



Declaring Local-Search Neighbourhoods in MiniZinc

Gustav Björdal¹

Pierre Flener,¹ Justin Pearson,¹ Peter J. Stuckey,² and Guido Tack³

¹Uppsala University, ²University of Melbourne, ³Monash University

Goal



- Extend the MiniZinc language with declarative neighbourhoods.
- Allow rapid prototyping with local search strategies.
- Recreate known local search strategies in MiniZinc.

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This presentation

Some background and a short walkthrough of our new MiniZinc syntax.

MiniZinc

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Constraint-based declarative modelling language.

Solver and technology independent: more than 15 backends.

High-level syntax.

Free and open-source under MPL 2.0.

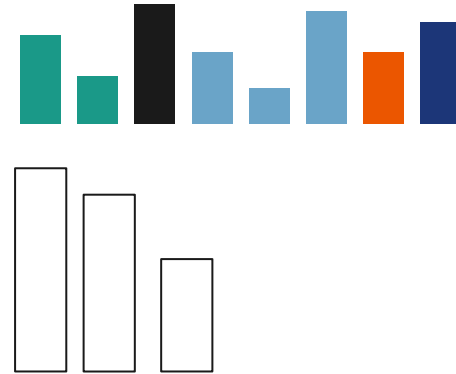
Annual MiniZinc Challenge.

www.minizinc.org

Running Example – Steel Mill Slab Design

Given

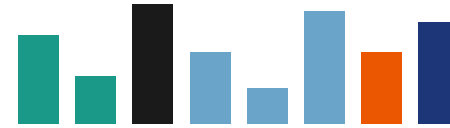
- orders of steel, each having a *size* and *colour*, to be cut from slabs
- available slab sizes



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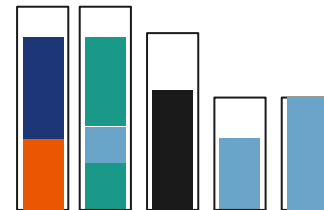
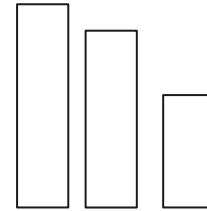
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Decide

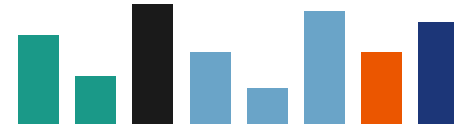
- how many slabs to use
- the size of each slab
- the slab each order is placed in



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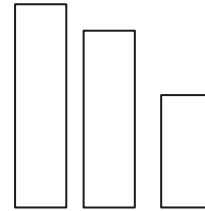
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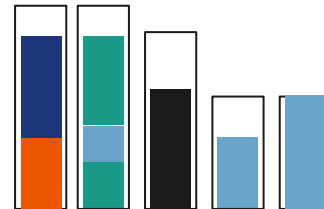
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Such that

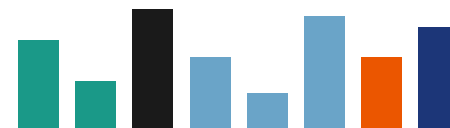
- at most two colours on each slab
- the total slack in the slabs is minimal



