

# A Serial Schedule-Generation Scheme for Preemptive Project Scheduling Problems with Generalized Precedence Relations and Regular Min-Max Criteria

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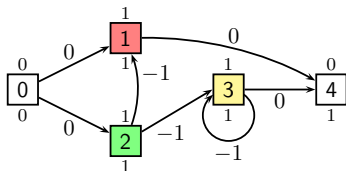
## Outline

- 1 Preemptive project scheduling problem
  - Problem statement
  - Descriptive model
- 2 Preemptive schedule-generation scheme
  - Motivation
  - Iteration of threshold problems
  - Schedule-generation scheme for given upper bound
  - Enhancements
- 3 Performance analysis
- 4 Conclusions

## Preemptive project scheduling problem

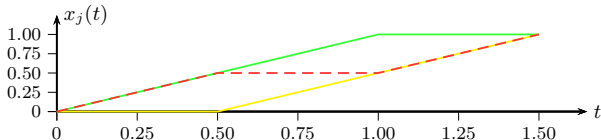
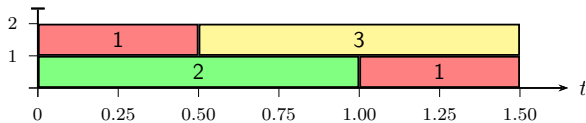
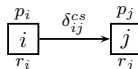
- Project consists of **preemptive activities**  $i \in V$  with durations  $p_i$
- Activities  $i$  require  $r_{ik}$  units of **renewable resources**  $k \in \mathcal{R}$  with capacities  $R_k$
- **Minimum time lags**  $\delta_{ij}^{cs}$  between completion time  $C_i$  of activity  $i$  and start time  $S_j$  of activity  $j$  with  $(i, j) \in E \subseteq V \times V$ 
  - $-\delta_{ij}^{cs} > 0$ : maximum start-to-completion time lag between  $j$  and  $i$
  - $-\delta_{ii}^{cs} \geq p_i$ : maximum makespan of  $i$
- **Sought: feasible schedule**  $x : t \mapsto (x_j(t))_{j \in V}$  **minimizing objective function**  $f_{max}(C) = \max_{j \in V} f_j(C_j)$  with regular  $f_j$ 
  - $x_j(t)$ : percentage of activity  $j$  processed by time  $t \geq 0$
  - $S_j = \sup\{t \geq 0 \mid x_j(t) = 0\}$ ,  $C_j = \inf\{t \geq 0 \mid x_j(t) = 1\}$
  - $y_j(t) := p_j \cdot \frac{d^+ x_j}{dt}(t) = \begin{cases} 1, & \text{if } j \text{ is in progress at time } t \\ 0, & \text{otherwise} \end{cases}$

## Example



$$R = 2, f_{max} = C_{max}$$

Legend:



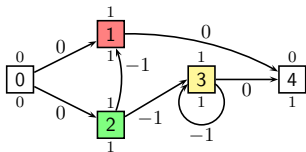
## Conceptual model

$$(P) \begin{cases} \text{Minimize} & f_{max}(C) = \max_{j \in V} f_j(C_j) \\ \text{subject to} & r_k(t) := \sum_{i \in V} r_{ik} y_i(t) \leq R_k \quad (k \in \mathcal{R}; t \geq 0) \\ & S_j \geq C_i + \delta_{ij}^{cs} \quad ((i, j) \in E) \\ & S_j \geq 0, C_j \geq S_j + p_j \quad (j \in V) \end{cases}$$

- **Non-preemptive problem** contained as a special case ( $\delta_{ii}^{cs} = -p_i$ )
- Feasibility variant **strongly NP-hard**
- By convention  $\delta_{ij}^{cs} \leq 0$  ( $\delta_{ij}^{cs} > 0$  as dummy  $h$  with  $p_h = \delta_{ij}^{cs}$ )
- Each activity interrupted at most  $n - 1$  times
- **Schedule encoded as set  $\Sigma$**  of at most  $n^2$  **triples**  $(j, s_j, c_j)$  defining time intervals  $[s_j, c_j[$  during which parts of activities  $j$  are in progress

