

A compact MIP formulation for single machine scheduling to minimize a piecewise linear objective function

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Optimeo'08

The problem PWL (I)

Data

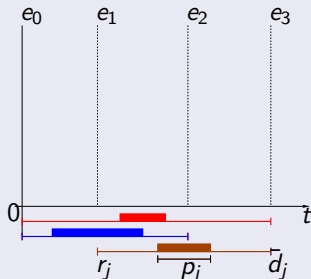
1 machine, m intervals

$I_u = [e_{u-1}, e_u]$, $u \in M$;

for each job $j \in N = \{1, \dots, n\}$

- r_j - release date,
- p_j - processing time,
- \bar{d}_j - deadline,

such that $\{r_j, \bar{d}_j\}_{j \in N} \subseteq \{e_u\}_{u \in M}$.



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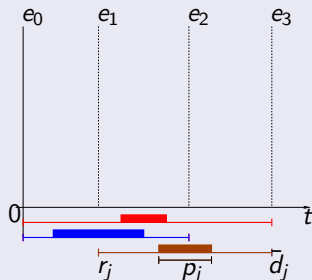
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Constraints

- the machine can process only one job at a time,
- preemption is not allowed,

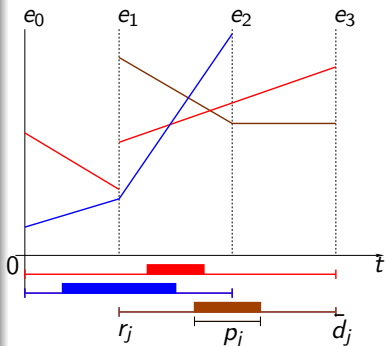
The problem PWL (II)

Objective

$\forall j \in N$, the cost function $F_j(C_j)$ is linear in each interval :

if $e_{u-1} < C_j \leq e_u$ then

$$F_j(C_j) = f_j^u + w_j^u \cdot (C_j - e_{u-1}).$$



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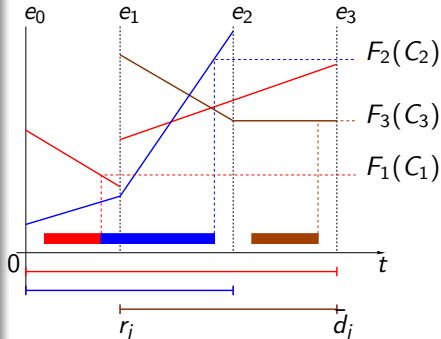
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We minimize the total cost :

$$\sum_{j \in N} F_j(C_j)$$



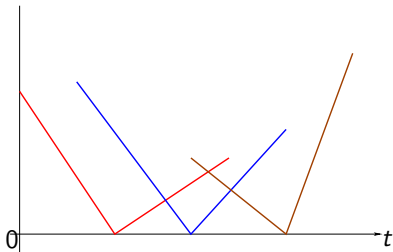
Generality of the problem PWL

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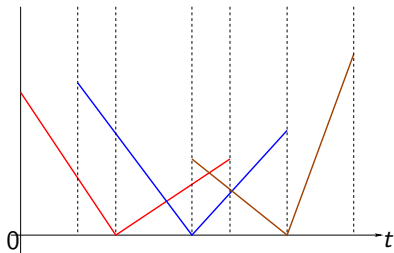
Earliness-tardiness scheduling



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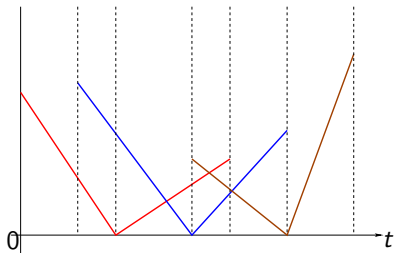
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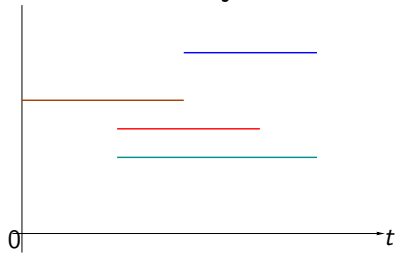
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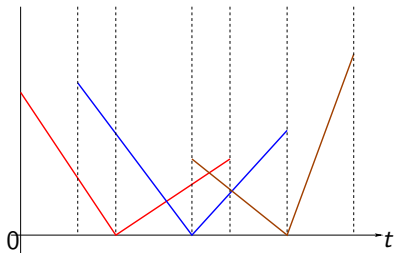
Maximizing weighted number of on-time jobs



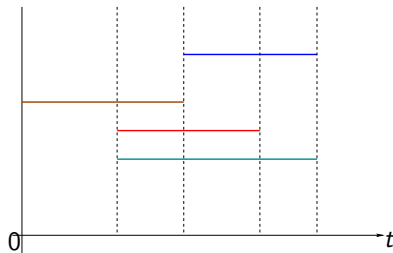
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Time-indexed formulation

Data is integer \rightarrow Time-indexed formulation [Sousa, Wolsey, 92]

