

# Insight on the parameterized complexity of scheduling problems

Designing FPT algorithms with dynamic programming

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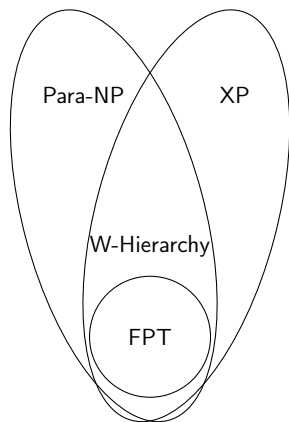
New challenges in scheduling Theory, Aussois 2024



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# Parameterized complexity classes



A parameterized problem of size  $n$  with parameter  $k$  :

## Definition

FPT is the class of problems solvable by a fixed-parameter tractable algorithm with **time complexity**  $\mathcal{O}(f(k) \times \text{poly}(n))$ , where  $f$  is a computable function and  $\text{poly}(n)$  a polynome of  $n$ .

# Challenges for scheduling problems

- Which problems can be shown FPT?
- Which relevant structural parameters for scheduling?
- How to design FPT algorithms?

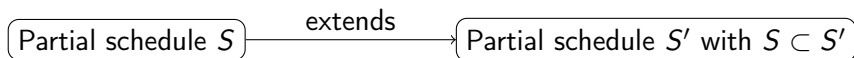
# Challenges for scheduling problems

- Which problems can be shown FPT?
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Two main techniques:

- 1 ILP with special structure (for example Hermelin, Karhi, Pinedo, and Shabtay, 2021)
- 2 Dynamic programming

# Dynamic programming for scheduling



- The second partial schedule extends the first one.
- A state records the minimum information needed from  $S$  to build feasible extensions  $S'$
- Dynamic programming algorithm starts from an empty schedule and builds a state graph.

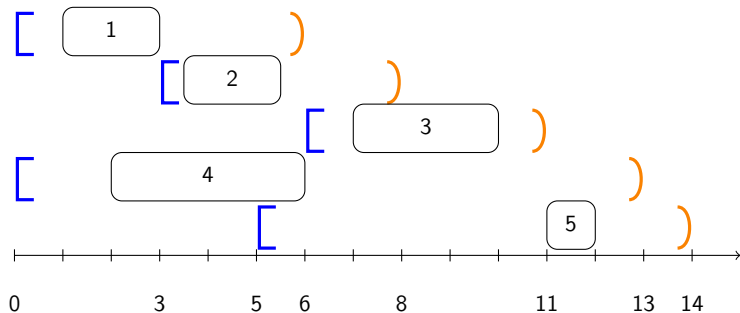
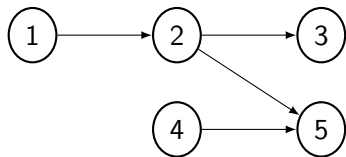
## Key arguments

- Key argument 1: A schedule is defined by a path on the state graph.
- Feasibility: finding a path from a source to a sink
- Optimization: finding a minimum cost path from a source to a sink
- Key Argument 2 : Bounding the number of states.

# Parameters: A single machine problem example

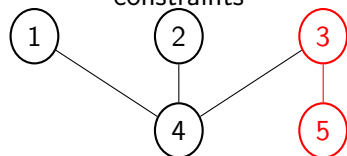
We consider  $1|prec, r_i, \bar{d}_i|C_{max}$   
usual parameters:

- $C_{max}$
- Slack  $\sigma = \max_i (\bar{d}_i - r_i - p_i)$
- $p_{max} = \max_i p_i$



# Compatibility graph

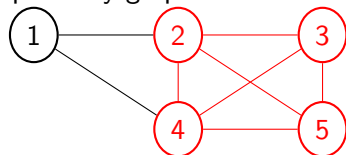
Compatibility graph of precedence constraints



$maxclique = w(G)$  (width of the precedence graph)

$P|_{prec, p_j = 1} | C_{max} \leq D$  is  
 XNLP-complete [Bodlaender et al,  
 2022]

