Solving a game of domino with MiniZinc and ASP

The problem is as follows:

Problem definition

Given a multiset of stones, each made of two labelled sides, try to place as many as possible of them on an $n \times n$ grid while respecting the following rules:

- Stones may not overlap, neither partially nor fully
- Stones may not leave the board, neither partially nor fully
- Each stone shall be placed so that at least one of its sides is adjacent to a side of the previous stone and with matching label
- Each stone may only be placed once

Stones are labelled with numbers from 0 to 6 included

A simple python program has been written to automatically generate instances and benchmark various models with them while recording solutions and statistics. Models, python program and benchmarks can all be found in this:repository

Code snippets of MiniZinc and ASP will be shown as screenshots rather than code blocks due to the lack of syntax highlighting

MiniZinc Program

The MiniZinc model expects the multiset of stones as an $n \times n$ matrix S where element S[a,b] indicates the amount of available stones with label (a,b). Such an input format has been chosen to inherently combat identical-stones symmetries (i.e. situations where the solver will try to swap two identical stones when trying to find a better solution)

The solution is calculated as two $2\times n^2$ matrixes P and C (Placements and Coordinates respectively in the program). P[1,n] indicates the stone label used at time n while P[2,n] indicates the label on the "other side" of that stone. Similarly C[1,n] and C[2,n] indicate the two coordinates of the n-th placement

Number labels are increased by 1 while within the program so that 0 can be used as a special value that indicates missing placements

The constraints that define the validity of a solution candidate are then as follows:

Please note how each placement describes only "half of a stone" rather than the full stone. For example a solution for a 4×4 grid that placed 8 stones will have 16 "placements". It is for this reason that even and odd numbered placements are treated differently: the former are always the "other side" to the previous placement rather than a new stone

Symmetry breaking

Several additional rules were then added to break symmetries, reducing the space of possible solutions

The most impactful and most numerous of those are related to a unique trait of the problem: the placement of the tiles on the grid and their order are two entirely separate problems and the former has a trivial solution

That is because every stone placement only cares about the previous placements and no other stone. This allows us to force the program to place the stones in a "snake-like" orderly pattern as such:

```
0 0 0 1 1
| | | | | |
2 4 1 3 3

| | | | | |
5 5 1 1 1
```

```
5-5 1-1 #
```

Where the top left corner is always the starting point (# represents an empty space and the two sides of a single stone are connected by a line)

This is achieved by using the following constraints

The performance impact of such constraints is massive, going from taking over 20 seconds to solve an 11×11 instance to just under 2 seconds

Some other constraints have been also added to improve performance

```
%Break symmetry that would allow to place the same tiles but in reverse order
constraint (n > 1) -> (Placements[1, 1] <= Placements[1, n^2 - non_placements]);
%Stop if all tiles have been used or entire grid has been filled
constraint non_placements >= max(n^2 - sum(i in 1..7, j in 1..7)(Stones[i,j]), 0);
%Coordinates of non-placements are irrelevant
constraint forall(i in 1..n^2)(Placements[1, i] = 0 -> (Coordinates[1,i] = 1 /\ Coordinates[2,i] = 1));
```

Iterative improvements through benchmarking

By analyzing the results of multiple benchmarks, several different performance enhancing features have been added to the program while iterating upon it. All such benchmarks can be found in the benchmarks folder, where the statistics and results are stored in a labelled CSV format and the solution to each specific instance is stored in a separate file in a human-readable format

To save some time each program being tested had a total maximum amount of time allowed for each benchmark batch. Once it ran out of such time it wasn't allowed to even attempt the harder instances in the batch. This maximum time has been set to be very generous however so that the benchmarks could still be useful and exhaustive for the better and more interesting programs

The first few tests where trying to check the effectiveness of the additional constraints explained above. After that various MiniZinc heuristics were tested. Finally several benchmarks have been ran to find the best restart strategy and the results are as follows:

