

# Effective Modeling in Answer Set Programming modulo Theories

Martin Gebser

University of Klagenfurt

Graz University of Technology

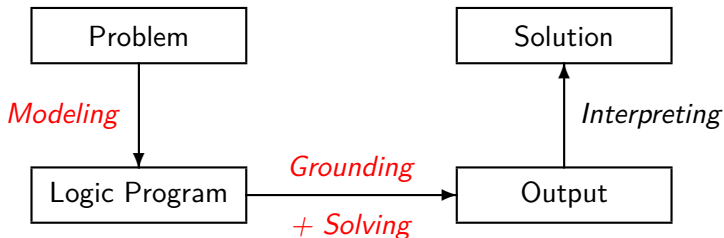


thanks to:

Dirk Abels, Mohammed El-Kholany, Giulia Francescutto, Gerhard Friedrich,  
Clémence Frioux, Christian Haubelt, Tomi Janhunnen, Julian Jordi, Roland Kaminski,  
Kai Neubauer, Max Ostrowski, Konstantin Schekotihin, Sebastian Schellhorn,  
Anne Siegel, Erich Teppan, Ambra Toletti, Torsten Schaub, Philipp Wanko

# Answer Set Programming

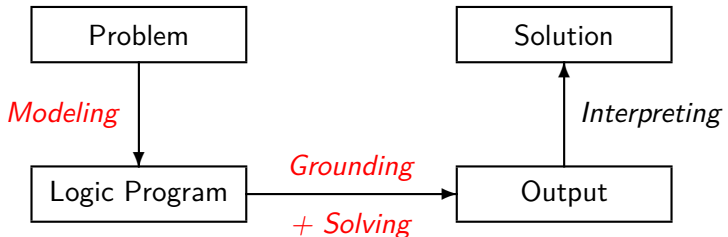
- ▶ Answer Set Programming (ASP) offers expressive first-order modeling language and powerful reasoning technology
  - **Ground instantiation** by semi-naive database evaluation [KS23]
  - **Search/Optimization** by conflict-driven learning [ADMR20, GKS12]



- ▶ Uniform problem representations separate instance data from **high-level problem encoding** for elaboration-tolerant modeling
  - Logic Program =  $\underbrace{\text{Facts}}_{\text{Instance}} + \underbrace{\text{Generate} + \text{Define} + \text{Test} + \text{Optimize}}_{\text{Encoding}}$

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# Answer Set Programming modulo Theories

- ▶ Large domains, even if discrete, lead to **grounding bottleneck**
- ▶ Even linear/logarithmic **propositional representation** by binary or order encoding [CB94, War98] **prohibitive for large ranges**
  - Quantitative resources, spatial coordinates, **time horizons**
- ▶ Extensions like **Constraint Answer Set Programming** (CASP) [Lie23] enable succinct representations of integer/real values
- ▶ Difference Logic (DL) constitutes a **tractable** CASP fragment
  - `clingo-dl` [JKO<sup>+</sup>17] supplies **propagator**, similar to stability [GKS12], acyclicity [BGJ<sup>+</sup>16], or constraint [CDRS20] checking
- 👉 Inspiration by Satisfiability modulo Theories (SMT) [BSST09]

From plain ASP to ASP modulo DL

```
% at(Y) = at(X) + D
at(Y,T+D) :- at(X,T), duration(X,Y,D).
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% at(X) - at(Y) <= -D   iff   at(Y) >= at(X) + D
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# Flexible Job-Shop Scheduling Problem (FJSSP)

- ▶ Classical NP-hard optimization problem [BEP<sup>+</sup>14]
- ▶ Various applications in manufacturing and logistics [XGP<sup>+</sup>19]
- ▶ Benchmark for Answer Set Programming (ASP), Constraint Programming (CP), Satisfiability (SAT) solver competitions

## Problem Specification

- ▶ An instance provides jobs and machines to process operations
  - Jobs = sequences of operations
  - An operation's processing time can vary between machines
- ▶ A schedule assigns a machine and start time to each operation
  - Processing times of operations on same machine do not overlap
  - Successor operations don't start before predecessor completion
- ▶ Minimization of makespan = maximum job completion time



Discrete start times give rise to potentially large integer ranges

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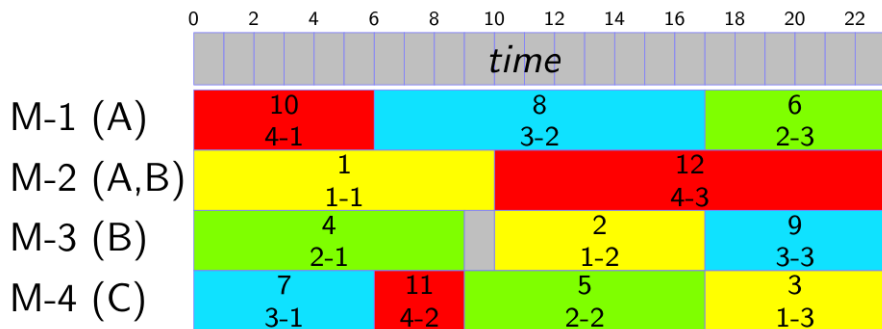


Discrete start times give rise to potentially **large integer ranges**

# FJSSP Instance

OpId	JobId	Pos	Type	M-1	M-2	M-3	M-4
1	1	1	A	11	10		
2	1	2	B		6	7	
3	1	3	C				6
4	2	1	B		8	9	
5	2	2	C				8
6	2	3	A	6	5		
7	3	1	C				6
8	3	2	A	11	10		
9	3	3	B		5	6	
10	4	1	A	6	5		
11	4	2	C				3
12	4	3	B		13	14	

# Optimal FJSSP Solution



**Makespan:** 23 time units

# Instance Representation