$X_{jt} \in \{0, 1\}$, $X_{jt} = 1$ iff job j is started at time moment t

$$\min \sum_{t=0}^T F_j(t + p_j) X_{jt}$$

$$\text{s.t.} \quad \sum_{t=r_j}^{\bar{d}_j - p_j} X_{jt} = 1, \quad j \in N,$$

$$\sum_{j \in N} \sum_{s=t-p_j+1}^t X_{js} \leq 1, \quad t \in [0, T),$$

Time-indexed formulation

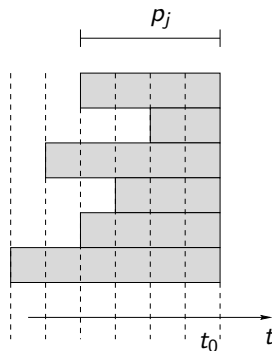
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Interval decomposition

The cost functions are **linear** in an interval.

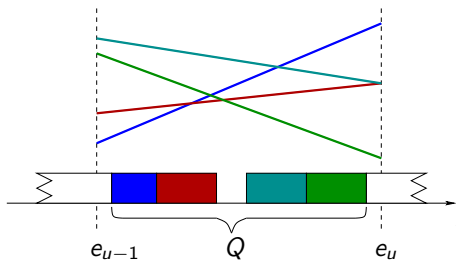
Interval decomposition

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Property. If all jobs in a set Q are started and completed in the same interval $[e_{u-1}, e_u]$, it is optimal to process them according to the Smith rule :

$$i \rightarrow j \text{ if } \frac{w_i^u}{p_i} > \frac{w_j^u}{p_j}.$$

Idle time between jobs with positive and negative weights w_j^u



Interval-indexed formulation

Variables

$X_j^u, Y_j^u \in \{0, 1\}$ — whether job j is started (completed) before e_u

W_u — length of the idle time in I_u

F_j^u — difference between the completion time and the border of I_u

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Variables constraints

$$\forall u \in M, \forall j \in N, \quad Y_j^{u-1} \leq Y_j^u, \quad X_j^{u-1} \leq X_j^u, \quad Y_j^u \leq X_j^u$$

$$\forall u \in M, \forall j \in NS_u, \quad X_j^{u-1} \leq Y_j^u$$

$$\forall u \in M, \forall j \in NB_u, \quad Y_j^u \leq X_j^{u-1}$$

$$\forall j \in N, r_j = e_u, \quad X_j^u = 0$$

$$\forall j \in N, \bar{d}_j = e_u, \quad Y_j^u = 1$$

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Knapsack constraints

$$\forall u \in M, \quad \sum_{j \in N} p_j Y_j^u + \sum_{v=1}^u W_v \leq e_u$$

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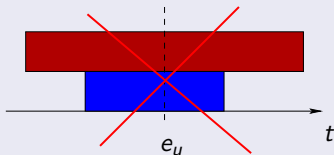
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Additional constraints I

$$\forall u \in M, \quad \sum_{i \in N} (X_i^u - Y_i^u) \leq 1$$



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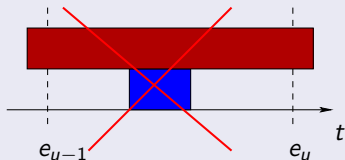
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NS_u, NB_u — set of “small” jobs and “big” jobs for I_u

Additional constraints II

$$\forall u \in M, \forall j \in NS_u, \quad \sum_{i \in NB_u} (X_i^{u-1} - Y_i^u) + Y_j^u - X_j^{u-1} \leq 1$$



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Theorem

Vector (X, Y, W) satisfies the constraints presented

\Leftrightarrow the corresponding schedule is feasible

Interval-indexed formulation

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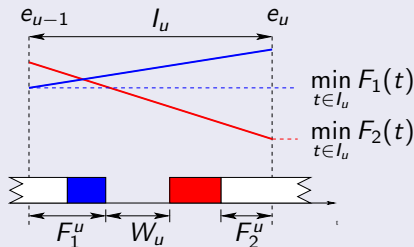
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$$\forall u \in M, \forall j \in N, \quad F_j^u \geq \alpha X + \beta Y + \gamma W$$

Tightening interval-indexed formulation

- Redundant constraints.

Example : $p_j \leq e_u - e_v : C_j \leq e_v \Rightarrow C_j \leq e_u$ ($Y_j^u \geq X_j^u$)

- Appropriate partition of the time horizon

Known order of jobs completed but **not necessarily started** in the same interval.

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Numerical experiments

Cplex 10.0, standard settings and only standard cuts,
1000 seconds time limit

- **Minimizing total weighted earliness and tardiness with release dates** ($1 \mid r_j \mid \alpha_j E_j + \beta_j T_j$) — some 15 jobs instances could not be solved (all are solved if number of intervals is small).
- **Minimizing total weighted earliness and tardiness** ($1 \parallel \alpha_j E_j + \beta_j T_j$) — all 15 jobs instances are solved (majority of 20 jobs instances are not solved).
- **Minimizing total weighted tardiness** ($1 \parallel w_j T_j$) — all 20 jobs instances are solved (some 30 jobs instances are not solved). Here we need only Y and F variables.

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Similar performance with the Time-indexed formulation

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