Compatibility graph of time windows



$maxclique$  is the nb of overlapping intervals

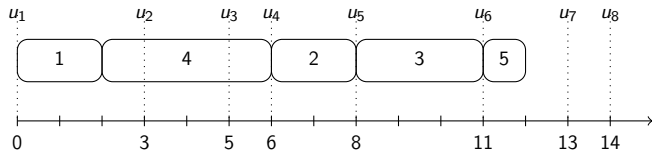
$Pathwidth = maxclique - 1 = 3$

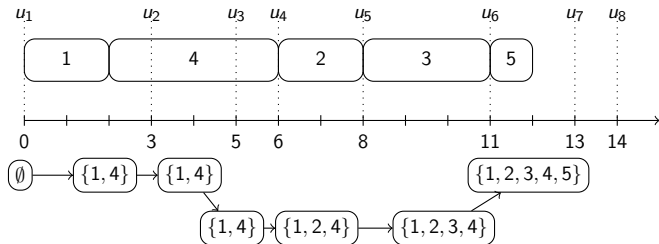
$P|_{prec, p_i = 1, r_i, d_i} \star$  FPT [Munier  
 Kordon 2021]

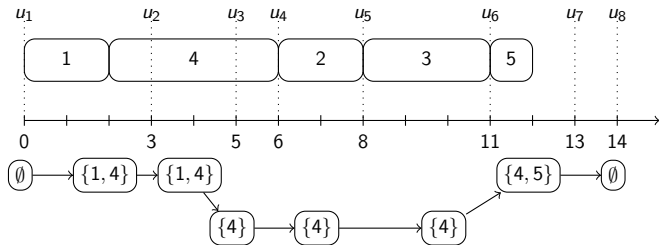


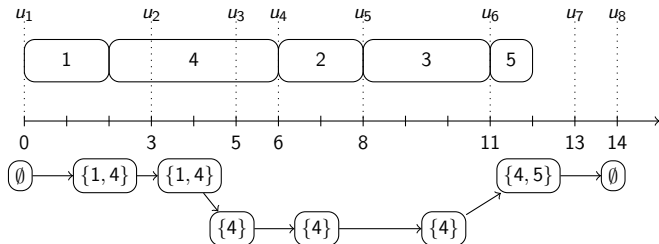
Single machine with pathwidth  $\mu$ 

Sorting the end points of time windows  $u_1, \dots, u_K$



Single machine with pathwidth  $\mu$ Recording the set of scheduled jobs before each  $u_{k+1}$ 

Single machine with pathwidth  $\mu$ Removing jobs that must end before  $u_{k+1}$ 

Single machine with pathwidth  $\mu$ State  $L_k$  of stage  $k$ 

$L_k$  Set of jobs started before  $u_{k+1}$  and with deadline  $> u_{k+1}$  Evaluation of a state: minimum makespan of a feasible schedule of  $L_k$  plus jobs with deadlines  $\leq u_{k+1}$ .

# Results

Remark: nb of jobs  $i$  with  $r_i \leq u_k < u_{k+1} < d_i$  is bounded by  $\mu + 1$   
 In a state  $L_k$  of stage  $k$ ,  $L_k$  is a subset of a set of size at most  $\mu + 1$ .

## Theorem

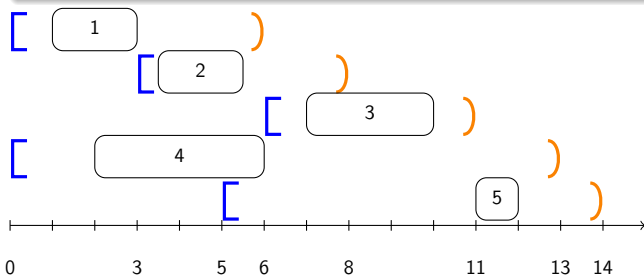
- The number of states is  $\mathcal{O}(n \cdot 2^{\mu+2})$ .
- The number of arcs is  $\mathcal{O}(n \cdot 2^{2(\mu+2)})$
- Finding the optimal schedule is FPT with respect to  $\mu$

## q-proper level parameter

Proper set  $\Pi_i$  of a job  $i$ Set of jobs whose interval strictly encloses  $[r_i, \bar{d}_i)$ 