Program	Board size	Stones	Average time (secs)	Average cost	Samples				
Trivial									
best model candidate A	6	41	0.47	0	10				
best model candidate B	6	41	0.45	0	10				
Very easy									
best model candidate A	10	56	1.27	0	10				
best model candidate B	10	56	1.26	0	10				
Easy									
best model candidate A	18	179	4.28	0	10				
best model candidate B	18	179	4.29	0	10				
Medium									
best model candidate A	25	343	29.91	1	6				
best model candidate B	25	343	29.66	1	6				
Hard									
best model candidate A	30	495	78.25	0	4				
best model candidate B	30	495	80.32	0	4				
Very hard									
best model candidate A	33	598	56.40	1	3				
best model candidate B	33	598	55.89	1	3				
Extreme									
best model candidate A	40	880	279.66	0	2				
best model candidate B	40	880	143.91	0	2				

Unfortunately any instance bigger than 40×40 would result in memory issues on the benchmarking machine and thus couldn't be tested

The best program in full is thus as follows:

```
include "globals.mzn";
par int: n:
array [1..7,1..7] of int: Stones;
array [1..2, 1..n^2] of var 0..7: Placements;
array [1..2, 1..n^2] of var 1..n: Coordinates;
constraint forall(i in 1..n^2)(Placements[1, i] = 0 <-> Placements[2, i] = 0);
constraint all_different(i in 1..max(1, (n^2) - non_placements))(Coordinates[1,i]*(n+1) + Coordinates[2,i]);
constraint forall(i in (1..n^2 div 2))(Placements[1, 2*i] = Placements[2, 2*i - 1] /\ Placements[2, 2*i] = Placements[1, 2*i - 1]);
constraint forall(i in (1..(n^2-1) \text{ div } 2))(Placements[1, 2*i + 1] = Placements[1, 2*i])
                                                 \/ Placements[1, 2*i+1] = 0 \/ Placements[2, 2*i+1] = 0);
constraint forall(i in 2..max(2,n^2 - non_placements))(abs(Coordinates[1, i] - Coordinates[1, i-1]) + abs(Coordinates[2,i] - Coordinates[2, i-1]) = 1);
                                               sum(k \text{ in } 1..n^2)(if (Placements[1,k] = i / Placements[2,k] = j)
                                                                 then 1 else 0 endif)
                                             <= Stones[i,j]);
%If the board has an odd side length, the last placement is always empty constraint (n mod 2 = 1) -> (Placements[1, n^2] = 0);
constraint forall(i in 1..n^2-1)(Placements[1, i] = 0 -> Placements[1, i+1] = 0);
constraint Coordinates[1.1] = 1 /\ Coordinates[2.1] = 1:
constraint forall(i in 1..(n^2-1))((Placements[1, i] != 0 /\ Placements[1, i+1] != 0) -> (
    (Coordinates[2,i] <= Coordinates[2,i+1])</pre>
%Prevent placements from going "down" twice in a row constraint forall(i in 1..(n^2-2))((Placements[1, i] != 0 /\ Placements[1, i+1] != 0 /\ Placements[1, i+2] != 0) -> (
    ({\tt Coordinates[2,i+1]} \; \leftarrow \; ({\tt Coordinates[2,i+1]} \; \rightarrow \; ({\tt Coordinates[2,i+1]} \; = \; {\tt Coordinates[2,i+2]})
constraint forall(i in 1..(n^2-1))((Placements[1, i] != 0 /\ Placements[1, i+1] != 0) \rightarrow (
    ({\tt Coordinates[1,i]} \ {\tt 1} \ / \setminus \ {\tt Coordinates[1,i]} \ {\tt < n)} \ {\tt ->} \ ({\tt Coordinates[2,i]} \ {\tt = Coordinates[2,i+1]})
*Break symmetry that would allow to place the same tiles but in reverse order constraint (n > 1) -> (Placements[1, 1] <= Placements[1, n^2 - non_placements]);
constraint non_placements >= max(n^2 - sum(i in 1..7, j in 1..7)(Stones[i,j]), 0);
constraint \ forall (i \ in \ 1...n^2) (Placements[1, \ i] = 0 \ \rightarrow \ (Coordinates[1,i] = 1 \ / \ Coordinates[2,i] = 1));
var int: non_placements = count(Placements[1, 1..n^2], 0);
     :: relax_and_reconstruct([Placements[j,i] | i in 1..n^2, j in 1..2], 15)
         restart_luby(floor((n^2) / 4))
      else
        restart_none
    :: if n > 33 then
   int_search(Placements, dom_w_deg, indomain_random)
       else
         int_search(Placements, input_order, indomain_min)
       endif
    :: int_search(Coordinates, input_order, indomain_min)
    :: int_search(Coordinates[1,..], input_order, indomain_min)
:: int_search(Coordinates[2,..], input_order, indomain_min)
    minimize non_placements;
```