```
operation(1,1,1).    mode(1,1,11).    mode(1,2,10).  
operation(2,1,2).    mode(2,2,6).    mode(2,3,7).  
operation(3,1,3).    mode(3,4,6).  
  
operation(4,2,1).    mode(4,2,8).    mode(4,3,9).  
operation(5,2,2).    mode(5,4,8).  
operation(6,2,3).    mode(6,1,6).    mode(6,2,5).  
  
operation(7,3,1).    mode(7,4,6).  
operation(8,3,2).    mode(8,1,11).    mode(8,2,10).  
operation(9,3,3).    mode(9,2,5).    mode(9,3,6).  
  
operation(10,4,1).    mode(10,1,6).    mode(10,2,5).  
operation(11,4,2).    mode(11,4,3).  
operation(12,4,3).    mode(12,2,13).    mode(12,3,14).  
  
time(0..50).
```

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# Vanilla ASP Encoding

% Choose one machine per operation

{process(X,M,P) : mode(X,M,P)} = 1 :- operation(X,J,N).

% Operations (of distinct jobs) on same machine have to be ordered

ordered(X,Y) :- operation(X,J1,N1), process(X,M,P1),  
                  operation(Y,J2,N2), process(Y,M,P2), J1 != J2.

% Choose the operation start times

{start(X,S) : time(S)} = 1 :- operation(X,J,N).

% Derive the operation end times

end(X,S+P) :- start(X,S), process(X,\_,P).

% Operations (of distinct jobs) must be processed without overlaps

:- ordered(X,Y), start(X,S1), start(Y,S2), end(X,E1), S1 <= S2, S2 < E1.

% Successor operation must not start before end of its predecessor

:- operation(X,J,N), operation(Y,J,N+1), start(Y,S), end(X,E), S < E.

% Derive and minimize the makespan

makespan(K) :- K = #max{E : end(X,E),  
  operation(X,J,N), not operation(X+1,J,N+1)}.

:- makespan(K). [K]



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# Let's Run!

```
clingo instance.lp vanilla.lp --stats
```

```
Solving...
```

```
Answer: 1
```

```
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```

```
...
```

```
Answer: 28
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```
process(1,2,10) process(2,2,6) ... start(12,9)
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Optimization: 23
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Time           : 8.981s (Solving: 3.69s ...)
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Variables      : 2099
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```
Constraints    : 1980906
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Highly problematic for ground instantiation size:

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% Choose an order of processing two operations on the same machine

{order(X,Y)} :- ordered(X,Y).  
  order(Y,X) :- ordered(X,Y), not order(X,Y).

% Derive lower bounds on operation start times

start(X,0) :- operation(X,J,1).  
start(Y,E) :- end(X,E), order(X,Y).  
start(Y,E) :- end(X,E), operation(X,J,N), operation(Y,J,N+1).

% Derive lower bounds on operation end times

end(X,S+P) :- start(X,S), process(X,\_,P), time(S).

% Restrict lower bounds on start times to range given by time predicate

:- start(X,S), not time(S).

% Derive processing interval and minimize the makespan

makespan(K) :- end(X,K), operation(X,J,N), not operation(X+1,J,N+1).  
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# Let's Run Again!

```
clingo instance.lp compact.lp --stats & time(0..5000)
```

```
Solving...
```

```
Answer: 1
```

```
process(1,1,11) process(2,2,6) ... start(12,8)
```

```
...
```

```
Answer: 12
```

```
process(1,2,10) process(2,2,6) ... start(12,9)
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```
Optimization: 23
```

```
OPTIMUM FOUND
```

```
Time           : 0.026s (Solving: 0.01s ...)
```

```
Variables      : 3816
```

```
Constraints    : 12713
```



Many time points highly problematic for ground instantiation size

# Let's Run Again!

```
clingo instance.lp compact.lp --stats & time(0..5000)
```

```
Solving...
```

```
Answer: 1
```

```
process(1,2,10) process(2,2,6) ... start(12,8)
```

```
...
```

```
Answer: 5
```

```
process(1,2,10) process(2,2,6) ... start(12,9)
```

```
Optimization: 23
```

```
OPTIMUM FOUND
```

```
Time           : 3.235s (Solving: 1.46s ...)
```

```
Variables      : 483966
```

```
Constraints    : 1685813
```



Many time points highly problematic for ground instantiation size

# ASP modulo DL Encoding

```
% Choose one machine per operation
{process(X,M,P) : mode(X,M,P)} = 1 :- operation(X,J,N).

% Operations (of distinct jobs) on same machine have to be ordered
ordered(X,Y) :- operation(X,J1,N1), process(X,M,P1),
                operation(Y,J2,N2), process(Y,M,P2), J1 < J2.

% Choose an order of processing two operations on the same machine
{order(X,Y)} :- ordered(X,Y).
order(Y,X) :- ordered(X,Y), not order(X,Y).

% Derive lower bounds on operation start times
start(X,0) :- operation(X,J,1).
start(Y,E) :- end(X,E), order(X,Y).
start(Y,E) :- end(X,E), operation(X,J,N), operation(Y,J,N+1).

% Derive lower bounds on operation end times
end(X,S+P) :- start(X,S), process(X,_,P), time(S).

