CS3243 Group 25 Project 2: Sudoku Solver

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1   Problem Specification

CSP search algorithm uses the structure of states and heuristics to solve problems. We are given the task of implementing a Sudoku solver using backtracking algorithm and heuristics.

The puzzle is modeled as a list of 81 cells where each cell is represented by a (x, y) coordinate where $1 \leq \mathtt{x}, \mathtt{y} \leq 9$. Every cell is assigned a value $\mathtt{v}$ where $0 \leq \mathtt{v} \leq 9$. An empty cell is where $v = 0$. The variables are the coordinates of empty cells in the given input puzzle. A list `unassigned_var_list` contains the unassigned variables. The constraints are that for any of the 9 rows, 9 columns and 9 diagonals, no value can appear more than once. To enforce this constraint, at the start, we initialise 27 lists containing the legal values for each row, column and diagonal after removing the pre-assigned values in the given input puzzle. Then, we keep track of the domain of each variable containing the legal values. The domains are updated to be consistent with the 27 lists.

Afterwards, we begin the backtracking algorithm. A state is represented by the `unassigned_var_list` and `assignments`. When a value is assigned to a variable, the variable is removed from `unassigned_var_list`, its domain becomes empty and `assignments` is updated. After every assignment, we use heuristics to remove illegal values from the domains of neighbouring variables. Hence, the constraints are guaranteed not to be violated throughout.

An assignment is complete when the `unassigned_var_list` is empty. Afterwhich, we add the `assignments` to the original puzzle.

2   Algorithm variants

We implemented a backtracking solver that incorporates heuristics in 3 aspects: Variable Ordering, Value Ordering and Inference

2.1   Variable Ordering

We prioritized the variable for expansion that is the most constrained. For this, we experimented with 2 heuristics: Minimum Remaining Value (MRV) and Degree.

Most Constrained Variable: Minimum Remaining Value Minimum Remaining Value chooses the variable with the least number of legal values. This is simply the variable with the smallest domain size. MRV is implemented in the function `minimum_remaining_values`.

Most Constraining Variable: Degree Degree heuristic acts as a tie-breaker for the MRV heuristic. Among variables that have least and equal domain sizes, the variable with the most unassigned neighbours is prioritized. Degree heuristic is implemented in the function `degree`.

2.2   Value Ordering: Least Constraining Value

Least Constraining Value heuristic prioritizes the value of a variable that results in the minimum reduction in the domains of neighbouring variables. In our implementation, we count the number of conflicts between the value and the values of its neighbours. A conflict happens when the value of the target variable is contained in the domain of its neighbour. The value with the least number of conflicts is prioritized.

2.3   Inference

Inference is effective in reducing the domains of variables. Whenever a value is assigned to a variable, we can infer new domain reductions on its neighbouring variables. We have experimented with 2 variants: Forward Checking (FC) and Arc Consistency (AC-3).

Forward Checking Forward Checking ensures that a value assignment is consistent with the domains of all its neighbouring variables. Fundamentally, it ensures that the assigned variable is arc-consistent with the other variables. FC prunes the the domain values of variables that are in conflict with the value assignment. Failure is detected when any of the reduced domains is empty as a result. This is implemented in the function `forward_checking`.

AC-3 Whereas FC ensures arc consistency between a certain variable and the other variables, AC-3 ensures arc consistency between any 2 variables after every value assignment. This requires an initilisation of binary constraints where we create a tuple for each pair of neighbour variables as seen in the function `init_binary_constraints`. After every assignment, we loop through the list of binary constraints, reducing the domains whenever necessary. Whenever a domain is reduced using the `revise` function, the binary constraints of the neighbours are added to the queue. Evidently, this can be computationally expensive. This is implemented in the function `ac_3` and the subfunction `revise`.

3   Experimental Setup

We run different variants of the aforementioned heuristics. We compare the performance of each variant by measuring the running time. This allows us to compare the efficiency of the variants in terms of time. The runtime of the variant accounts for all the work done and computations by the involved heuristics.

