

Sudoku Solver

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Introduction

Enjoying a worldwide popularity, Sudoku never fails to appeal to programmers. This report mainly introduces the implementation of some human-like algorithms to solve Sudoku in Haskell.

Algorithm

Solving Algorithm 1: Depth-first Search (Brute Force/Backtracking) (Tony's Skeleton Solution)

The solver searches the puzzle from left to right, top to bottom, attempting to find a digit from 1-9 in each blank cell until it satisfies the three constraints of the game rules. If the constraints cannot be satisfied whatever the value of the position is, the solver backtracks to the previous blank position and replaced the filled digit there into another new potential digit. In this way the solver continuously performs the search until all the potential solutions are searched out and recorded.

Solving Algorithm 2: Prioritize the Guessing Order (Extension 1)

In the solver level 0, I implement a function called `expand` which is to guess the value in a cell without considering the guessing order. Based on human logic, the simplest optimization figured out is to prioritize the guessing order, and that is what the Extension 1 requires. It provides a solution that we can fill the block with maximum candidates in it. Using the same logic mentioned in the assignment description, I defined a new data type `Choices` to describe the list of values/possibilities to choose for a single position. If the `Choices` for a unit cell contains only one value, this value must be the solution filled here. Therefore, we can always give preference to the position with the fewest potential possibilities (larger than 1), thereby reducing redundancy dramatically, especially when the Sudoku puzzle is complicated by naïve guessing.

Solving Algorithm 3: Constraint Propagation (Extension 2)

- **Hidden Singles (Solver level 1 in code)**

Optimize the Backtracking process by trimming the known value from the potential `Choices` of other cells which share `block(row/column/box)` with the known position. This constraint is consistent with the game rule that, in a block, one digit can merely occur once. (Note: column, row & box are all called as "block".)

1	8 6	9
8 6	4	5
7	2 5 3	3

To illustrate the operation, we consider an example above: due to the uniqueness principle, we have to delete the potential Choices of digit 3 and 5 from the second cell in the last row. The given circumstance is a box, the same with all other blocks (row/column) as well. So, each known unit cell can constrain 3 blocks related to it, thereby removing this known digit from the candidates of other cells in this 3 blocks and resulting in a dramatic decrease in redundancy.

- **Prune Naked Pairs (Solver level 2 in code)**

A 3*3 box is given below to explain this optimization:

1	8 6	9
8 6	4	5
7	2 8 6	3

As the given box illustrates, the digit(s) in each cell is the potential Choices. We consider a certain situation: if two cells in a block contain 2 same choice values 8 and 6, then no other cell contain any of them due to the simple reason that if there exist some other cells contained this 2 values, and a digit were selected, it is inevitable that one of the original cell would have no choice to fill and would the program would backtrack which led to redundancy. Therefore, we can remove this 2 values from all other related cells' Choice list to avoid meaningless backtracking and help to reduce redundancy. (Note: define "related cells" as those cells sharing blocks with the given cell.)

- **Prune Naked Triples (Solver level 3 in code)**

1 2	2 8 6	8 9
2 8 6	4	5 6
7	2 8 6	3

This level just resembles pruning naked pairs: if there exist 3 cells containing 3 same values, remove these 3 digits from all other cells' choice list.

Generating Algorithm: Las Vegas Algorithm (Extension 3)

Jiang (2009) suggests that using digging holes method to generate a Sudoku puzzle by erasing several digits in confirmed cells of a terminal pattern. In order to boost the generation process, Jiang's team (2009) applies the greedy strategy to digging holes. Once a filled cell is dug out, the following operations are forbidden from filling another digit into the cell again.

- Check the whether the given Sudoku is valid
- Set constraints to the Sudoku when finished
- Check whether a Sudoku has a unique solution by a Depth-First-Search solver
- Prune the searching: Set Non-duplicates constraints in terms of each block)
- Perform propagating at a dug-out puzzle to raise the diversity of the output Sudoku

Implementation

1. Basic Skeleton for Brute Force: (Solver - LEVEL 0)

This part consists of some basic functions, which completely follow the given assignment instructions.

Three functions were constructed to constrain the validity of a Sudoku: isSudoku constrains the whether this puzzle is 9*9 representation; okSudoku constrains the duplicates; noBlanks constrains the Sudoku is finished without “Nothing” cell respectively.

The given skeleton uses the Pos type to represent the cell position. In order to find all the blanks (“Nothing” cells), I indexed the Sudoku puzzle and mapped the getBlanks function to the whole puzzle after the previous map getBlanksRow and finally constructed the allBlankPos.

Update and (!!=) are helper functions to change the position coordinate. I also add the QuickCheck property here to check whether the length remains the same after replacing an element.

Solve level 0 is a naïve searching without any optimization, Brute Force solution. The main purpose here is to discuss different situations and how to deal with the input. Apparently, Guards are quite useful. First, we have to discuss whether the input can be converted to a valid Sudoku. Since the input type is string but our main type during operating process is Sudoku, we have to construct fromString & toString to convert types and complete the IO function. In order to convert 81-character canonical string to a 9*9 Sudoku, firstly we have to group the string into a matrix (a nested list consisting 9 lists and each list has 9 members). Then we have to split the string as 9 sub-strings and covert the sub-string to a row (list). As helper functions, group, chunk and convert were constructed

For the boxes function, it converts a matrix to a list of blocks(rows). Each row contains 3 elements and it has 3 rows for a box. The chop helps to split sub-lists into 3 elements. We zipped the 3-element groups using transpose. The last step is to flatten it using concat and map.