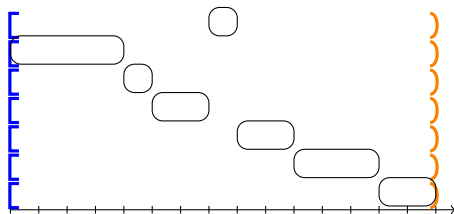
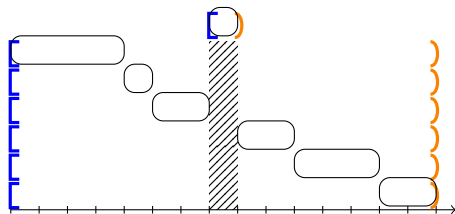
## q-proper parameter

$$q = \max_i |\Pi_i|$$



$$\begin{aligned} \Pi_1 &= \Pi_2 = \\ \Pi_3 &= \emptyset, \\ \Pi_2 &= \{4\}, \\ \Pi_3 &= \{4, 5\}, \\ q &= 2 \end{aligned}$$

## q-proper level motivation


 $\mu = n - 1, q = 0$ , Easy problem

 $\mu = n - 1, q = n - 1$ , NP-hard problem

# Assumptions

## Consistency of time windows

We assume release times and deadlines consistent with precedence constraints: if  $i \rightarrow j$  then  $r_j \geq r_i + p_i \wedge \bar{d}_i \leq \bar{d}_j - p_j$

## Indexation

We assume jobs are indexed such that : if  $i < j$  then

$$\bar{d}_i < \bar{d}_j \vee ((\bar{d}_i = \bar{d}_j) \wedge (r_i \leq r_j))$$



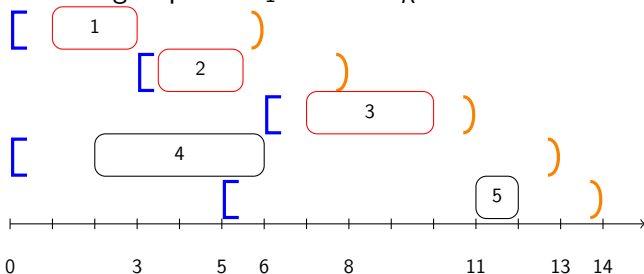
# Summits

## Definition

A summit is a job  $i$  such that the time window  $[r_i, \bar{d}_i)$  does not strictly enclose another time window:

$$\forall j \neq i, i \notin \Pi_j$$

Increasing sequence  $s_1 < \dots < s_K$  of summits



Summits

$$s_1 = 1$$

$$s_2 = 2$$

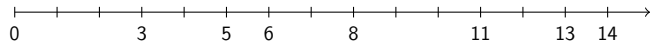
$$s_3 = 3$$

# Structure of a schedule : a dominance rule

## Dominance rule lemma

If there is a feasible schedule then there is a minimum makespan schedule  $\sigma$  of the form  $T_1 s_1 \dots T_N s_N T_{N+1}$  where:

- every job in sequence  $T_1$  is in set  $\Pi_{s_1}$ ,
- for all  $k$  in  $[2, N]$  every job in sequence  $T_k$  is in set  $\Pi_{s_{k-1}} \cup \Pi_{s_k}$ , and
- every job in sequence  $T_{N+1}$  is in set  $\Pi_{s_N}$ .



$$T_1 = \emptyset, T_2 = \{4\}, T_3 = \emptyset, T_4 = \{5\}$$

# Dominance rule detailed

## Lemma

In a dominant schedule  $T_1 s_1 \dots T_N s_N T_{N+1}$ ,

- All jobs with index  $< s_k$  are performed before  $s_k$
- $T_k = C_{k-1} A_k B_k$

Where

- $C_{k-1} = T_k - \Pi_{s_k}$  scheduled by increasing index
- $A_k = T_k \cap \Pi_{s_{k-1}} \cap \Pi_{s_k}$  scheduled by increasing index
- $B_k = T_k - \Pi_{s_{k-1}}$  scheduled by increasing  $r_i$

Inspired by a similar decomposition for problems without prec:

J. Erschler, Gérard Fontan, Colette Mercé, and François Roubellat. *A new dominance concept in scheduling  $n$  jobs on a single machine with ready times and due dates*. Oper. Res., 31(1):114–127, 1 (1983)

# Dynamic programming

- A state of stage  $k$  records the minimum information about jobs performed before summit  $s_k$
- Only jobs with index  $> s_k$  are useful to store

## Lemma

$$L_k = \bigcup_{1 \leq h \leq k} T_h \cap \{s_k + 1, \dots, n\}$$

$$L_k \subset \Lambda_k = \bigcup_{1 \leq h \leq k} \Pi_{s_h} \cap \{s_k + 1, \dots, n\} = \Pi_{s_k}$$