% Restrict lower bounds on start times to range given by time predicate
:- start(X,S), not time(S).

% Derive processing interval and minimize the makespan
makespan(K) :- end(X,K), operation(X,J,N), not operation(X+1,J,N+1).
makespan(K) :- makespan(K+1), 0 < K.
:~ makespan(K). [1,K]
```

# ASP modulo DL Encoding

```
% Choose one machine per operation
{process(X,M,P) : mode(X,M,P)} = 1 :- operation(X,J,N).

% Operations (of distinct jobs) on same machine have to be ordered
ordered(X,Y) :- operation(X,J1,N1), process(X,M,P1),
                operation(Y,J2,N2), process(Y,M,P2), J1 < J2.

% Choose an order of processing two operations on the same machine
{order(X,Y)} :- ordered(X,Y).
order(Y,X) :- ordered(X,Y), not order(X,Y).

% Derive lower bounds on operation start times
start(X) >= 0 :- operation(X,J,1).
start(Y) >= end(X) :- order(X,Y).
start(Y) >= end(X) :- operation(X,J,N), operation(Y,J,N+1).

% Derive lower bounds on operation end times
end(X,S+P) :- start(X,S), process(X,_,P), time(S).

% Restrict lower bounds on start times to range given by time predicate
:- start(X,S), not time(S).

% Derive processing interval and minimize the makespan
makespan(K) :- end(X,K), operation(X,J,N), not operation(X+1,J,N+1).
makespan(K) :- makespan(K+1), 0 < K.
:~ makespan(K). [1,K]
```

# ASP modulo DL Encoding

```
% Choose one machine per operation
{process(X,M,P) : mode(X,M,P)} = 1 :- operation(X,J,N).

% Operations (of distinct jobs) on same machine have to be ordered
ordered(X,Y) :- operation(X,J1,N1), process(X,M,P1),
                  operation(Y,J2,N2), process(Y,M,P2), J1 < J2.

% Choose an order of processing two operations on the same machine
{order(X,Y)} :- ordered(X,Y).
order(Y,X) :- ordered(X,Y), not order(X,Y).

% Derive lower bounds on operation start times
&diff{ 0 - start(X)} <= 0 :- operation(X,J,1).
&diff{end(X) - start(Y)} <= 0 :- order(X,Y).
&diff{end(X) - start(Y)} <= 0 :- operation(X,J,N), operation(Y,J,N+1).

% Derive lower bounds on operation end times
end(X,S+P) :- start(X,S), process(X,_,P), time(S).

% Restrict lower bounds on start times to range given by time predicate
:- start(X,S), not time(S).

% Derive processing interval and minimize the makespan
makespan(K) :- end(X,K), operation(X,J,N), not operation(X+1,J,N+1).
makespan(K) :- makespan(K+1), 0 < K.
:~ makespan(K). [1,K]
```



# ASP modulo DL Encoding

```
% Choose one machine per operation
```

```
{process(X,M,P) : mode(X,M,P)} = 1 :- operation(X,J,N).
```

```
% Operations (of distinct jobs) on same machine have to be ordered
```

```
ordered(X,Y) :- operation(X,J1,N1), process(X,M,P1),  
                  operation(Y,J2,N2), process(Y,M,P2), J1 < J2.
```

```
% Choose an order of processing two operations on the same machine
```

```
{order(X,Y)} :- ordered(X,Y).  
order(Y,X) :- ordered(X,Y), not order(X,Y).
```

```
% Derive lower bounds on operation start times
```

```
&diff{ 0 - start(X)} <= 0 :- operation(X,J,1).  
&diff{end(X) - start(Y)} <= 0 :- order(X,Y).  
&diff{end(X) - start(Y)} <= 0 :- operation(X,J,N), operation(Y,J,N+1).
```

```
% Derive lower bounds on operation end times
```

```
end(X,S+P) :- start(X,S), process(X,_,P), time(S).
```

```
% Restrict lower bounds on start times to range given by time predicate
```

```
:- start(X,S), not time(S).
```

```
% Derive processing interval and minimize the makespan
```

```
makespan(K) :- end(X,K), operation(X,J,N), not operation(X+1,J,N+1).
```

```
makespan(K) :- makespan(K+1), 0 < K.
```

```
:~ makespan(K). [1,K]
```

# ASP modulo DL Encoding

```
% Choose one machine per operation
{process(X,M,P) : mode(X,M,P)} = 1 :- operation(X,J,N).

% Operations (of distinct jobs) on same machine have to be ordered
ordered(X,Y) :- operation(X,J1,N1), process(X,M,P1),
                  operation(Y,J2,N2), process(Y,M,P2), J1 < J2.

% Choose an order of processing two operations on the same machine
{order(X,Y)} :- ordered(X,Y).
order(Y,X) :- ordered(X,Y), not order(X,Y).

% Derive lower bounds on operation start times
&diff{ 0 - start(X)} <= 0 :- operation(X,J,1).
&diff{end(X) - start(Y)} <= 0 :- order(X,Y).
&diff{end(X) - start(Y)} <= 0 :- operation(X,J,N), operation(Y,J,N+1).

% Derive lower bounds on operation end times
end(X) >= start(X) + P :- process(X,_,P).

% Restrict lower bounds on start times to range given by time predicate
:- start(X,S), not time(S).

