d_j,d_k-p_k\}$  if  $j\to k$  does not increase  $L_{max}$  in any feasible schedule

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## Algorithm $1|prec|L_{max}$

1. make due dates consistent: set  $d_j = \min\{d_j, \min_{k|j \to k} d_k - p_k\}$ 2. apply EDD rule with modified due dates

Remarks on Algorithm  $1|prec|L_{max}$ 

- $\bullet$  leads to an optimal solution
- Step 1 can be realized in  $O(n^2)$
- problem  $1|prec|L_{max}$  can be solved without knowledge of the processing times, whereas Lawler's Algorithm (which also solves this problem) in general needs this knowledge (Exercise),

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• Problem  $1|r_j, prec|C_{max} \propto 1|prec|L_{max}$ 

# Single machine models: Maximum Lateness <u>Problem $1|r_j|L_{max}$ </u>:

- problem  $1|r_j|L_{max}$  is NP-hard
- Proof: by reduction from 3-PARTITION (on the board)

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Lecture 3

Problem  $1|pmtn, r_j|L_{max}$ :

• preemptive EDD-rule: at each point in time, schedule an available job (job, which release date has passed) with earliest due date.

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• preemptive EDD-rule leads to at most k preemptions (k = number of distinct release dates)

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Problem  $1|pmtn, r_j|L_{max}$ :

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- preemptive EDD-rule leads to at most k preemptions (k = number of distinct release dates)
- preemptive EDD solves problem  $1|pmtn, r_j|L_{max}$
- Proof (on the board) uses following results:

$$\begin{aligned} -L_{max} &\geq r(S) + p(S) - d(S) \text{ for any } S \subset \{1, \dots, n\}, \text{ where} \\ r(S) &= \min_{j \in S} r_j, \ p(S) = \sum_{j \in S} p_j, \ d(S) = \max_{j \in S} d_j \\ -\text{ preemptive EDD leads to a schedule with} \\ L_{max} &= \max_{S \subset \{1, \dots, n\}} r(S) + p(S) - d(S) \end{aligned}$$

Remarks on preemptive EDD-rule for  $1|pmtn, r_j|L_{max}$ :

- can be implemented in  $O(n \log(n))$
- is an 'on-line' algorithm
- $\bullet$  after modification of release and due-dates, preemptive EDD solves also  $1|prec,pmtn,r_{j}|L_{max}$

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Approximation algorithms for problem  $1|r_j|L_{max}$ :

• a polynomial algorithm A is called an  $\alpha$ -approximation for problem P if for every instance I of P algorithm A yields an objective value  $f_A(I)$  which is bounded by a factor  $\alpha$  of the optimal value  $f^*(I)$ ; i.e.  $f_A(I) \leq \alpha f^*(I)$ 

-7-

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- for the objective  $L_{max}$ ,  $\alpha$ -approximation does not make sense since  $L_{max}$  may get negative
- for the objective  $T_{max}$ , an  $\alpha$ -approximation with a constant  $\alpha$  implies  $\mathcal{P} = \mathcal{NP}$  (if  $T_{max} = 0$  an  $\alpha$ -approximation is optimal)

- Single machine models: Maximum Lateness
- The head-body-tail problem  $(1|r_j, d_j < 0|L_{max})$ 
  - n jobs
  - $\bullet$ jobj:release date  $r_{j}$  (head), processing time  $p_{j}$  (body), delivery time  $q_{j}$  (tail)

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- starting time  $S_j \ge r_j$ ;
- completion time  $C_j = S_j + p_j$
- delivered at  $C_j + q_j$
- goal: minimize  $\max_{j=1}^{n} C_j + q_j$

The head-body-tail problem  $(1|r_j, d_j < 0|L_{max})$ , (cont.)

- define  $d_j = -q_j$ , i.e. the due dates get negative!
- result:  $\min_{j=1}^{n} C_j + q_j = \min_{j=1}^{n} C_j d_j = \min_{j=1}^{n} L_j = L_{max}$
- head-body-tail problem equivalent with  $1|r_j|L_{max}$ -problem with negative due dates

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<u>Notation</u>:  $1|r_j, d_j < 0|L_{max}$ 

- an instance of the head-body-tail problem defined by n triples  $(r_j, p_j, q_j)$ is equivalent to an inverse instance defined by n triples  $(q_j, p_j, r_j)$
- for the head-body-tail problem considering approximation algorithms makes sense

Single machine models: Maximum Lateness <u>The head-body-tail problem  $(1|r_j, d_j < 0|L_{max})$ , (cont.)</u> •  $L_{max} \ge r(S) + p(S) + q(S)$  for any  $S \subset \{1, \ldots, n\}$ , where  $r(S) = \min_{j \in S} r_j, \ p(S) = \sum_{j \in S} p_j, \ q(S) = \min_{j \in S} q_j$ (follows from  $L_{max} \ge r(S) + p(S) - d(S)$  - slide 5)

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## Single machine models: Maximum Lateness Approximation ratio for EDD for problem $1|r_j, d_j < 0|L_{max}$

 $\bullet$  structure of an schedule

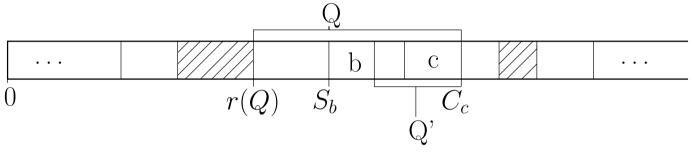
$$\begin{array}{c|c} Q \\ \hline \\ \hline \\ 0 \\ \end{array} \\ t = r(Q) \\ \hline \\ C_c \\ \end{array} \\ \hline \\ C_c \\ \end{array}$$

