

# Exact algorithms for the bilevel scheduling of uniform parallel machines in the context of coupling maintenance and scheduling decisions

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## 1 Introduction

- Problem definition
- Block structure property
- Complexity by reduction

## 2 Exact methods

- MIP
- Column generation

## 3 Computational results

## 4 Conclusions

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# Bilevel scheduling: problem definition

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Global  
scheduler

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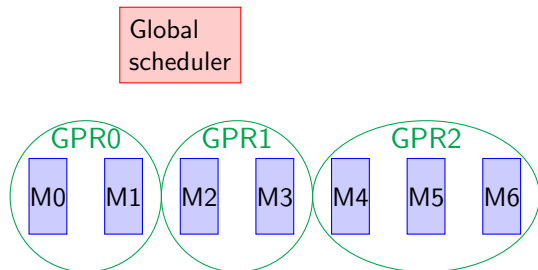
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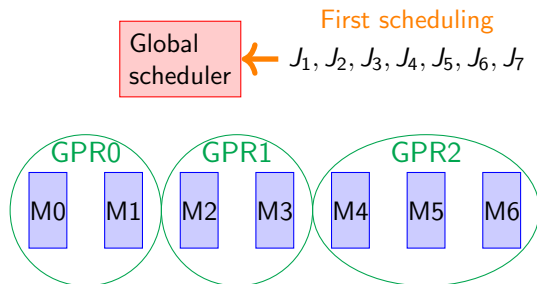
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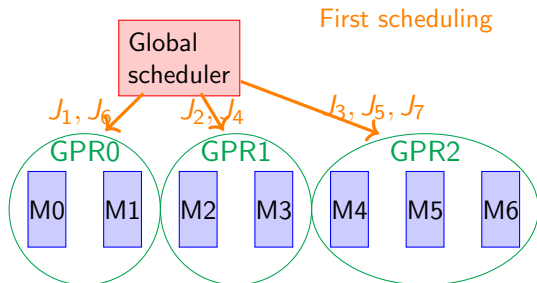
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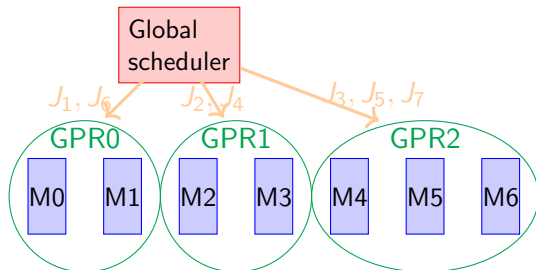
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After period  $T$  :



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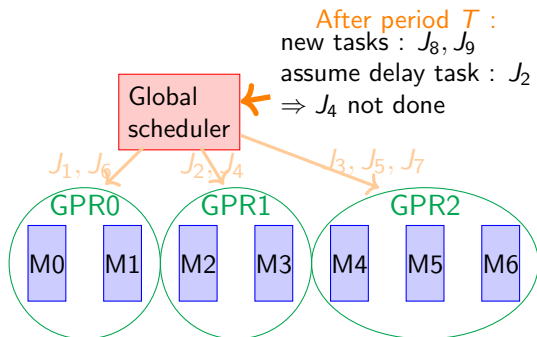
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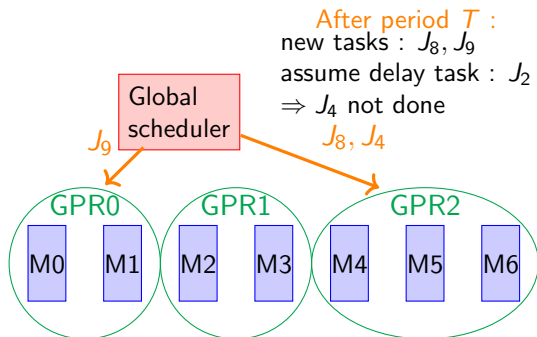
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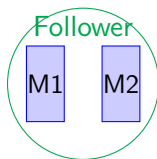
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- A follower composed of several machines