% Derive processing interval and minimize the makespan
makespan(K) :- end(X,K), operation(X,J,N), not operation(X+1,J,N+1).
makespan(K) :- makespan(K+1), 0 < K.
:~ makespan(K). [1,K]
```

# ASP modulo DL Encoding

```
% Choose one machine per operation
{process(X,M,P) : mode(X,M,P)} = 1 :- operation(X,J,N).

% Operations (of distinct jobs) on same machine have to be ordered
ordered(X,Y) :- operation(X,J1,N1), process(X,M,P1),
                 operation(Y,J2,N2), process(Y,M,P2), J1 < J2.

% Choose an order of processing two operations on the same machine
{order(X,Y)} :- ordered(X,Y).
order(Y,X) :- ordered(X,Y), not order(X,Y).

% Derive lower bounds on operation start times
&diff{ 0 - start(X)} <= 0 :- operation(X,J,1).
&diff{end(X) - start(Y)} <= 0 :- order(X,Y).
&diff{end(X) - start(Y)} <= 0 :- operation(X,J,N), operation(Y,J,N+1).

% Derive lower bounds on operation end times
&diff{start(X) - end(X)} <= -P :- process(X,_,P).

% Restrict lower bounds on start times to range given by time predicate
:- start(X,S), not time(S).

% Derive processing interval and minimize the makespan
makespan(K) :- end(X,K), operation(X,J,N), not operation(X+1,J,N+1).
makespan(K) :- makespan(K+1), 0 < K.
:~ makespan(K). [1,K]
```

# ASP modulo DL Encoding

```
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{process(X,M,P) : mode(X,M,P)} = 1 :- operation(X,J,N).

% Operations (of distinct jobs) on same machine have to be ordered
ordered(X,Y) :- operation(X,J1,N1), process(X,M,P1),
                  operation(Y,J2,N2), process(Y,M,P2), J1 < J2.

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{order(X,Y)} :- ordered(X,Y).
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&diff{ 0 - start(X)} <= 0 :- operation(X,J,1).
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&diff{end(X) - start(Y)} <= 0 :- operation(X,J,N), operation(Y,J,N+1).

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:~ makespan(K). [1,K]
```

# ASP modulo DL Encoding

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{process(X,M,P) : mode(X,M,P)} = 1 :- operation(X,J,N).

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ordered(X,Y) :- operation(X,J1,N1), process(X,M,P1),
                  operation(Y,J2,N2), process(Y,M,P2), J1 < J2.

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order(Y,X) :- ordered(X,Y), not order(X,Y).

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&diff{end(X) - start(Y)} <= 0 :- operation(X,J,N), operation(Y,J,N+1).

% Derive lower bounds on operation end times
&diff{start(X) - end(X)} <= -P :- process(X,_,P).

% Restrict lower bounds on start times to range given by time predicate
start(X) <= S :- operation(X,J,N), not operation(X+1,J,N+1),
                  time(S), not time(S+1).

% Derive processing interval and minimize the makespan
makespan(K) :- end(X,K), operation(X,J,N), not operation(X+1,J,N+1).
makespan(K) :- makespan(K+1), 0 < K.
```

# ASP modulo DL Encoding

```
% Choose one machine per operation
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&diff{end(X) - start(Y)} <= 0 :- operation(X,J,N), operation(Y,J,N+1).

% Derive lower bounds on operation end times
&diff{start(X) - end(X)} <= -P :- process(X,_,P).

% Restrict lower bounds on start times to range given by time predicate
&diff{start(X) - 0} <= S :- operation(X,J,N), not operation(X+1,J,N+1),
                             time(S), not time(S+1).

% Derive processing interval and minimize the makespan
makespan(K) :- end(X,K), operation(X,J,N), not operation(X+1,J,N+1).
makespan(K) :- makespan(K+1), 0 < K.
```

# ASP modulo DL Encoding

```
% Choose one machine per operation
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&diff{start(X) - end(X)} <= -P :- process(X,_,P).

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&diff{start(X) - 0} <= S :- operation(X,J,N), not operation(X+1,J,N+1),
                             time(S), not time(S+1).

% Derive processing interval and minimize the makespan
makespan(K) :- end(X,K), operation(X,J,N), not operation(X+1,J,N+1).
makespan(K) :- makespan(K+1), 0 < K.
```

# ASP modulo DL Encoding

```
% Choose one machine per operation
{process(X,M,P) : mode(X,M,P)} = 1 :- operation(X,J,N).

% Operations (of distinct jobs) on same machine have to be ordered
ordered(X,Y) :- operation(X,J1,N1), process(X,M,P1),
                  operation(Y,J2,N2), process(Y,M,P2), J1 < J2.

% Choose an order of processing two operations on the same machine
{order(X,Y)} :- ordered(X,Y).
order(Y,X) :- ordered(X,Y), not order(X,Y).

% Derive lower bounds on operation start times
&diff{ 0 - start(X)} <= 0 :- operation(X,J,1).
&diff{end(X) - start(Y)} <= 0 :- order(X,Y).
&diff{end(X) - start(Y)} <= 0 :- operation(X,J,N), operation(Y,J,N+1).

% Derive lower bounds on operation end times
&diff{start(X) - end(X)} <= -P :- process(X,_,P).

% Restrict lower bounds on start times to range given by time predicate
&diff{start(X) - 0} <= S :- operation(X,J,N), not operation(X+1,J,N+1),
                             time(S), not time(S+1).

% Derive the makespan
makespan >= end(X) :- operation(X,J,N), not operation(X+1,J,N+1).
```