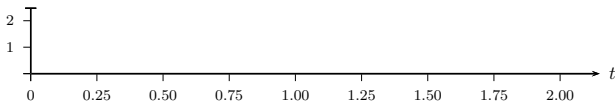
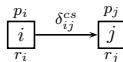
## Why naïve adaption of SGS by Franck et al. (2001) fails

- In each iteration select eligible activity  $j^*$  to be started or resumed
- Schedule  $j^*$  at earliest time- and resource-feasible point in time
- Allow to suspend execution of  $j^*$  at next decision point (release, start, or completion of some activity)



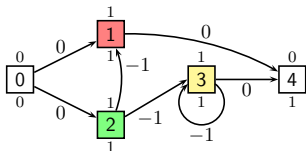
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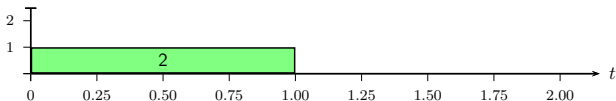
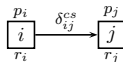
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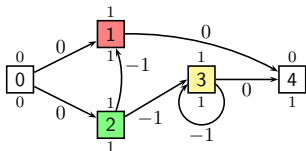
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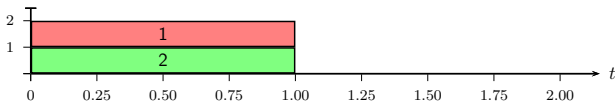
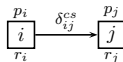
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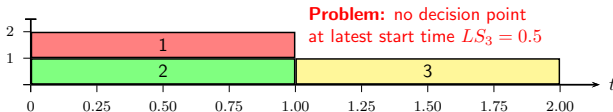
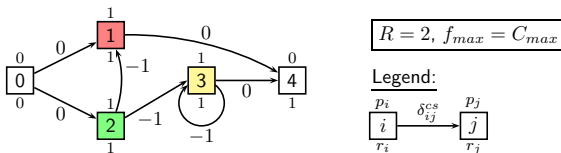
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- In each iteration select eligible activity  $j^*$  to be started or resumed
- Schedule  $j^*$  at earliest time- and resource-feasible point in time
- Allow to suspend execution of  $j^*$  at next decision point (release, start, or completion of some activity)



- Need additional decision points arising from objective function value

## The solution: iterate threshold instances

- Given upper bound  $v$  on objective function value  $f_{max}(C)$ ,

$$f_{max}(C) = \max_{j \in V} f_j(C_j) \leq v \Leftrightarrow f_j(C_j) \leq v \quad (j \in V)$$

$$\Leftrightarrow C_j \leq f_j^{-1}(v) \quad (j \in V)$$

with  $f_j^{-1}(v) := \sup\{C_j \mid f_j(C_j) \leq v\}$  and  $\sup \emptyset := -\infty$

- Introducing  $-\delta_{j0}^{cs} = f_j^{-1}(v)$  for  $j \in V$  ensures  $f_{max}(C) \leq v$
- Perform binary search for smallest  $v$  over interval  $[lb, ub]$ 
  - Start with  $v = \frac{lb+ub}{2}$
  - Apply preemptive SGS to instance with  $-\delta_{j0}^{cs} = f_j^{-1}(v)$ 
    - If feasible schedule is found:  $f_{max}^* \leq f_{max}(C) \leq v$ ,  
put  $ub := f_{max}(C)$
    - Otherwise: put  $lb := v$
  - Recursively continue search on  $[lb, ub]$  until  $ub - lb \leq \varepsilon$

## Basic principle of preemptive SGS

- Translate **upper bound**  $v$  into **time lags**  $-\delta_{j0}^{cs} = f_j^{-1}(v)$  for  $j \in V$
- Compute distance matrix  $D^{cs} = (d_{ij}^{cs})_{i,j \in V}$  of **transitive time lags**
  - $ES_j = d_{0j}^{cs}$ : earliest time-feasible start time of activity  $j$
  - $LC_j = -d_{j0}^{cs}$ : latest time-feasible completion times of activity  $j$
- Initialize earliest time-feasible start times  $es_j := ES_j$  of **pending parts** of activities  $j$
- Activity  $j$  **eligible** for being started or resumed if all predecessors  $i \in Pred^{\prec}(j)$  have been completed, i. e.,  $Pred^{\prec}(j) \subseteq \mathcal{C}$ , where

$$i \prec j \Leftrightarrow i \text{ must be started no later than } j \text{ but not vice versa}$$

$$\Leftrightarrow \max\{d_{ij}^{cs}, d_{i0}^{cs} + es_j\} + p_i \geq 0 \wedge \max\{d_{ji}^{cs}, d_{j0}^{cs} + es_i\} + p_j < 0$$