Leader



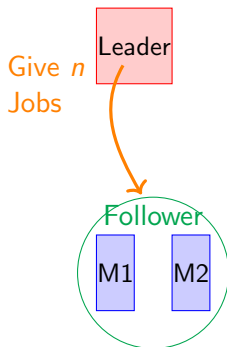
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We have 2 agents:

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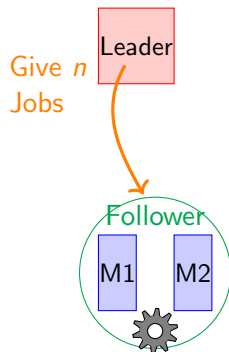
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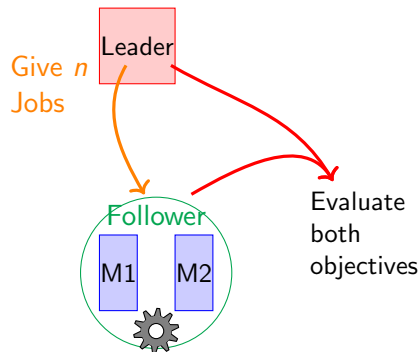
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The follower returns the schedule  $\rightarrow \min \sum_j w_j U_j^L$  **among all optimal schedules** for  $\sum_j C_j^F$ .

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$\Leftrightarrow \text{Lex} \left( \sum_j C_j^F, \sum_j w_j U_j^L \right)$  for the follower.



# Bilevel scheduling: problem definition

## Definition

We can define our problem as:

$$\begin{aligned} \min_{\substack{\mathcal{I} \subset \mathcal{J} \\ |\mathcal{I}|=n}} \sum_{j \in \sigma^*} w_j U_j^L \\ \text{st.} \quad \sigma^* = \operatorname{argmin}_{\sigma \in \mathfrak{S}_{\mathcal{I}}} \left( \operatorname{Lex} \left( \sum_{j \in \sigma} C_j^F, \sum_{j \in \sigma} w_j U_j^L \right) \right) \end{aligned} \quad (1)$$

where:

- $\mathfrak{S}_{\mathcal{I}}$  is the set of schedules formed of all jobs in  $\mathcal{I}$ .
- $C_j^F$  is the completion time of the job  $j$ .
- $U_j^L = \begin{cases} 1 & \text{if } C_j^F > d_j \\ 0 & \text{otherwise} \end{cases}$

## 1 Introduction

- Problem definition
- **Block structure property**
- Complexity by reduction

# Bilevel scheduling: block structure property

## Property

Theorem  $Q||\sum C_j$ <sup>1</sup>

$Q||\sum C_j$  is polynomial in  $\mathcal{O}(n \log(n))$

Assignment problem: **larger** processing times  $p_1 \geq p_2 \geq \dots \geq p_n$  are assigned to the **smaller** coefficients  $t_j = \frac{k}{V_i}$ .

Here, job  $j$  will be in the  $k$ -th position (counting from the end) on machine  $i$ .

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We focus on a special case with 2 speeds:  $Q | V_i \in \{V_0, V_1\} | \sum_j C_j$