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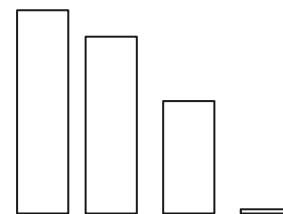
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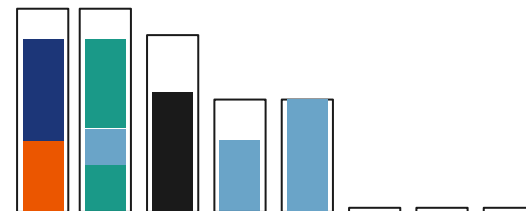
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- ~~— how many slabs to use~~
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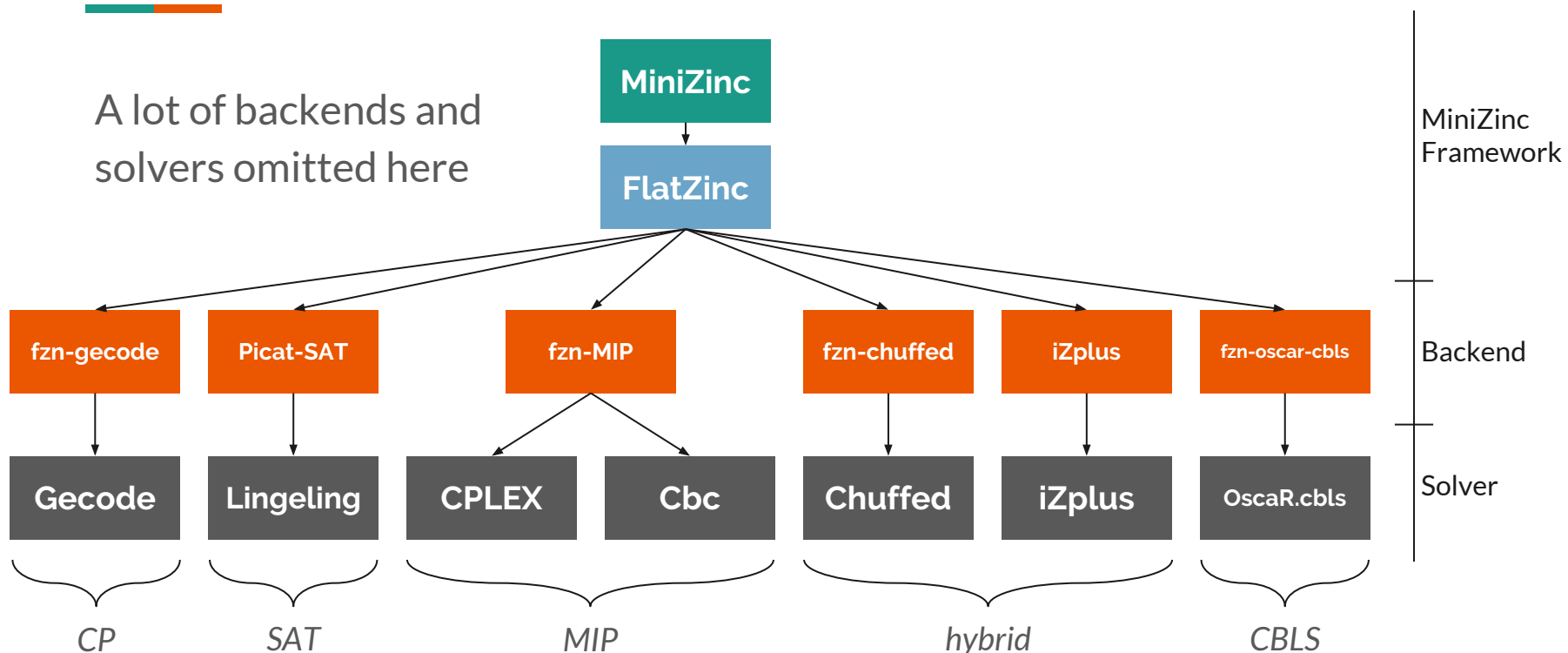


MiniZinc Model

```
% Some parameters omitted
...
array [Orders] of int: size;
array [Orders] of Colors: color;
array [0..maxCapa] of 0..maxCapa: slack = ... ;
% Variables:
array [Orders] of var Slabs: placedIn;
% Constraints:
array [Slabs] of var 0..maxCapa: load;
constraint bin_packing_load(load, placedIn, size);
array [Slabs] of var 0..2: nColors;
constraint forall(s in Slabs)(nColors[s] = ... );
% Objective:
var int: objective = sum(s in Slabs)(slack[load[s]]);
solve minimize objective;
```