- **Select** some **eligible activity**  $j^* \in \mathcal{E}$  by applying priority rule  $\pi$

## Basic principle of preemptive SGS

- Determine earliest resource-feasible start time  $s_{j^*} \geq es_{j^*}$
- If  $s_{j^*} \leq LC_{j^*} - p_{j^*}$ , execute  $j^*$  up to next decision point  $c_{j^*}$
- Set  $\mathcal{D}$  of decision points contains
  - earliest completion time  $s_{j^*} + p_{j^*}$  of  $j^*$
  - start and completion times  $s_j, c_j$  of executed activity parts
  - earliest and latest start and completion times  
 $es_j, es_j + p_j, LC_j - p_j, LC_j$  of pending activity parts
- Decrease residual processing time  $p_{j^*}$  by  $c_{j^*} - s_{j^*}$  and update earliest start and latest completion times of pending activities
- If  $s_{j^*} > LC_{j^*} - p_{j^*}$ , try to repair the schedule by calling unscheduling procedure
- Proceed until all activities  $j \in V$  have been entirely added to the schedule, i. e.,  $\mathcal{C} = V$

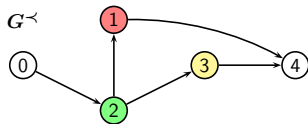
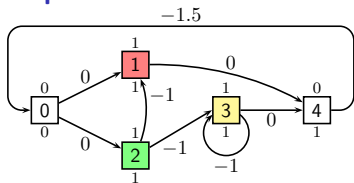
## Basic principle of unscheduling procedure

- If  $s_{j^*} > LC_{j^*} - p_{j^*}$ , SGS has run into a **deadlock**
- **Clear parts of the schedule** to remove deadlock
- Identify set  $\mathcal{U}$  of started **activities**  $j \in \mathcal{S}$  that determined the **latest completion time**  $LC_{j^*}$  of activity  $j^*$

$$\mathcal{U} = \{j \in \mathcal{S} \mid LC_{j^*} = S_j - d_{j^*j}^{cs}\}$$

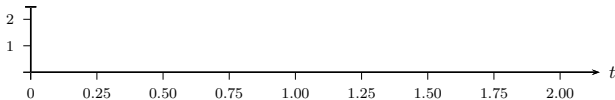
- **Remove all activity parts** with  $s_j \geq \min_{i \in \mathcal{U}} S_i$  from the schedule
- **Delay activities**  $j \in \mathcal{U}$  by  $\Delta = s_{j^*} + p_{j^*} - LC_{j^*}$ : put  $es_j := S_j + \Delta$
- If  $es_j + p_j > -d_{j0}^{cs}$ , deadlock could not be resolved: **terminate**
- Otherwise **return to preemptive SGS**

## Example

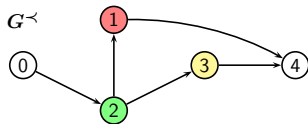
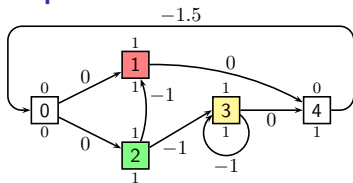


$$D^{cs} = \begin{pmatrix} 0 & 0 & 0 & 0 & 1 \\ -1.5 & -1.5 & -1.5 & -1.5 & 0 \\ -1.5 & -1 & -1.5 & -1 & 0 \\ -1.5 & -1.5 & -1.5 & -1 & 0 \\ -1.5 & -1.5 & -1.5 & -1.5 & 0 \end{pmatrix}$$