ASP Program

The ASP Program has been developed by first trying to "port over" the same ideas used in MiniZinc but that route quickly proved itself to be wrong and inefficient

A novel approach was thus needed and so the ASP model ended up being based on follow predicates. In this program each stone is given as a stone((N1, N2), ID) fact where (N1, N2) are the numbers on the stone and ID is a unique identifier for it so that multiple identical stones could be provided

A follow((N1, ID1, T1), (N2, ID2, T2)) predicate thus indicates that the number N2 from the stone identified by ID2 has been placed at time step T2 and is followed by the number N1 from the stone identified by ID1 at time step T1. Similarly to before we work with single numbers and single cells rather than entire stones at once. The reason for the inclusion of a time variable in the follows predicate will be explained later

Such predicates are handled by the following rules:

It is worth noting that the triple (-1, -1, T) in the left side of a follow indicates that the last placement has been made at time T-1 and such placement thus doesn't have any following it. Most of those rules are self-explanatory but they often use some helper predicates defined as such:

```
%%%%%%% BASIC HELPER FACTS %%%%%%%
coordinates(1..n,1..n).
time(1..n**2).
stone_opposite(B, (A, ID)) :- stone((A, B), ID).
stone_opposite(A, (B, ID)) :- stone((A, B), ID).
stone_half(N, ID) :- stone((N, _), ID).
stone_half(N, ID) :- stone((_, N), ID).
```

The last two rules for the follow predicate are the most relevant ones: the former defines the rules of how a stone can follow another one while the latter ensures that every half-stone placed is followed by its other half. Rather than checking whether the time is odd or even like in MiniZinc these two rules alternate each other by checking the "shape" of the follow that came before the one being currently analyzed

Both to decide the coordinates of each placement and to conform to a certain expected output format all the <code>follow</code> predicates are then used to infer <code>placement(X, Y, T, N, ID)</code> predicates which indicate that number $\mathbb N$ from the stone identified by $\mathbb ID$ has been placed at coordinates $\mathbb X$, $\mathbb Y$ at time step $\mathbb T$. This is achieved using the following rules:

follow utilizes a time step variable T to distinguish the left and the right triple in every possible scenario so that the third placement rule cannot recursively "place" the same follow predicate over and over. Without the time step variable or a similar additional variable a predicate such as follow((2,4), (2,4)) would result in the stone stone((2,2), 4) being placed over and over again over the entire grid

And finally the time of the final placement is maximized, resulting in the following complete program:

This program, however, is considerably slower than those defined using MiniZinc, as shown by the following benchmarks:

Program	Board size	Stones	Average time (secs)	Average cost	Samples			
Easy								
Clingo follows-based model	4	13	0.20	0	20			
Medium								
Clingo follows-based model	5	15	57.33	1.2	10			
Hard								
Clingo follows-based model	7	27	395.26	5	1			