Current Landscape

A lot of backends and solvers omitted here



(Constraint-Based) Local Search

Local Search



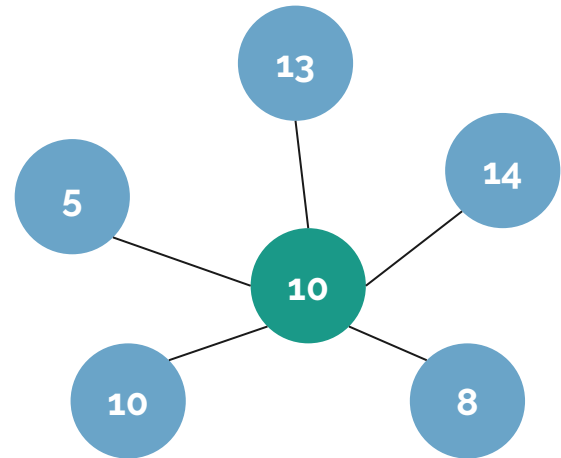
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2. While some condition holds:
 - a. Generate a neighbourhood of similar assignments
 - b. Move to a best neighbour
3. Return the best solution found

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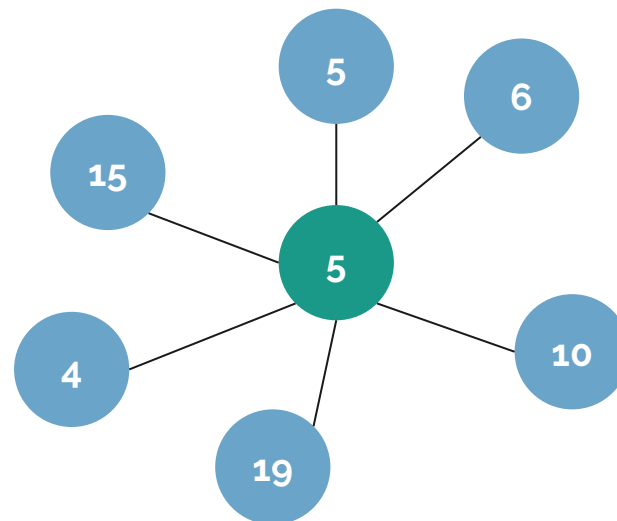
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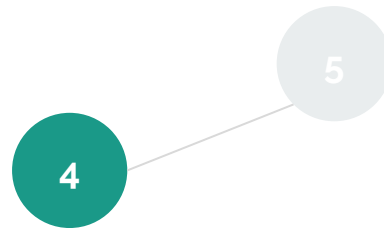
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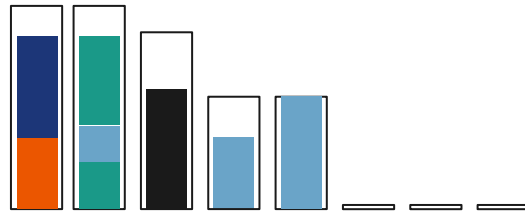