$j$	1	2	3	$\mathcal{S} = \{0\}$
$es_j$	0	0	0	$\mathcal{C} = \{0\}$
$LC_j$	1.5	1.5	1.5	
$S_j$	-	-	-	
$C_j$	-	-	-	

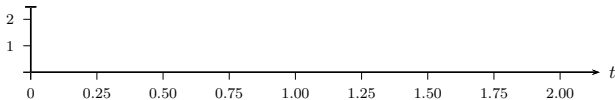


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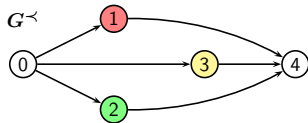
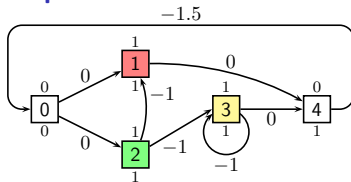
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$j$	1	2	3	$\mathcal{S} = \{0\}$
$es_j$	0	0	0	$\mathcal{C} = \{0\}$
$LC_j$	1.5	1.5	1.5	
$S_j$	-	-	-	
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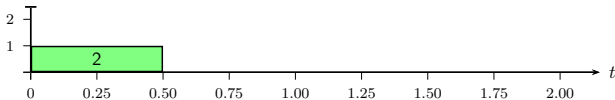
$$\mathcal{E} = \{2\}, j^* = 2, s_2 = 0, \mathcal{D} = \{0, 0.5, 1, 1.5\}, c_2 = 0.5, p_2 = 0.5$$

## Example



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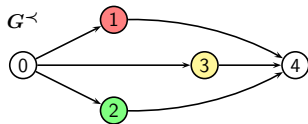
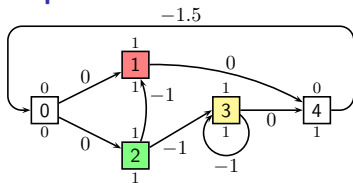
$j$	1	2	3	$\mathcal{S} = \{0, 2\}$
$es_j$	0	0.5	0	$\mathcal{C} = \{0\}$
$LC_j$	1.5	1.5	1.5	
$S_j$	-	0	-	
$C_j$	-	-	-	



$$\mathcal{E} = \{1, 2, 3\}, j^* = 3, s_3 = 0, \mathcal{D} = \{0, 0.5, 1, 1.5\}, c_3 = 0.5, p_3 = 0.5$$

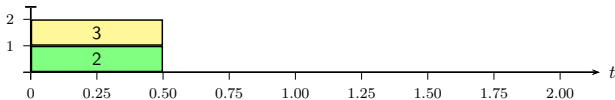


## Example



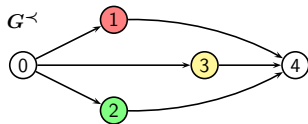
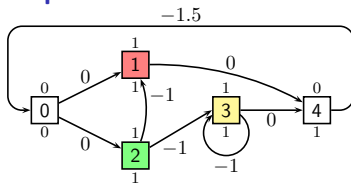
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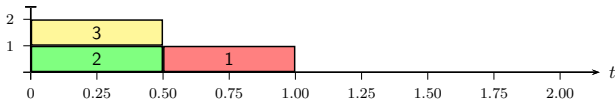
$$\mathcal{E} = \{1, 2, 3\}, j^* = 1, s_1 = 0.5, \mathcal{D} = \{0, 0.5, 1, 1.5\}, c_1 = 1, p_1 = 0.5$$

## Example



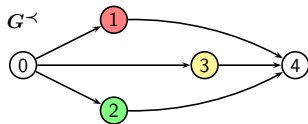
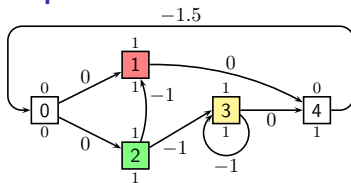
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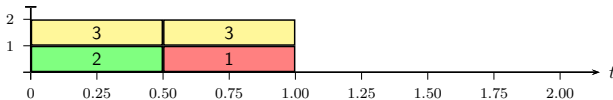
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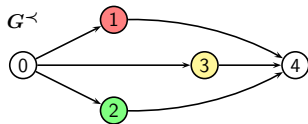
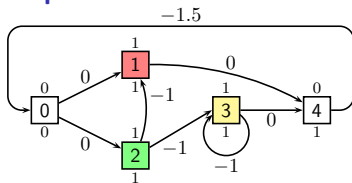
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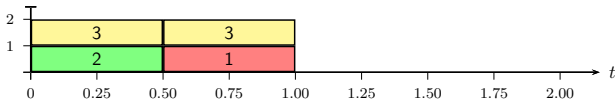
$$\mathcal{E} = \{1, 2\}, j^* = 2, s_2 = 1, \mathcal{D} = \{0, 0.5, 1, 1.5\}$$