We find it inaccurate and difficult to compare the search space of different variants via computing the number of traversed nodes. It is impossible to establish a fair definition of what a node represents due to the different implementations of various heuristics. For instance, for AC-3, on top of the number of value assignments made before arriving at a solution, it also requires checking every binary constraint everytime an assignment occurs. For FC, it requires traversing the domains of remaining unassigned variables. One way to resolve is simply to define it to be the number of assignments made (equivalently, the number of calls to the `recursive_backtrack` function. However, that is clearly too trivial and inaccurate as the additional work done by the heuristics is not accounted for.

We also test the variants on the world's hardest Sudoku puzzle(Fig 1) found online (https://www.conceptispuzzles.com/index.aspx?uri=info/article/424).

```
_ _ _ _ _ _ _ _ _
|8 0 0 0 0 0 0 0 0|
|0 0 3 6 0 0 0 0 0|
|0 7 0 0 9 0 2 0 0|
|0 5 0 0 0 7 0 0 0|
|0 0 0 0 4 5 7 0 0|
|0 0 0 1 0 0 0 3 0|
|0 0 1 0 0 0 0 6 8|
|0 0 8 5 0 0 0 1 0|
|0 9 0 0 0 0 4 0 0|
 _ _ _ _ _ _ _ _ _
```

Fig. 1. World's Hardest Sudoku

4   Results and Discussion

The results for runtime performance are given in Fig 2 and 3. Runtimes longer than 300 seconds are indicated by a '-' symbol.

From Fig 2, FC+MRV+LCV and AC3+MRV+LCV have the fastest overall run times for the Sudoku puzzle. Between these 2 variants, FC+MRV+LCV is faster for easier puzzles, whereas AC3+MRV+LCV is faster for harder puzzles. This is because AC3 leads to a significant reduction in search space, especially for harder puzzles with enormous search space. For simpler puzzles, however,

ie. input1 and hardest, the reduction in runtime due to smaller search space in AC3 is outweighed by the reduction in runtime due to less node traversal and computation in FC. This is due to the amount of computation needed per assignment. AC3 has to ensure arc consistency between every 2 variables, whereas FC simply ensures consistency between a certain variable and other variables. This is further confirmed when you compare AC3 and FC in Fig 3.

Implementing Degree heuristic, denoted by D , slows down the search. This is because the increase in time spent on computing the degree of every variable outweighs the reduction in time due to smaller search space. In fact, the purpose of the Degree heuristic is merely for tie-breaking for MRV. The extra work of computing the degree is not worth it.

It can also be noted that MRV and LCV play a significant role in the decrease in runtime.

Run time (seconds)	FC+MRV+LCV+D	AC3+MRV+LCV+D	FC+MRV+LCV	AC3+MRV+LCV
Sudoku (input1)	0.912	0.826	0.347	0.251
Sudoku (input2)	0.082	0.085	0.003	0.016
Sudoku (input3)	0.023	0.009	0.052	0.002
Sudoku (input4)	0.018	0.012	0.019	0.003
Sudoku (hardest)	0.901	0.825	0.349	0.253

Fig. 2. Experiments performed to measure runtime with various heuristics on Sunfire

Run time (seconds)	FC+MRV	FC+LCV	AC3+MRV	AC3+LCV	AC3	FC
Sudoku (input1)	0.912	-	0.616	-	120.42	236.654
Sudoku (input2)	0.013	0.095	0.022	0.033	0.141	0.520
Sudoku (input3)	0.023	0.002	0.146	0.023	0.023	0.110
Sudoku (input4)	0.011	0.003	0.802	0.011	0.018	0.350
Sudoku (hardest)	0.926	-	0.614	-	120.390	236.671

Fig. 3. Experiments performed to measure runtime with various heuristics on Sunfire

5 Further Evaluation between FC+MRV+LCV and AC3+MRV+LCV

If we must decide between these 2 variants, we will choose AC3+MRV+LCV. Even though this variant is slower for simpler puzzles, the difference in time is small ($\leq 0.1s$). For complex puzzles, however, the time difference is much larger, up to 2 times faster(compare AC3 and FC). Using the stronger but more computationally expensive AC3 heuristic is preferred to FC.