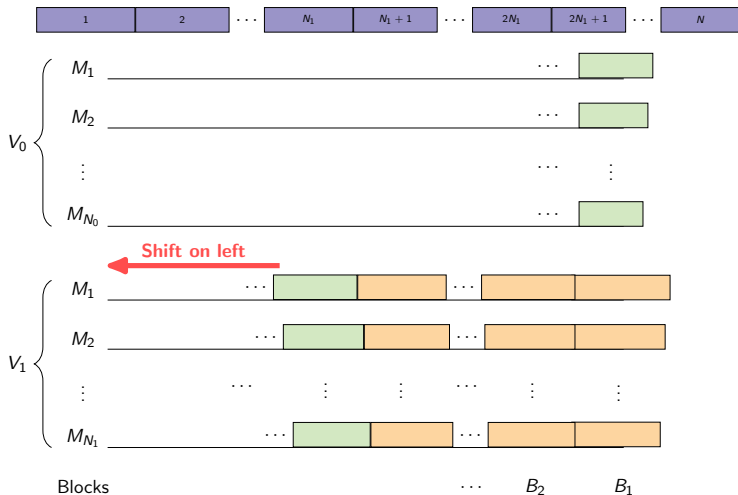
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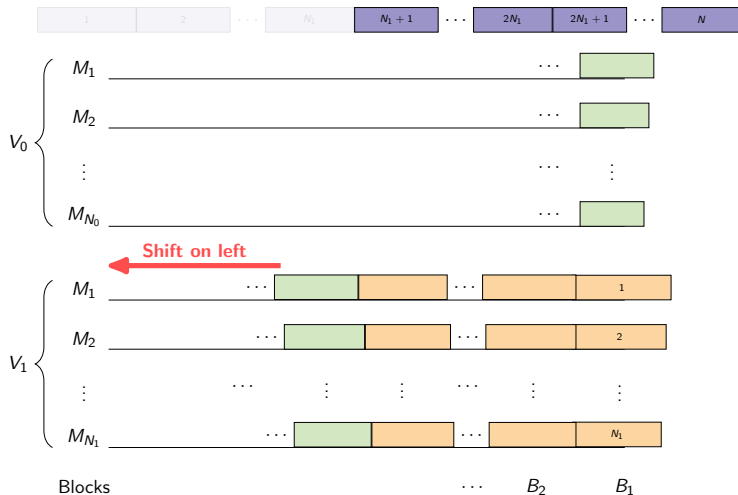
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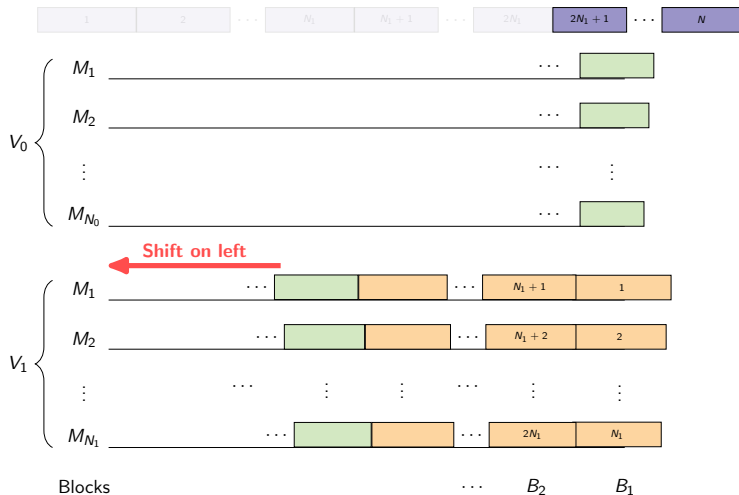
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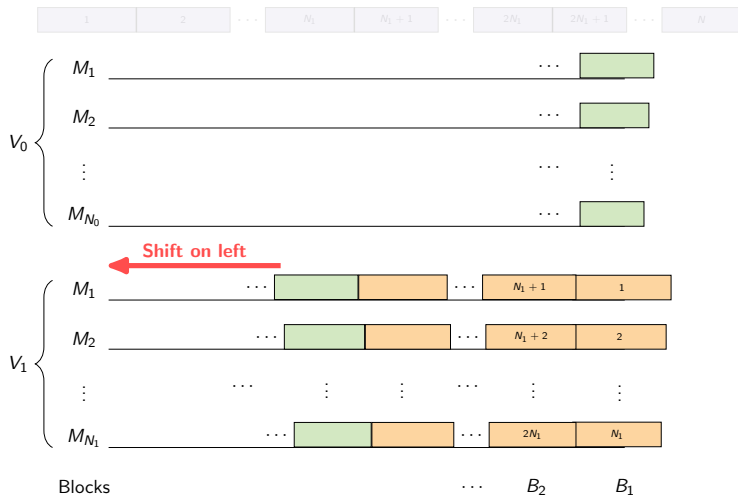
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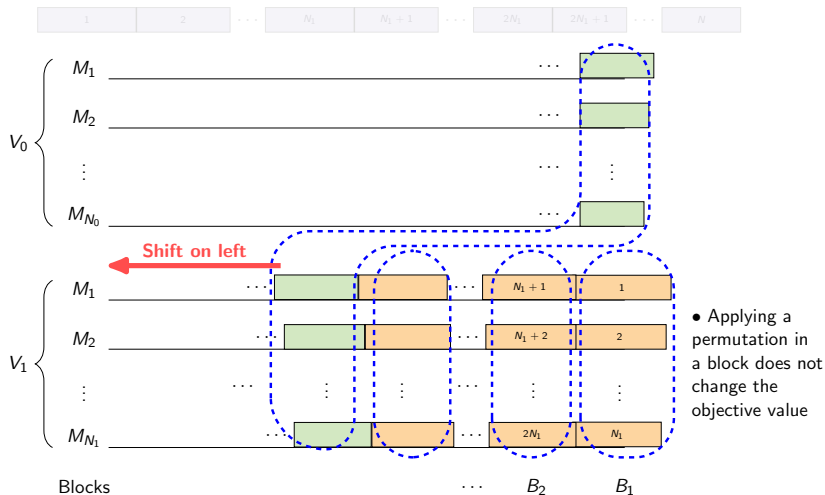