## Example



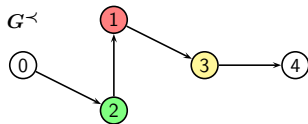
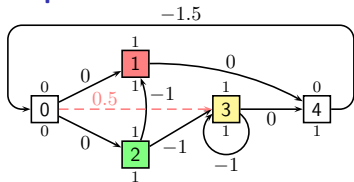
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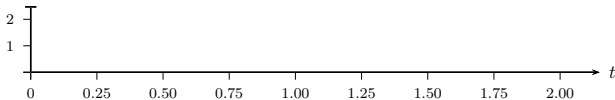
$s_2 + p_2 > LC_2$ : *unschedule*(2, 0.5),  $\mathcal{U} = \{3\}$ ,  $t^* = S_3 = 0$ ,  $es_3 = 0.5$

## Example



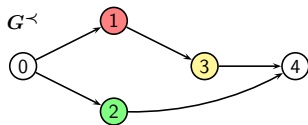
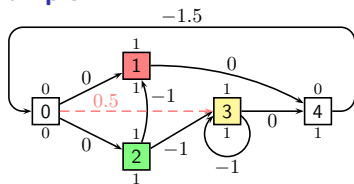
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$S_j$	-	-	-	
$C_j$	-	-	-	



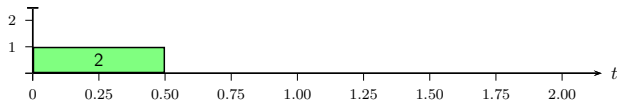
$$\mathcal{E} = \{2\}, j^* = 2, s_2 = 0, \mathcal{D} = \{0, 0.5, 1, 1.5\}, c_2 = 0.5, p_2 = 0.5$$

## Example



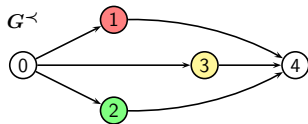
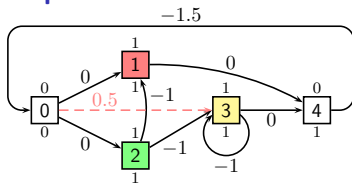
$$D^{cs} = \begin{pmatrix} 0 & 0 & 0 & 0 & 1 \\ -1.5 & -1.5 & -1.5 & -1.5 & 0 \\ -1.5 & -1 & -1.5 & -1 & 0 \\ -1.5 & -1.5 & -1.5 & -1 & 0 \\ -1.5 & -1.5 & -1.5 & -1.5 & 0 \end{pmatrix}$$

$j$	1	2	3	$\mathcal{S} = \{0, 2\}$
$es_j$	0	0.5	0.5	$\mathcal{C} = \{0\}$
$LC_j$	1.5	1.5	1.5	
$S_j$	-	0	-	
$C_j$	-	-	-	



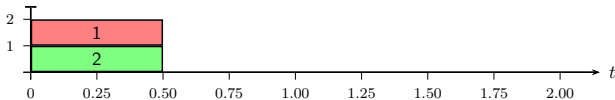
$$\mathcal{E} = \{1, 2\}, j^* = 1, s_1 = 0, \mathcal{D} = \{0, 0.5, 1, 1.5\}, c_1 = 0.5, p_1 = 0.5$$