# ASP modulo DL Encoding

```
% Choose one machine per operation
{process(X,M,P) : mode(X,M,P)} = 1 :- operation(X,J,N).

% Operations (of distinct jobs) on same machine have to be ordered
ordered(X,Y) :- operation(X,J1,N1), process(X,M,P1),
                  operation(Y,J2,N2), process(Y,M,P2), J1 < J2.

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order(Y,X) :- ordered(X,Y), not order(X,Y).

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&diff{ 0 - start(X)} <= 0 :- operation(X,J,1).
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&diff{end(X) - start(Y)} <= 0 :- operation(X,J,N), operation(Y,J,N+1).

% Derive lower bounds on operation end times
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% Restrict lower bounds on start times to range given by time predicate
&diff{start(X) - 0} <= S :- operation(X,J,N), not operation(X+1,J,N+1),
                           time(S), not time(S+1).

% Derive the makespan
&diff{end(X) - makespan} <= 0 :- operation(X,J,N), not operation(X+1,J,N+1).
```

# Let's Run clingo-dl!

```
clingo-dl instance.lp difference.lp --stats  
--minimize-variable=makespan & time(0..5000)
```

Solving...

Answer: 1

process(1,1,11) process(2,3,7) ... dl(start(12),9)

...

Answer: 11

process(1,2,10) process(2,3,7) ... dl(start(12),10)

DL Optimization: 23

UNSATISFIABLE

Time : 0.013s (Solving: 0.00s ...)

Variables : 141

Constraints : 256



Time range makes no difference for ground instantiation size

# Let's Run clingo-dl!

```
clingo-dl instance.lp difference.lp --stats  
--minimize-variable=makespan & time(0..5000)
```

```
Solving...
```

```
Answer: 1
```

```
process(1,1,11) process(2,3,7) ... dl(start(12),9)
```

```
...
```

```
Answer: 10
```

```
process(1,2,10) process(2,3,7) ... dl(start(12),10)
```

```
DL Optimization: 23
```

```
UNSATISFIABLE
```

```
Time          : 0.031s (Solving: 0.00s ...)
```

```
Variables     : 140
```

```
Constraints   : 199
```



Time range makes no difference for ground instantiation size

# And The Winner is ...

- ▶ Time points don't increase ground instantiation size with DL
- ▶ Makespan minimization refers to a **single DL variable's value**



Perfect for: `clingo-dl --minimize-variable=makespan`

- ▶ Turning from makespan minimization

```
makespan(K) :- end(X,K), operation(X,J,N), not operation(X+1,J,N+1).  
makespan(K) :- makespan(K+1), 0 < K.  
:~ makespan(K). [1,K]
```

- ▶ Native support of (lexicographic) multi-criteria optimization:

```
makespan(K) :- end(X,K), operation(X,J,N), not operation(X+1,J,N+1).  
makespan(K) :- makespan(K+1), 0 < K.  
:~ makespan(K). [1@2,K]
```

- ▶ The optimization features provided for DL variables are not yet on a par with functionalities readily available in plain ASP



Research and implementation progress needed to **take burden of scripting custom DL optimization algorithms from the user**

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# And The Winner is ...

- ▶ Time points don't increase ground instantiation size with DL
- ▶ Makespan minimization refers to a single DL variable's value



Perfect for: `clingo-dl --minimize-variable=makespan`

- ▶ Turning from makespan to **total processing time** minimization

`:~ end(X,K), operation(X,J,N), not operation(X+1,J,N+1). [K,J]`

is easy to do in plain ASP

- ▶ Native support of (lexicographic) multi-criteria optimization:

`makespan(K) :- end(X,K), operation(X,J,N), not operation(X+1,J,N+1).`

`makespan(K) :- makespan(K+1), 0 < K.`

`:~ makespan(K). [1@2,K]`

- ▶ The optimization features provided for DL variables are not yet on a par with functionalities readily available in plain ASP



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# And The Winner is ...

- ▶ Time points don't increase ground instantiation size with DL
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Perfect for: `clingo-dl --minimize-variable=makespan`

- ▶ Turning from makespan to total processing time minimization

`:~ end(X,K), operation(X,J,N), not operation(X+1,J,N+1). [K@1,J]`

is easy to do in plain ASP

- ▶ Native support of (lexicographic) **multi-criteria optimization**:

`makespan(K) :- end(X,K), operation(X,J,N), not operation(X+1,J,N+1).`

`makespan(K) :- makespan(K+1), 0 < K.`

`:~ makespan(K). [1@2,K]`

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Perfect for: `clingo-dl --minimize-variable=makespan`

- ▶ Turning from makespan to total processing time minimization

`:~ end(X,K), operation(X,J,N), not operation(X+1,J,N+1). [K@1,J]`

is easy to do in plain ASP

- ▶ Native support of (lexicographic) multi-criteria optimization:

`makespan(K) :- end(X,K), operation(X,J,N), not operation(X+1,J,N+1).`

`makespan(K) :- makespan(K+1), 0 < K.`

`:~ makespan(K). [1@2,K]`

- ▶ The optimization features provided for DL variables are not yet on a par with functionalities readily available in plain ASP

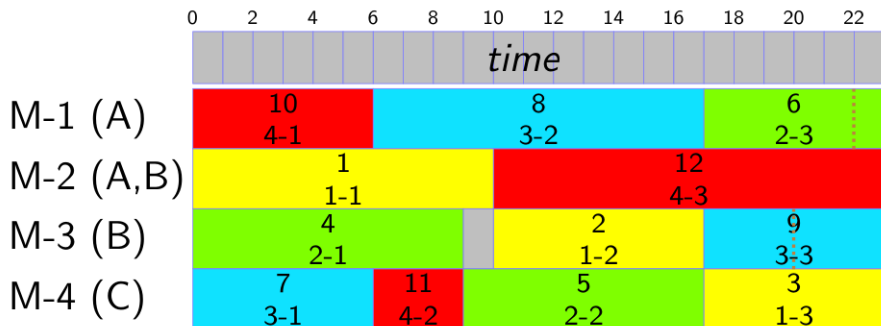


Research and implementation progress needed to **take burden of scripting custom DL optimization algorithms from the user**



# Let's Add Job Dealines

deadline(1,23).      deadline(2,22).  
deadline(3,21).      deadline(4,23).