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Bilevel optimisation problems are difficult in terms of complexity .  
What happens for our problem ?

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### Theorem

The problem  $Q|V_i \in \{V_0, V_1\}|Lex \left( \sum_j C_j^F, \sum_j w_j U_j^L \right)$  is  $\mathcal{NP}$ -hard in the strong sense by reduction from the numerical 3-dimensional matching ( $NUM - 3DM$ ).

*Proof omitted*

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- MIP
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### Constraints:

- Block structure: **fill from the RIGHT to LEFT**
- SPT order on each machine schedules
- A job  $j$  that is not "potentially" tardy is necessarily early and fulfils the **due date constraint**:
- Standards constraints: a job cannot be duplicated, only selected jobs can be tardy...

## 2 Exact methods

- MIP
- Column generation

# Exact methods

## Column generation: Master problem

We are extending the approach of Van Den Akker<sup>2</sup>.

The master problem: set-covering problem with an exponential number of binary variables and a linear number of constraints.

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Let  $s$  be a machine schedule (column) defined as a sequence of jobs that can be assigned to any machine.

The jobs in  $s$  must be numbered according to SPT.

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$$a_{j,s} = \begin{cases} 1 & \text{if the job } J_j \text{ is in the schedule } s \\ 0 & \text{otherwise} \end{cases}$$

$$t_{j,s} = \begin{cases} 1 & \text{if the job } J_j \text{ is late in the schedule } s \\ 0 & \text{otherwise} \end{cases}$$

Let  $C_j(s)$  be the completion time of the job  $j$  in the schedule  $s$

$$C_j(s) = \sum_{k=1}^j a_{k,s} \frac{p_k}{V^\gamma} \quad (2)$$



# Exact methods

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We want to compute a LB to the  $\sum_j w_j U_j$ , in the **bilevel scenario**

$\Rightarrow$  integrate the constraint " *when the  $n$  jobs are selected, they must be scheduled to minimize  $\text{Lex}(\sum_j C_j, \sum_j w_j U_j)$*  "

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For one schedule  $s$ , we have the cost  $c_s$  defined as follows:

$$c_s = \sum_{j=1}^N w_j t_{j,s} + \alpha \sum_{j=1}^N C_j(s) \quad (3)$$

where:  $\alpha = \frac{1}{\sum_{j=1}^N (w_j + p_j)}$

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Let be  $x_s \in \{0, 1\}$  where:  $x_s = \begin{cases} 1 & \text{if the column } s \text{ is selected} \\ 0 & \text{otherwise} \end{cases}$