## Example



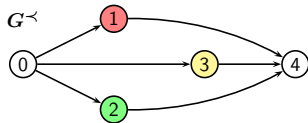
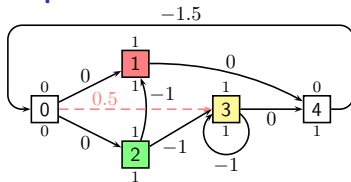
$$D^{cs} = \begin{pmatrix} 0 & 0 & 0 & 0 & 1 \\ -1.5 & -1.5 & -1.5 & -1.5 & 0 \\ -1.5 & -1 & -1.5 & -1 & 0 \\ -1.5 & -1.5 & -1.5 & -1 & 0 \\ -1.5 & -1.5 & -1.5 & -1.5 & 0 \end{pmatrix}$$

$j$	1	2	3	$\mathcal{S} = \{0, 1, 2\}$
$es_j$	0.5	0.5	0.5	$\mathcal{C} = \{0\}$
$LC_j$	1.5	1	1.5	
$S_j$	0	0	-	
$C_j$	-	-	-	



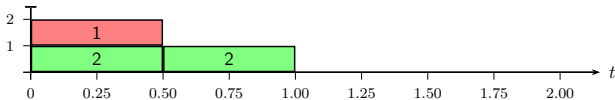
$$\mathcal{E} = \{1, 2, 3\}, j^* = 2, s_2 = 0.5, \mathcal{D} = \{0, 0.5, 1, 1.5\}, c_2 = 1, p_2 = 0$$

## Example



$$D^{cs} = \begin{pmatrix} 0 & 0 & 0 & 0 & 1 \\ -1.5 & -1.5 & -1.5 & -1.5 & 0 \\ -1.5 & -1 & -1.5 & -1 & 0 \\ -1.5 & -1.5 & -1.5 & -1 & 0 \\ -1.5 & -1.5 & -1.5 & -1.5 & 0 \end{pmatrix}$$

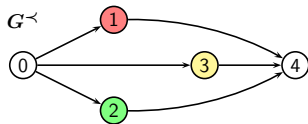
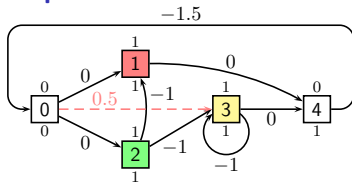
$j$	1	2	3	$\mathcal{S} = \{0, 1, 2\}$
$es_j$	0.5	0.5	<b>0.5</b>	$\mathcal{C} = \{0, 2\}$
$LC_j$	1.5	1	1.5	
$S_j$	0	0	-	
$C_j$	-	1	-	



$$\mathcal{E} = \{1, 3\}, j^* = 3, s_3 = 0.5, \mathcal{D} = \{0, 0.5, 1, 1.5\}, c_3 = 1, p_3 = 0.5$$

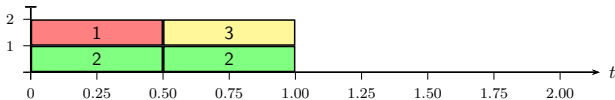


## Example



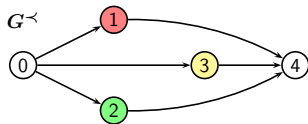
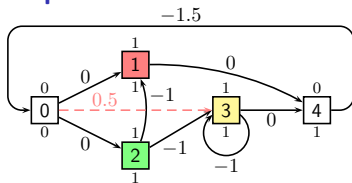
$$D^{cs} = \begin{pmatrix} 0 & 0 & 0 & 0 & 1 \\ -1.5 & -1.5 & -1.5 & -1.5 & 0 \\ -1.5 & -1 & -1.5 & -1 & 0 \\ -1.5 & -1.5 & -1.5 & -1 & 0 \\ -1.5 & -1.5 & -1.5 & -1.5 & 0 \end{pmatrix}$$

$j$	1	2	3	$\mathcal{S} = \{0, 1, 2, 3\}$
$es_j$	0.5	0.5	1	$\mathcal{C} = \{0, 2\}$
$LC_j$	1.5	1	1.5	
$S_j$	0	0	0.5	
$C_j$	-	1	-	



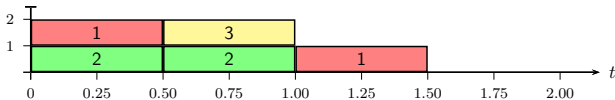
$$\mathcal{E} = \{1, 3\}, j^* = 1, s_1 = 1, \mathcal{D} = \{0, 0.5, 1, 1.5\}, c_1 = 1.5, p_1 = 0$$