Job 2 is **delayed** by one and Job 3 by two time units

# Minimize Number of Delayed Jobs

```
% Choose one machine per operation
{process(X,M,P) : mode(X,M,P)} = 1 :- operation(X,J,N).

% Operations (of distinct jobs) on same machine have to be ordered
ordered(X,Y) :- operation(X,J1,N1), process(X,M,P1),
                  operation(Y,J2,N2), process(Y,M,P2), J1 < J2.

% Choose an order of processing two operations on the same machine
{order(X,Y)} :- ordered(X,Y).
order(Y,X) :- ordered(X,Y), not order(X,Y).

% Derive lower bounds on operation start times
&diff{ 0 - start(X)} <= 0 :- operation(X,J,1).
&diff{end(X) - start(Y)} <= 0 :- order(X,Y).
&diff{end(X) - start(Y)} <= 0 :- operation(X,J,N), operation(Y,J,N+1).

% Derive lower bounds on operation end times
&diff{start(X) - end(X)} <= -P :- process(X,_,P).

% Restrict lower bounds on start times to range given by time predicate
&diff{start(X) - 0} <= S :- operation(X,J,N), not operation(X+1,J,N+1),
                             time(S), not time(S+1).

% Derive the makespan
&diff{end(X) - makespan} <= 0 :- operation(X,J,N), not operation(X+1,J,N+1).
```

# Minimize Number of Delayed Jobs

```
% Choose one machine per operation
{process(X,M,P) : mode(X,M,P)} = 1 :- operation(X,J,N).

% Operations (of distinct jobs) on same machine have to be ordered
ordered(X,Y) :- operation(X,J1,N1), process(X,M,P1),
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{order(X,Y)} :- ordered(X,Y).
order(Y,X) :- ordered(X,Y), not order(X,Y).

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&diff{ 0 - start(X)} <= 0 :- operation(X,J,1).
&diff{end(X) - start(Y)} <= 0 :- order(X,Y).
&diff{end(X) - start(Y)} <= 0 :- operation(X,J,N), operation(Y,J,N+1).

% Derive lower bounds on operation end times
&diff{start(X) - end(X)} <= -P :- process(X,_,P).

% Restrict lower bounds on start times to range given by time predicate
&diff{start(X) - 0} <= S :- operation(X,J,N), not operation(X+1,J,N+1),
                             time(S), not time(S+1).

% Minimize number of delayed jobs
:~ operation(X,J,N), not operation(X+1,J,N+1),
  deadline(J,D), not end(X) <= D. [1,J]
```

# Minimize Number of Delayed Jobs

```
% Choose one machine per operation
{process(X,M,P) : mode(X,M,P)} = 1 :- operation(X,J,N).

% Operations (of distinct jobs) on same machine have to be ordered
ordered(X,Y) :- operation(X,J1,N1), process(X,M,P1),
                  operation(Y,J2,N2), process(Y,M,P2), J1 < J2.

% Choose an order of processing two operations on the same machine
{order(X,Y)} :- ordered(X,Y).
order(Y,X) :- ordered(X,Y), not order(X,Y).

% Derive lower bounds on operation start times
&diff{ 0 - start(X)} <= 0 :- operation(X,J,1).
&diff{end(X) - start(Y)} <= 0 :- order(X,Y).
&diff{end(X) - start(Y)} <= 0 :- operation(X,J,N), operation(Y,J,N+1).

% Derive lower bounds on operation end times
&diff{start(X) - end(X)} <= -P :- process(X,_,P).

% Restrict lower bounds on start times to range given by time predicate
&diff{start(X) - 0} <= S :- operation(X,J,N), not operation(X+1,J,N+1),
                             time(S), not time(S+1).

% Minimize number of delayed jobs
:~ operation(X,J,N), not operation(X+1,J,N+1),
   deadline(J,D), not end(X) <= D. [1,J]
```

# Minimize Number of Delayed Jobs

```
% Choose one machine per operation
{process(X,M,P) : mode(X,M,P)} = 1 :- operation(X,J,N).

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&diff{start(X) - 0} <= S :- operation(X,J,N), not operation(X+1,J,N+1),
                             time(S), not time(S+1).

% Minimize number of delayed jobs
:~ operation(X,J,N), not operation(X+1,J,N+1),
    deadline(J,D), not &diff{end(X) - 0} <= D. [1,J]
```

# Let's Run with Delayed Job Minimization!

```
clingo-dl instance.lp deadline.lp differences.lp  
--stats & time(0..1000)
```

Solving...

Answer: 1

process(1,1,11) process(2,3,7) ... dl(start(12),9)

...

Answer: 4

process(1,1,11) process(2,2,6) ... dl(start(12),10)

Optimization: 1

OPTIMUM FOUND

Time : 0.002s (Solving: 0.00s ...)