# Exact methods

## Column generation: Master problem

We have the following master's problem formulation :

$$\text{Minimize } \sum_{s^1 \in S_1} c_{s^1} x_{s^1} + \sum_{s^0 \in S_0} c_{s^0} x_{s^0} \quad (4)$$

subject to:

$$\sum_{s^0 \in S_0} a_{j,s^0} x_{s^0} + \sum_{s^1 \in S_1} a_{j,s^1} x_{s^1} \leq 1 \quad \forall j = 1, \dots, N \quad (5a)$$

$$\sum_{s^0 \in S_0} x_{s^0} \leq N_0 \quad (5b)$$

$$\sum_{s^1 \in S_1} x_{s^1} \leq N_1 \quad (5c)$$

$$\sum_{j=1}^N \left( \sum_{s^0 \in S_0} a_{j,s^0} x_{s^0} + \sum_{s^1 \in S_1} a_{j,s^1} x_{s^1} \right) = n \quad (5d)$$

# Exact methods

## Column generation: Pricing problem

We compute the reduced cost  $r_s$  of any machine schedule  $s$  as follow:

$$r_s = c_s - \tilde{\lambda} - \sum_{j=1}^N \lambda_j a_{j,s} - \lambda_{N+1} \sum_{j=1}^N a_{j,s} \quad (6)$$

We solve the *pricing problem*, that involves finding the column with the minimum reduced cost.

- We achieve this through dynamic programming in  $\mathcal{O}\left(\sum_{j=1}^N p_j \times N^2\right)$  time and then by backtracking.
- Using a heuristic that generate column with negative reduced cost.

# Exact methods

## Branch and Bounds

We identify the more fractional job  $j$ .

Let the fractional completion time:

$$\bar{C}(j) = \sum_{s \in S^*(j)} C_j(s) x_s^*$$

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Branching scheme, create 5 nodes :

- $j$  is on high speed machines &  $C_j \leq \lfloor \bar{C}(j) \rfloor$
- $j$  is on high speed machines &  $C_j \geq \lfloor \bar{C}(j) \rfloor + 1$
- $j$  is on low speed machines &  $C_j \leq \lfloor \bar{C}(j) \rfloor$
- $j$  is on low speed machines &  $C_j \geq \lfloor \bar{C}(j) \rfloor + 1$
- Don't take job  $j$



# Exact methods

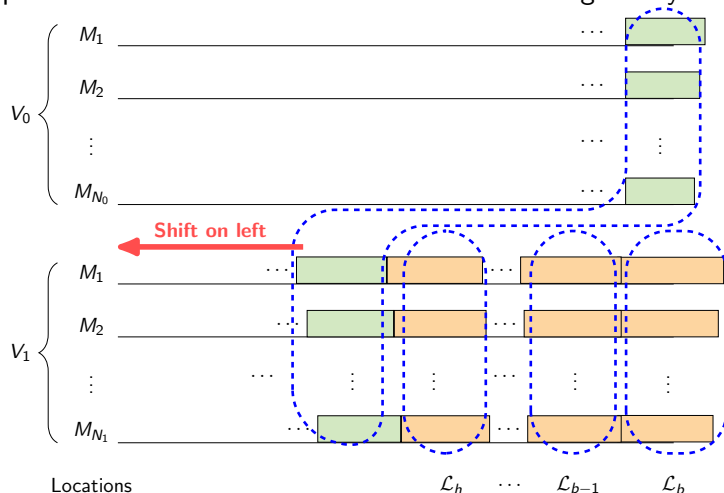
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When we reach a node with  $n$  selected jobs, we **start solving the bilevel** problem and branch on the **location available** given by **block structure**.

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### 3 Computational results

# Computational results

25 types of instances with:  $p_j \in \llbracket 1, 100 \rrbracket$ ,  $w_i \in \llbracket 1, 100 \rrbracket$  and  $d_i \in \llbracket p_j \times (1 - tf - \frac{rdd}{2}), p_j \times (1 - tf + \frac{rdd}{2}) \rrbracket$  where  $tf, rdd \in \{0.2, 0.4, 0.6, 0.8, 1.0\}$