## Example



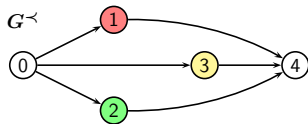
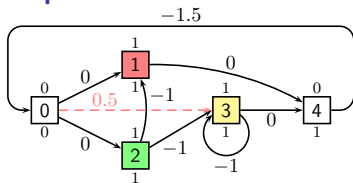
$$D^{cs} = \begin{pmatrix} 0 & 0 & 0 & 0 & 1 \\ -1.5 & -1.5 & -1.5 & -1.5 & 0 \\ -1.5 & -1 & -1.5 & -1 & 0 \\ -1.5 & -1.5 & -1.5 & -1 & 0 \\ -1.5 & -1.5 & -1.5 & -1.5 & 0 \end{pmatrix}$$

$j$	1	2	3	$\mathcal{S} = \{0, 1, 2, 3\}$
$es_j$	0.5	0.5	1	$\mathcal{C} = \{0, 1, 2\}$
$LC_j$	1.5	1	1.5	
$S_j$	0	0	0.5	
$C_j$	1.5	1	-	



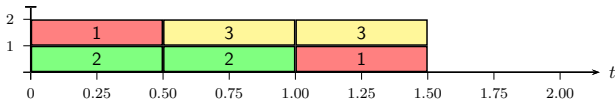
$$\mathcal{E} = \{3\}, j^* = 3, s_3 = 1, \mathcal{D} = \{0, 0.5, 1, 1.5\}, c_3 = 1.5, p_3 = 0$$

## Example



$$D^{cs} = \begin{pmatrix} 0 & 0 & 0 & 0 & 1 \\ -1.5 & -1.5 & -1.5 & -1.5 & 0 \\ -1.5 & -1 & -1.5 & -1 & 0 \\ -1.5 & -1.5 & -1.5 & -1 & 0 \\ -1.5 & -1.5 & -1.5 & -1.5 & 0 \end{pmatrix}$$

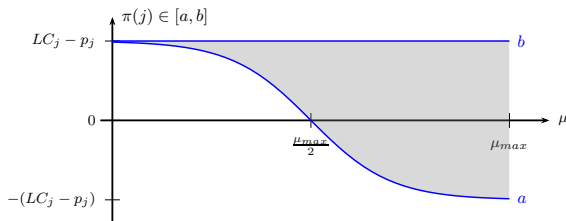
$j$	1	2	3	$\mathcal{S} = \{0, 1, 2, 3\}$
$es_j$	0.5	0.5	1	$\mathcal{C} = \{0, 1, 2, 3\}$
$LC_j$	1.5	1	1.5	
$S_j$	0	0	0.5	
$C_j$	1.5	1	1.5	



## Randomized multi-start procedure

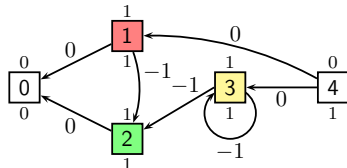
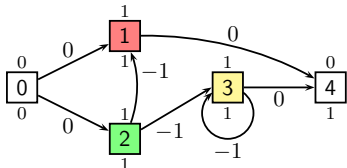
- In **multi-start procedure**, binary search algorithm can be run with different priority rules  $\pi$
- Start with **minimum latest start time rule**:  $\pi(j) = LC_j - p_j$
- Stepwise **increase random bias** according to sigmoid function  $\sigma(\mu)$  for passes  $\mu = 1, \dots, \mu_{max}$

$$\pi(j) = (LC_j - p_j) \cdot (1 - 2\sigma(\mu) \cdot rnd) \text{ with } \sigma(\mu) = \frac{1}{1 + e^{-\alpha \cdot (\mu - \mu_{max}/2)}}$$