Conflicts : 37

Constraints : 256



Time range makes no difference for optimization performance

# Let's Run with Delayed Job Minimization!

```
clingo-dl instance.lp deadline.lp differences.lp  
--stats & time(0..1000)
```

```
Solving...
```

```
Answer: 1
```

```
process(1,1,11) process(2,3,7) ... dl(start(12),9)
```

```
...
```

```
Answer: 4
```

```
process(1,1,11) process(2,2,6) ... dl(start(12),10)
```

```
Optimization: 1
```

```
OPTIMUM FOUND
```

```
Time           : 0.005s (Solving: 0.00s ...)
```

```
Conflicts      : 43
```

```
Constraints    : 256
```



Time range makes no difference for optimization performance

# Minimize Sum of Job Delays

[...]

% Choose an order of processing two operations on the same machine

{order(X,Y)} :- ordered(X,Y).

order(Y,X) :- ordered(X,Y), not order(X,Y).

% Derive lower bounds on operation start times

&diff{ 0 - start(X)} <= 0 :- operation(X,J,1).

&diff{end(X) - start(Y)} <= 0 :- order(X,Y).

&diff{end(X) - start(Y)} <= 0 :- operation(X,J,N), operation(Y,J,N+1).

% Derive lower bounds on operation end times

&diff{start(X) - end(X)} <= -P :- process(X,\_,P).

% Restrict lower bounds on start times to range given by time predicate

&diff{start(X) - 0} <= S :- operation(X,J,N), not operation(X+1,J,N+1),  
time(S), not time(S+1).

% Minimize number of delayed jobs

:~ operation(X,J,N), not operation(X+1,J,N+1),  
deadline(J,D), not &diff{end(X) - 0} <= D. [1,J]



# Minimize Sum of Job Delays

[...]

```
% Choose an order of processing two operations on the same machine
{order(X,Y)} :- ordered(X,Y).
order(Y,X) :- ordered(X,Y), not order(X,Y).

% Derive lower bounds on operation start times
&diff{ 0 - start(X)} <= 0 :- operation(X,J,1).
&diff{end(X) - start(Y)} <= 0 :- order(X,Y).
&diff{end(X) - start(Y)} <= 0 :- operation(X,J,N), operation(Y,J,N+1).

% Derive lower bounds on operation end times
&diff{start(X) - end(X)} <= -P :- process(X,_,P).

% Restrict lower bounds on start times to range given by time predicate
&diff{start(X) - 0} <= S :- operation(X,J,N), not operation(X+1,J,N+1),
time(S), not time(S+1).

% Minimize sum of job delays
delay(X,D,0..S+P-D-1) :- operation(X,J,N), not operation(X+1,J,N+1),
mode(X,M,P), deadline(J,D),
time(S), not time(S+1).

:~ delay(X,D,T), not end(X) <= D+T. [1,X,T]
```

# Minimize Sum of Job Delays

[...]

```
% Choose an order of processing two operations on the same machine
{order(X,Y)} :- ordered(X,Y).
order(Y,X) :- ordered(X,Y), not order(X,Y).

% Derive lower bounds on operation start times
&diff{ 0 - start(X)} <= 0 :- operation(X,J,1).
&diff{end(X) - start(Y)} <= 0 :- order(X,Y).
&diff{end(X) - start(Y)} <= 0 :- operation(X,J,N), operation(Y,J,N+1).

% Derive lower bounds on operation end times
&diff{start(X) - end(X)} <= -P :- process(X,_,P).

% Restrict lower bounds on start times to range given by time predicate
&diff{start(X) - 0} <= S :- operation(X,J,N), not operation(X+1,J,N+1),
time(S), not time(S+1).

% Minimize sum of job delays
delay(X,D,0..S+P-D-1) :- operation(X,J,N), not operation(X+1,J,N+1),
mode(X,M,P), deadline(J,D),
time(S), not time(S+1).

:~ delay(X,D,T), not end(X) <= D+T. [1,X,T]
```

# Minimize Sum of Job Delays

[...]

```
% Choose an order of processing two operations on the same machine
{order(X,Y)} :- ordered(X,Y).
order(Y,X) :- ordered(X,Y), not order(X,Y).

% Derive lower bounds on operation start times
&diff{ 0 - start(X)} <= 0 :- operation(X,J,1).
&diff{end(X) - start(Y)} <= 0 :- order(X,Y).
&diff{end(X) - start(Y)} <= 0 :- operation(X,J,N), operation(Y,J,N+1).

% Derive lower bounds on operation end times
&diff{start(X) - end(X)} <= -P :- process(X,_,P).

% Restrict lower bounds on start times to range given by time predicate
&diff{start(X) - 0} <= S :- operation(X,J,N), not operation(X+1,J,N+1),
time(S), not time(S+1).

% Minimize sum of job delays
delay(X,D,0..S+P-D-1) :- operation(X,J,N), not operation(X+1,J,N+1),
mode(X,M,P), deadline(J,D),
time(S), not time(S+1).

:~ delay(X,D,T), not &diff{end(X) - 0} <= D+T. [1,X,T]
```

# Let's Run with Delay Sum Minimization!

```
clingo-dl instance.lp deadline.lp differencet.lp  
--stats & time(0..1000)
```

Solving...

Answer: 1

process(1,1,11) process(2,3,7) ... dl(start(12),9)

...

Answer: 54

process(1,2,10) process(2,2,6) ... dl(start(12),9)

Optimization: 2

OPTIMUM FOUND

Time : 0.012s (Solving: 0.01s ...)