## MIP (125 instances)

MIP				
N	#OPT	$T_{min}$	$T_{avg}$	$T_1$
15	125	0.007	0.012	0.020
20	125	0.017	0.031	0.154
30	125	0.015	0.023	0.035
40	125	0.143	1.85	20.5
50	122	0.312	41.1	600
60	110	0.616	109	600
70	88	1.68	257	600
80	59	4.88	411	600
90	35	17.5	501	600

**Table:** Computing time for MIP method with 2 machines and  $n = \frac{N}{2}$

MIP				
N	#OPT	$T_{min}$	$T_{avg}$	$T_1$
15	125	0.009	0.014	0.039
20	125	0.025	0.057	0.254
30	125	0.074	0.279	2.46
40	125	0.126	5.54	120
50	116	0.38	113	600
60	68	1.29	347	615
70	35	2.66	474	600
80	17	16.1	545	600
90	11	32.3	570	600

**Table:** Computing time for MIP method with 4 machines and  $n = \frac{N}{2}$

## 4 Conclusions

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- Other heuristic methods are being considered, using machine learning to approximate follower's decision
- Other problems are considered with several followers



Thank you for your time!

Questions?

# Exact methods

## MIP formulation: constraints

### Block structure:

All locations, except perhaps those in the first block are occupied (**fill from the RIGHT to LEFT**):

$$\sum_{j=1}^N \sum_{(i,k) \in E \setminus \mathcal{L}_1} x_{ijk} = |E \setminus \mathcal{L}_1| = |E| - |\mathcal{L}_1| \quad (7)$$

The first block may not be totally filled. There are exactly  $|\mathcal{L}_1| + n - |E|$  jobs in the first block:

$$\sum_{j=1}^N \sum_{(i,k) \in \mathcal{L}_1} x_{ijk} = |\mathcal{L}_1| + n - |E| \quad (8)$$

Processing times must increase between 2 consecutive blocks:

$$\forall h < b, \forall (i, k) \in \mathcal{L}_h, \forall (i', k') \in \mathcal{L}_{h+1}, \sum_{j=1}^N x_{ijk} p_j \leq \sum_{j=1}^N x_{i'jk'} p_j \quad (9)$$

# Exact methods

## MIP formulation: constraints

### Block structure:

All locations, except perhaps those in the first block are occupied (**fill from the RIGHT to LEFT**):

$$\sum_{j=1}^N \sum_{(i,k) \in E \setminus \mathcal{L}_1} x_{ijk} = |E \setminus \mathcal{L}_1| = |E| - |\mathcal{L}_1| \quad (7)$$

The first block may not be totally filled. There are exactly  $|\mathcal{L}_1| + n - |E|$  jobs in the first block:

$$\sum_{j=1}^N \sum_{(i,k) \in \mathcal{L}_1} x_{ijk} = |\mathcal{L}_1| + n - |E| \quad (8)$$

Processing times must increase between 2 consecutive blocks:

$$\forall h < b, \forall (j, k) \in \mathcal{L}_h, \forall (i', k') \in \mathcal{L}_{h+1}, \sum_{j=1}^N x_{ijk} p_j \leq \sum_{j=1}^N x_{i'jk'} p_j \quad (9)$$

A job  $j$  that is not "potentially" tardy is necessarily early and checks the **due date constraint**:

$$\forall (i, k) \in E, C_{ik} \leq \sum_{j=1}^N x_{ijk} d_j + \left( \sum_{j=1}^N U_{ijk}^P \right) HV \quad (10)$$

# Exact methods

MIP formulation: constraints

## Other classical constraints:

Computes all completion times:

$$\forall (i, k) \in E, C_{ik} = C_{i,k-1} + \sum_{j=1}^N x_{ijk} \frac{p_j}{V_i} \quad (11)$$

where  $\forall i, C_{i,0} = 0$

Only a present job can be tardy:

$$\forall j, \forall (i, k) \in E, U_{ijk}^P \leq x_{ijk} \quad (12)$$

A job cannot be duplicated:

$$\forall j, \sum_{(i,k) \in E} x_{ijk} \leq 1 \quad (13)$$

Each location can have at most one job:

$$\forall (i, k) \in E, \sum_{j=1}^N x_{ijk} \leq 1 \quad (14)$$