- critical job c of a schedule: job with  $L_c = \max L_j$
- critical sequence Q: jobs processed in the interval  $[t, C_c]$ , where t is the earliest time that the machine is not idle in  $[t, C_c]$

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- if  $q_c = \min_{j \in Q} q_j$  the schedule is optimal since then  $L_{max}(S) = L_c = C_c - d_c = r(Q) + p(Q) + q(Q) \le L_{max}^*$
- Notation:  $L_{max}^*$  denotes the optimal value

- Approximation ratio for EDD for problem  $1|r_j, d_j < 0|L_{max}$ 
  - structure of an schedule



- $\bullet$  interference job b: last scheduled job from Q with  $q_b < q_c$
- Lemma: For the objective value  $L_{max}(EDD)$  of an EDD schedule we have

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1. 
$$L_{max}(EDD) - L_{max}^* < q_c$$
  
2.  $L_{max}(EDD) - L_{max}^* < p_b$ 

• <u>Theorem</u>: EDD is 2-approximation algorithm for  $1|r_j, d_j < 0|L_{max}$ 

- Approximation ratio for EDD for problem  $1|r_j, d_j < 0|L_{max}$ 
  - <u>Remarks</u>:
    - EDD is also a 2-approximation for  $1|prec,r_j,d_j<0|L_{max}$  (uses modified release and due dates)
    - by an iteration technique the approximation factor can be reduced to 3/2

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Enumerative methods for problem  $1|r_j|L_{max}$ 

- we again will use head-body-tail notation
- <u>Simple branch and bound method</u>:
  - branch on level i of the search tree by selecting a job to be scheduled on position i

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- if in a node of the search tree on level i the set of already scheduled jobs is denoted by S and the finishing time of the jobs from S by t, for position i we only have to consider jobs k with

$$r_k < \min_{j \notin S} (\max\{t, r_j\} + p_j)$$

- -lower bound: solve for remaining jobs  $1|r_j, pmtn|L_{max}$
- $-\operatorname{search}$  strategy: depth first search + selecting next job via lower bound

Advanced b&b-methods for problem  $1|r_j|L_{max}$ 

- node of search tree = restricted instance
- restrictions = set of precedence constraints
- branching = adding precedence constraints between certain pairs of jobs

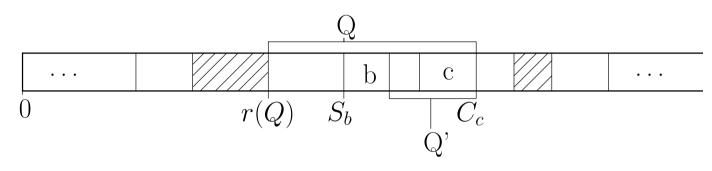
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- $\bullet$  after adding precedence constraints, modify release and due dates
- apply EDD to instance given in a node
  - critical sequence has no interference job: EDD solves instance optimal
    - $\rightarrow$  backtrack
  - $-\operatorname{critical}$  sequence has an interference job: branch

Advanced b&b-methods for problem  $1|r_i|L_{max}$  (cont.)

branching, given sequence Q, critical job c, interference job b, and set Q' of jobs from Q following b

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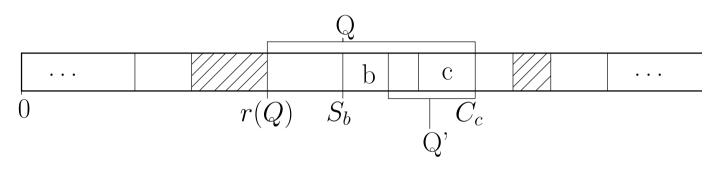


- $L_{max} = S_b + p_b + p(Q') + q(Q') < r(Q') + p_b + p(Q') + q(Q')$
- if b is scheduled between jobs of Q' the value is at least  $r(Q') + p_b + p(Q') + q(Q')$ ; i.e. worse than the current schedule

Advanced b&b-methods for problem  $1|r_j|L_{max}$  (cont.)

branching, given sequence Q, critical job c, interference job b, and set Q' of jobs from Q following b

-16-



- $L_{max} = S_b + p_b + p(Q') + q(Q') < r(Q') + p_b + p(Q') + q(Q')$
- if b is scheduled between jobs of Q' the value is at least  $r(Q') + p_b + p(Q') + q(Q')$ ; i.e. worse than the current schedule
- branch by adding either  $b \to Q'$  or  $Q' \to b$

Advanced b&b-methods for problem  $1|r_i|L_{max}$  (cont.)

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- lower bounds in a node: maximum of
  - lower bound of parent node
  - -r(Q') + p(Q') + q(Q') $-r(Q' \cup \{b\}) + p(Q' \cup \{b\}) + q(Q' \cup \{b\})$

using the modified release and due dates

- upper bound UB: best value of the EDD schedules
- discard a node if lower bound  $\geq UB$
- search strategy: select node with minimum lower bound

- Single machine models: Maximum Lateness Advanced b&b-methods for problem  $1|r_j|L_{max}$  (cont.)
  - speed up possibility:

 $-\operatorname{let} k \notin Q' \cup \{b\} \text{ with } r(Q') + p_k + p(Q') + q(Q') \ge UB$  $-\operatorname{if} r(Q') + p(Q') + p_k + q_k \ge UB \text{ then add } k \to Q'$  $-\operatorname{if} r_k + p_k + p(Q') + q(Q') \ge UB \text{ then add } Q' \to k$ 

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