Conflicts : 699

Constraints : 256



Time range may deteriorate optimization performance

# Let's Run with Delay Sum Minimization!

```
clingo-dl instance.lp deadline.lp differencet.lp  
--stats & time(0..1000)
```

Solving...

Answer: 1

process(1,1,11) process(2,3,7) ... dl(start(12),9)

...

Answer: 3842

process(1,2,10) process(2,2,6) ... dl(start(12),9)

Optimization: 2

OPTIMUM FOUND

Time : 345.868s (Solving: 345.84s ...)

Conflicts : 937922

Constraints : 256



Time range may deteriorate optimization performance

# Let's Run with Delay Sum Minimization!

```
clingo-dl instance.lp deadline.lp differencet.lp  
--stats --opt-heuristic=1 & time(0..1000)
```

```
Solving...
```

```
Answer: 1
```

```
process(1,2,10) process(2,2,6) ... dl(start(12),25)
```

```
...
```

```
Answer: 59
```

```
process(1,2,10) process(2,2,6) ... dl(start(12),9)
```

```
Optimization: 2
```

```
OPTIMUM FOUND
```

```
Time           : 0.198s (Solving: 0.17s ...)
```

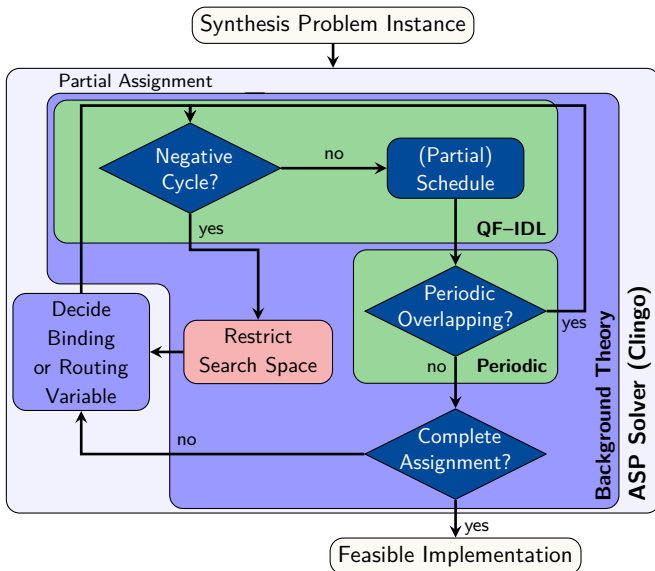
```
Conflicts      : 667
```

```
Constraints    : 256
```



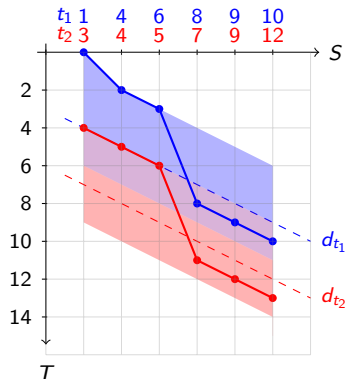
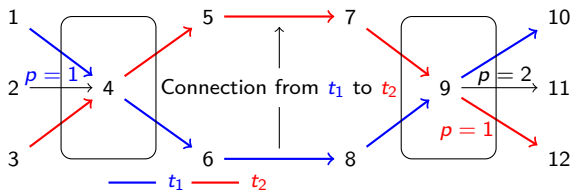
Time range may deteriorate optimization performance

# Symbolic System Synthesis [NWSH17]



# Train Scheduling [AJO<sup>+</sup>19]

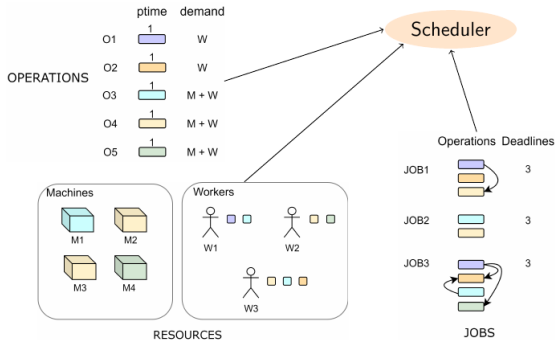
- Minimize **sum** of route penalties  $p$  and delays  $d$  at train stations



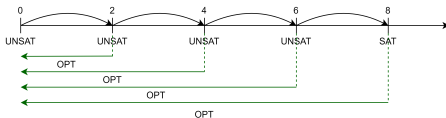


# Fault Analysis Lab Scheduling [FSE21]

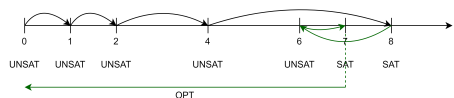
- Minimize **sum** of jobs' tardiness **w.r.t. feasible maximum** tardiness



- Linear search scheme:



- Exponential search scheme:



# Conclusion

- ▶ ASP modulo DL supplies **compact representation of quantities**
- ▶ Particularly useful to express **time, delays and non-overlapping**
- ▶ Optimization by `clingo-dl` limited to **single DL variable value**
- ▶ **Summation** and/or **lexicographic optimization** need
  - Dedicated propagators [NWSH18] or
  - Reification in plain/multi-shot ASP [AJO<sup>+</sup>19, FSE21]
- 👉 `clingo-dl` is great for solving **simplistic** ASP modulo DL **decision and optimization problems**
- 👉 **Richer optimization functionalities** on DL variables' values would be the cherry on the cake

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