## Forward and backward scheduling

- Scheduling pending activity parts at their **earliest** feasible start times corresponds to **forward scheduling**
- **Backward scheduling** can be emulated by applying SGS on “inverted” project network (Hanzálek and Šůcha 2009)
- Substitute time lags  $\delta_{ij}^{cs}$  into time lags  $\bar{\delta}_{ji}^{cs} = \delta_{ij}^{cs}$
- Original and inverted instances  $I, \bar{I}$  of threshold problem **equivalent**
- Schedule  $\bar{\Sigma}$  for  $\bar{I}$  transformed into schedule  $\Sigma$  for  $I$  by **replacing** triples  $(j, \bar{s}_j, \bar{c}_j) \in \bar{\Sigma}$  with triples  $(j, \bar{C}_{max} - \bar{c}_j, \bar{C}_{max} - \bar{s}_j)$



## Experimental performance analysis

- Testsets UBO10, UBO20, UBO50, UB100 with 90 instances each (Franck et al. 2001)
- Objective function **project duration**  $f_{max}(C) = C_{max}$
- Lower and upper bounds for **preemptive** instances computed with MILP formulation (S. and Paetz 2015)
- Upper bounds for **non-preemptive instances** according to benchmark files on ProGen/max homepage<sup>1</sup>
- **Tested configurations**
  - Forward scheduling:  $\mu_{max} = 100, \varepsilon = 10^{-4}$
  - Backward and forward scheduling:  $\mu_{max} = 50, \varepsilon = 10^{-4}$
- Multi-start procedure coded in **C#**
- Intel i5 PC with **3.4 GHz clock pulse** and 8 GB RAM, OS Win 7 Professional 64 Bit

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<sup>1</sup> [www.wiwi.tu-clausthal.de/abteilungen/produktion/forschung/schwerpunkte/project-generator/](http://www.wiwi.tu-clausthal.de/abteilungen/produktion/forschung/schwerpunkte/project-generator/)

## Computational results

### Forward scheduling, 100 passes

	$\Delta_{milp}$	$\Delta_{npmtn}$	$p_{imp}$	$p_{fnd}$	$t_{cpu}$	$\#it$
$n = 10$	0.50 %	-1.71 %	27.78 %	4.44 %	0.14 s	1668
$n = 20$	0.06 %	-1.68 %	36.67 %	10.00 %	0.37 s	1746
$n = 50$	n/a	-1.35 %	44.44 %	3.33 %	3.15 s	1728
$n = 100$	n/a	-1.16 %	46.67 %	0.00 %	24.50 s	1837

### Forward and backward scheduling, 50 passes

	$\Delta_{milp}$	$\Delta_{npmtn}$	$p_{imp}$	$p_{fnd}$	$t_{cpu}$	$\#it$
$n = 10$	0.17 %	-2.04 %	31.11 %	4.44 %	0.14 s	1662
$n = 20$	-1.08 %	-2.66 %	46.67 %	11.11 %	0.34 s	1715
$n = 50$	n/a	-2.12 %	52.22 %	2.22 %	3.37 s	1732
$n = 100$	n/a	-1.41 %	57.78 %	1.11 %	25.16 s	1835

## Summary

- SGS for preemptive project scheduling problems with generalized precedence relations and regular min-max criteria
- Create new decision points by iterating threshold instances
- Multi-start heuristic with randomly biased priority indices
- Improvements on benchmark results for non-preemptive problems
  - Preemption gains obtained for 47% of the instances
  - Feasible schedules generated for 17 out of 66 instances that are infeasible when preemption is not allowed

## Future research

- Performance evaluation for other regular min-max criteria
- Metaheuristics embedding preemptive SGS



## References



Franck B, Neumann K, Schwindt C (2001)

Truncated branch-and-bound, schedule-construction, and schedule-improvement procedures for resource-constrained project scheduling  
OR Spektrum 23: 297–324



Hanzálek Z, Šůcha P (2009)

Time Symmetry of Project Scheduling with Time Windows and Take-Give Resources  
In: 4th Multidisciplinary International Scheduling Conference: Theory and Applications, pp 239–253



Schwindt C, Paetz T (2015)

Continuous Preemption Problems

In: Schwindt C, Zimmermann J (eds) Handbook on Project Management and Scheduling Vol. 1. Springer, Cham, Heidelberg, New York, Dordrecht, London, pp 251–295