Local Search



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Steel Mill Slab Design

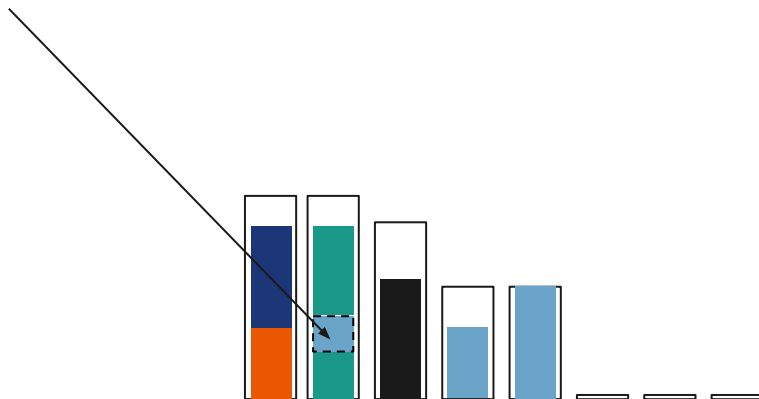


An initial assignment

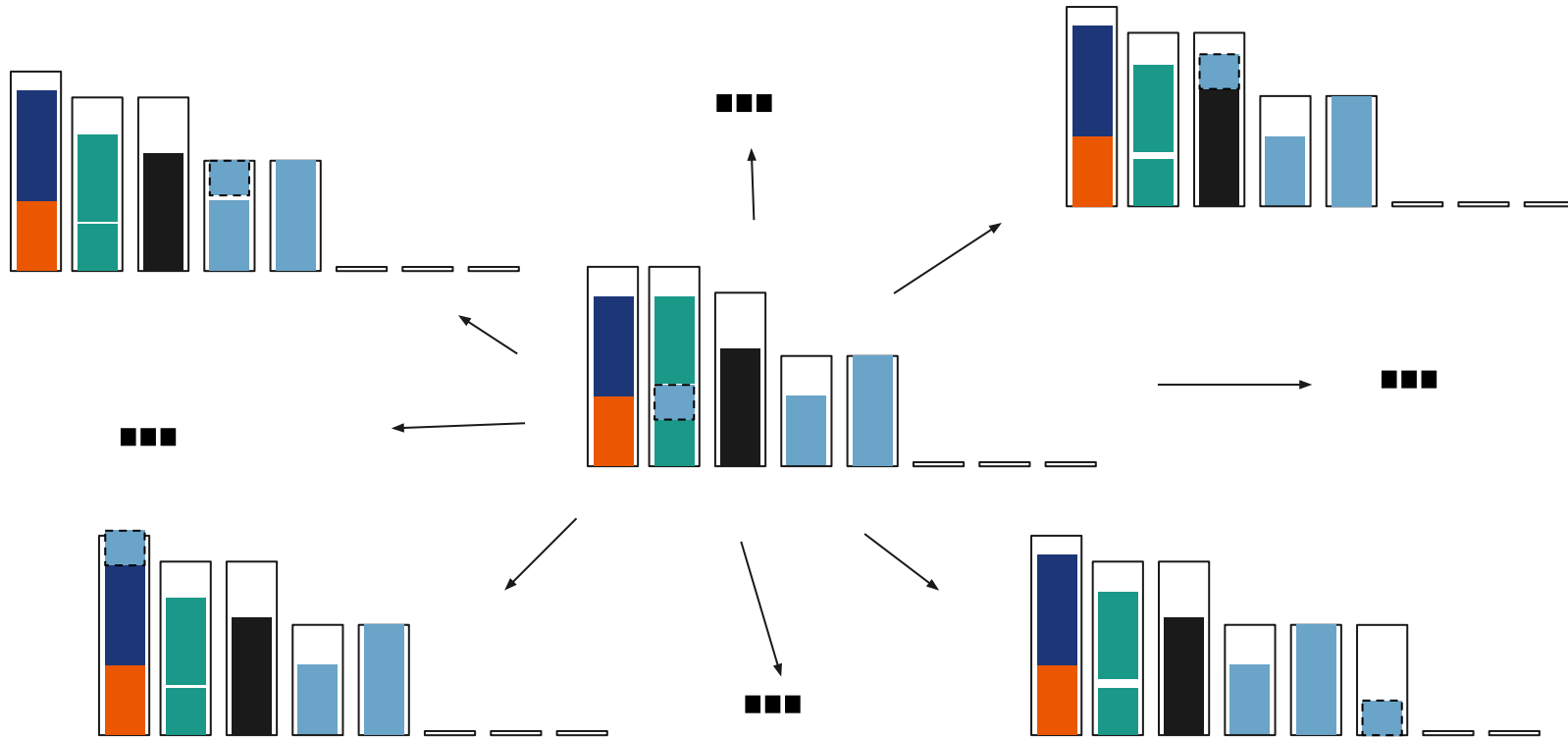
Steel Mill Slab Design



Move an order to another slab.