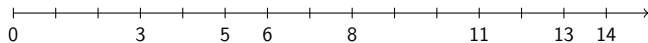
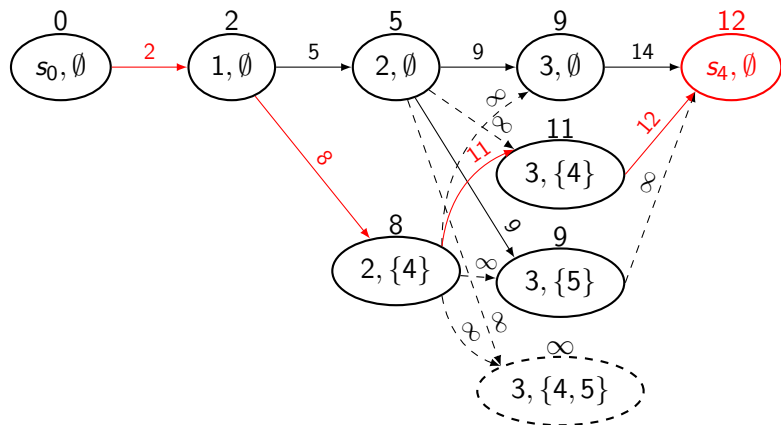
# Results

- Cardinality of the set  $\Lambda_k$  bounded by  $q$
- Evaluation of a state  $L_k$  of stage  $k$  : minimum makespan of a schedule of  $\{1, \dots, s_k\} \cup L_k$  ending by  $s_k$ .
- This evaluation can be performed knowing the evaluation of states of the lower stages.

## Theorem

- *The number of states is  $\mathcal{O}(n2^q)$*
- *Solving  $1|prec, r_i, \bar{d}_i|C_{max}$  is FPT with respect to  $q$ .*

## State graph



## Relations between parameters

### Lemma

*In any instance of  $1|prec, r_i, \bar{d}_i|*$ ,  $q \leq \mu$ .*

### Corollary

A FPT algorithm for  $q$  there is a FPT algorithm for parameter  $\mu$ . If a problem is para-NP-hard for parameter  $\mu$ , it will be also for parameter  $q$

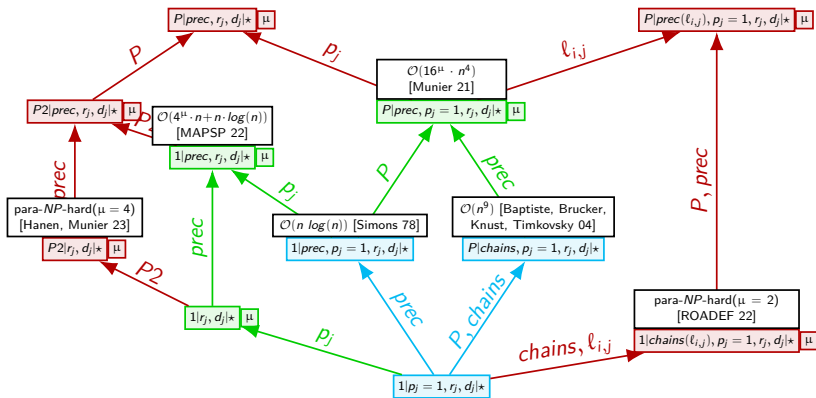
### Lemma

*In any feasible instance of  $1|prec, r_i, \bar{d}_i|*$   $\mu \leq 2\sigma$  (slack);*

### Corollary

A FPT algorithm for  $\mu$  is a FPT algorithm for parameter  $\sigma$ .

## Map

Parameter  $\mu$ 



## Concluding remarks

- We provided two FPT algorithms for  $1|prec, r_i, \bar{d}_i|C_{max}$  with parameters  $\mu$  and  $q$ .
- Currently studying  $P|tree, p_i = 1, r_i, \bar{d}_i|*$  with respect to  $q$
- Building maps for scheduling relevant parameters

Challenging questions for the study of parameterized complexity of scheduling problems :

- Are FPT algorithms efficient in practice?
- What about Kernelization techniques?

Maher Mallem, Claire Hanen, Alix Munier-Kordon:

*Parameterized Complexity of Single-machine Scheduling with Precedence, Release Dates and Deadlines, MAPSP2022*

A new structural parameter on single machine scheduling with release dates and deadlines, ISCO 2024