Steel Mill Slab Design



Design Aspects



Initialisation and neighbourhood

- How do we initialise?
- What are the moves?

Heuristic

- How do we explore the neighbourhood?

Meta-heuristic

- How do we prevent the search from getting stuck?

Constraint-Based Local Search (CBLS)



CBLS = CP-style declarative modelling + local search

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CBLS for MiniZinc

MiniZinc is a modelling language: must use black-box local search.

Works well but can be hit and miss.

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CBLS for MiniZinc

MiniZinc is a modelling language: must use black-box local search.

Works well but can be hit and miss.

Idea

Allow modellers to define (part of) a local search strategy in MiniZinc.

Declarative Neighbourhoods in MiniZinc

What is a Neighbourhood?



A neighbourhood is a **set of moves**.

For example:

$\{X \leftarrow v \mid v \in 1..10\}$ i.e., $\{X \leftarrow 1, X \leftarrow 2, \dots, X \leftarrow 10\}$

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- which move to select

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We provide syntax only for **defining the neighbourhood**, not heuristics.

Basic Syntax

Move operators

$X := v$

$X[i] := v$

$X ::= Y$

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`X[i] := v`

`X :=: Y`

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`moves(v in 1..10)(X := v)`

Compound moves

`moves(i, j in Idx)(Xs[i] := Xs[j] /\ Xs[j] := Xs[i])`

Pre- and Post-Conditions



Moves can have pre- and post-conditions in the form of CSPs.

Pre-condition

```
moves(i in Idx, v in Dom where Xs[i] > v)(Xs[i] := v)
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Post-condition

```
moves(i in Idx, v in Dom)(  
  Xs[i] := v /\ ensuring(alldifferent(Xs))
```

Initialisation

A neighbourhood can have an initialisation post-condition that must hold upon the initial assignment (and restarts).

Permutation neighbourhood

```
initially(alldifferent(Xs)) /\  
moves(i, j in Idx where i < j)(Xs[i] := Xs[j])
```

Union of Neighbourhoods

The union of two neighbourhoods can be expressed.

Swaps and assigns

```
moves (i in Idx, v in Dom)(Xs[i] := v)
```

```
union
```

```
moves (i, j in Idx where i < j)(Xs[i] :=: Xs[j])
```

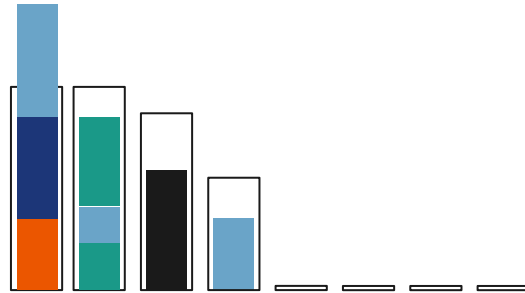
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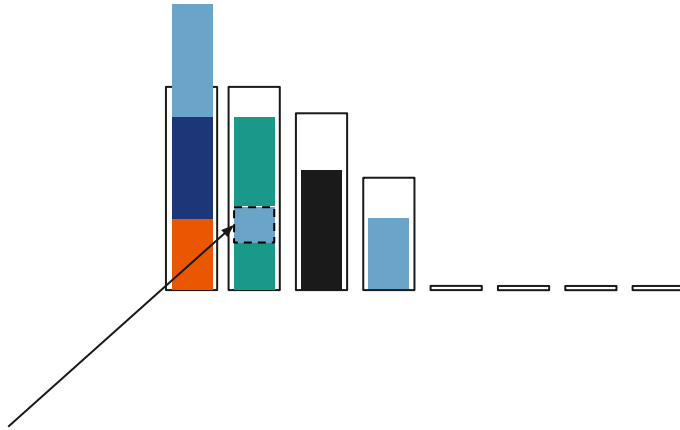
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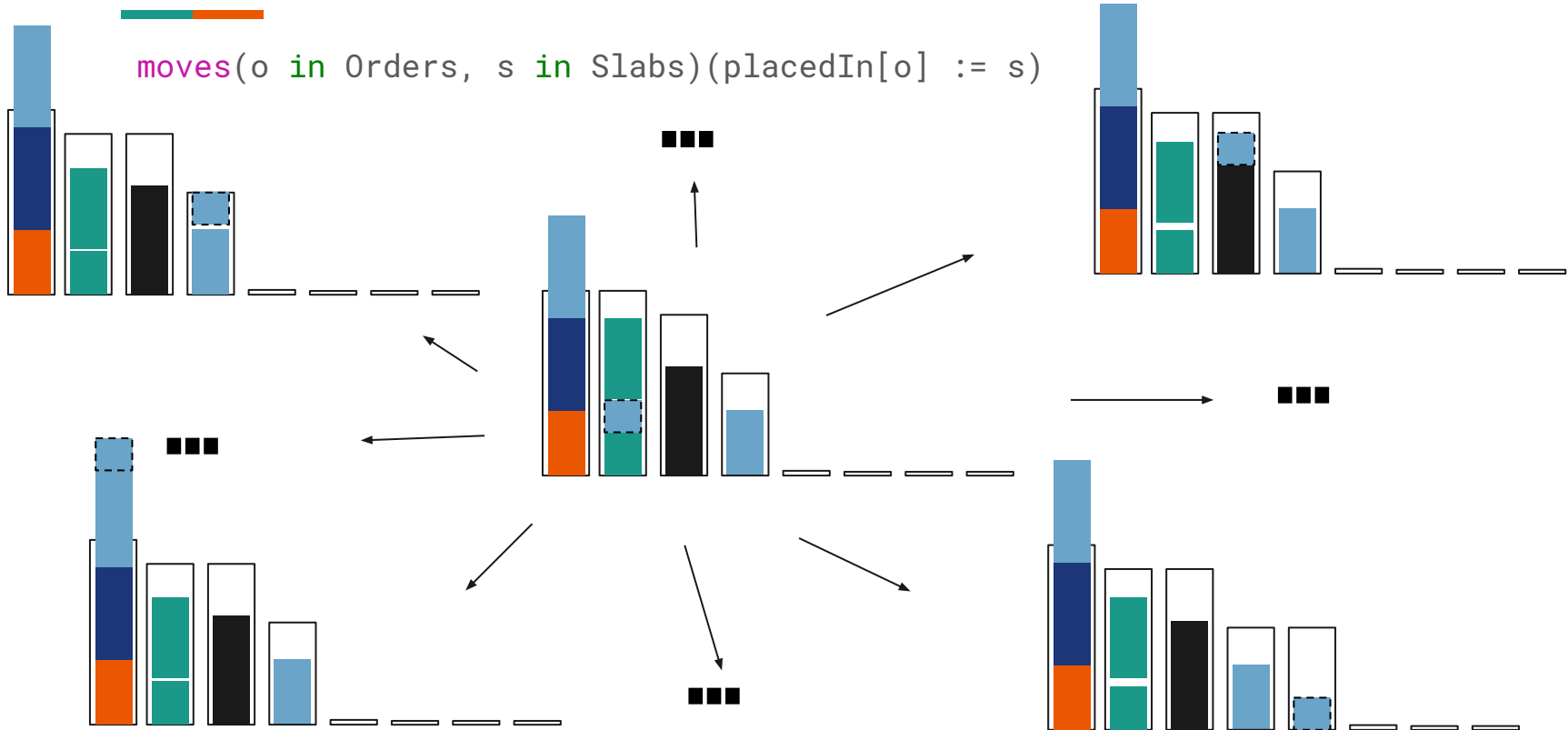
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Example 2 – Only Feasible Solutions

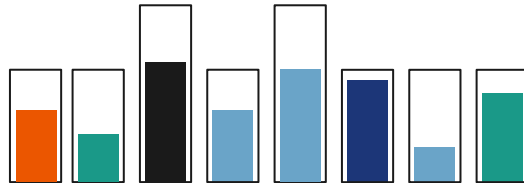


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initially(forall(o in Orders)(placedIn[o] = o))  
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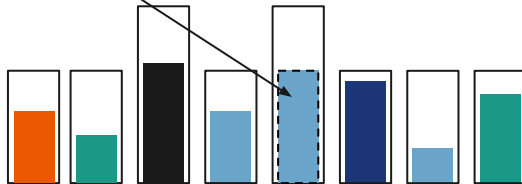
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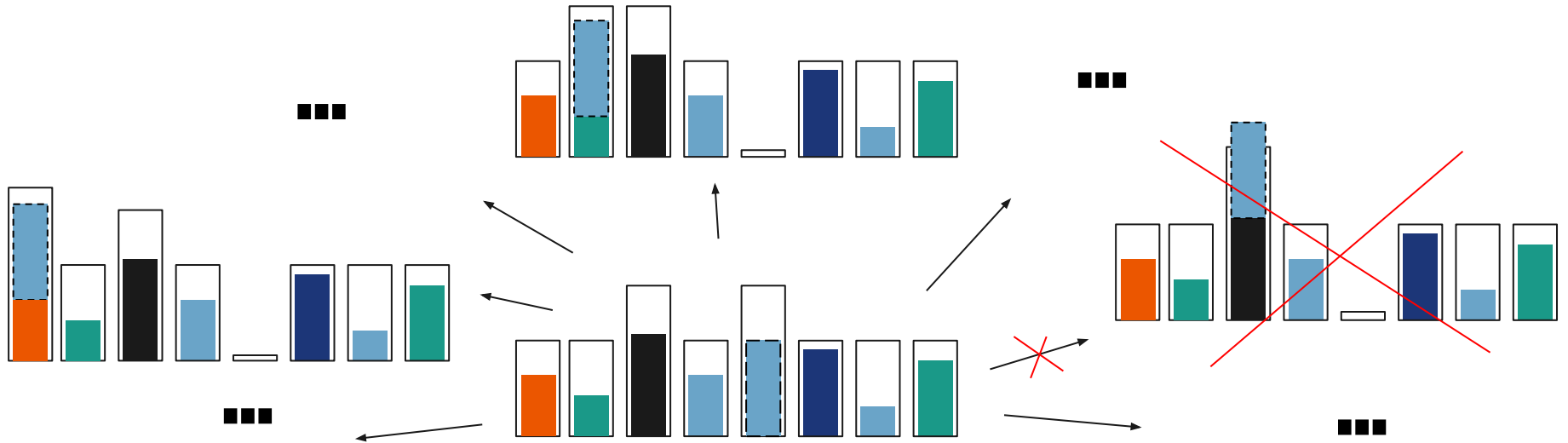
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Move an order to another slab,
preserving the colour and
capacity constraints.



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Under the Hood

Some Highlights



We extend FlatZinc with **flat functions**.

Initialisation done using the CP solver `OscAR.cp`.

Move pre- and post-conditions checked using `OscAR.cbis` constraint system.

Experimental Evaluation

Setup



1. Steel mill slab design is solved using CBLS in [1].
We recreated their neighbourhoods in MiniZinc.
2. We added neighbourhoods to existing MiniZinc models.

We only compare the results versus fzn-oscar-cbls (black-box).

1. Schaus, P., Van Hentenryck, P., Monette, J.N., Coffrin, C., Michel, L., Deville, Y.: Solving steel mill slab problems with constraint-based techniques: CP, LNS, and CBLS. *Constraints* 16(2), 125–147 (April 2011).

Results Compared to Black-Box



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For neighbourhoods similar to the black-box ones:

- an overall **slowdown**
- this is expected

Conclusion & Future Work

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A good starting point

- Grey-box
- Declarative neighbourhoods are expressive and powerful
- One can now experiment with neighbourhoods in MiniZinc

More to do

- White-box
- Declarative language for expressing heuristics and meta-heuristics
- Overheads to trim

Questions?

Thank You!