This kind of Top-down logic is frequently applied in Haskell programming. The same with the toString part. Just analyse what we have and what we want, so what kind of function we should construct to help me to approach the expected output. That is why selectRow is constructed in toString part.

2. Prioritize the Guessing Order: (replace expand by expandFirst Function)

Bird (2006) points out in his paper: A program to solve Sudoku. The best choice of cell to perform expansion is the smallest-choices one (greater than 1). min choice can also be improved by adding a breaking situation that when a cell only has 2 choices, it will be selected and the filter can be ended.

3. Constraint Propagation

1) Hidden Singles (Solver - LEVEL 1)

First, we have to filter the single value choice list, and remove it using minus, and then the pruning part: according to the calculation given by Bird (2006) based on 3 laws:

$$\text{filter } (p \cdot f) = \text{map } f \cdot \text{filter } p \cdot \text{map } f$$

$$\text{filter } (\text{all } p) \cdot \text{cp} = \text{cp} \cdot \text{map } (\text{filter } p)$$

$$\text{filter } \text{nodups} \cdot \text{cp} = \text{filter } \text{nodups} \cdot \text{cp} \cdot \text{reduce}$$

Let f is one of the rows/cols/boxes,

$$\text{filter } (\text{all } \text{nodups} \cdot f) \cdot \text{mcp} = \text{filter } (\text{all } \text{nodups} \cdot f) \cdot \text{mcp} \cdot \text{pruneBy } f$$

since,

$$\text{filter } p \cdot \text{filter } q = \text{filter } q \cdot \text{filter } p \text{ (in this case)}$$

we can finally rewrite:

$$\text{prune} = \text{pruneBy } \text{boxs} \cdot \text{pruneBy } \text{cols} \cdot \text{pruneBy } \text{rows}$$

Calculation process is shown below (Bird, 2006):

Here is the calculation. Let f be one of *rows*, *cols* or *boxs*:

$$\begin{aligned} & \text{filter } (\text{all } \text{nodups} \cdot f) \cdot \text{mcp} \\ = & \quad \{\text{since } \text{filter } (p \cdot f) = \text{map } f \cdot \text{filter } p \cdot \text{map } f \text{ if } f \cdot f = \text{id}\} \\ & \text{map } f \cdot \text{filter } (\text{all } \text{nodups}) \cdot \text{map } f \cdot \text{mcp} \\ = & \quad \{\text{since } \text{map } f \cdot \text{mcp} = \text{mcp} \cdot f \text{ if } f \in \{\text{boxs}, \text{cols}, \text{rows}\}\} \\ & \text{map } f \cdot \text{filter } (\text{all } \text{nodups}) \cdot \text{mcp} \cdot f \\ = & \quad \{\text{definition of } \text{mcp}\} \\ & \text{map } f \cdot \text{filter } (\text{all } \text{nodups}) \cdot \text{cp} \cdot \text{map } \text{cp} \cdot f \\ = & \quad \{\text{since } \text{filter } (\text{all } p) \cdot \text{cp} = \text{cp} \cdot \text{map } (\text{filter } p)\} \\ & \text{map } f \cdot \text{cp} \cdot \text{map } (\text{filter } \text{nodups} \cdot \text{cp}) \cdot f \\ = & \quad \{\text{property of } \text{reduce}\} \\ & \text{map } f \cdot \text{cp} \cdot \text{map } (\text{filter } \text{nodups} \cdot \text{cp} \cdot \text{reduce}) \cdot f \\ = & \quad \{\text{since } \text{filter } (\text{all } p) \cdot \text{cp} = \text{cp} \cdot \text{map } (\text{filter } p)\} \\ & \text{map } f \cdot \text{filter } (\text{all } \text{nodups}) \cdot \text{cp} \cdot \text{map } (\text{cp} \cdot \text{reduce}) \cdot f \\ = & \quad \{\text{since } \text{map } f \cdot \text{filter } p = \text{filter } (p \cdot f) \cdot \text{map } f \text{ if } f \cdot f = \text{id}\} \\ & \text{filter } (\text{all } \text{nodups} \cdot f) \cdot \text{map } f \cdot \text{mcp} \cdot \text{map } \text{reduce} \cdot f \\ = & \quad \{\text{since } \text{map } f \cdot \text{mcp} = \text{mcp} \cdot f \text{ if } f \in \{\text{boxs}, \text{cols}, \text{rows}\}\} \\ & \text{filter } (\text{all } \text{nodups} \cdot f) \cdot \text{mcp} \cdot f \cdot \text{map } \text{reduce} \cdot f \\ = & \quad \{\text{definition of } \text{pruneBy } f; \text{ see below}\} \\ & \text{filter } (\text{all } \text{nodups} \cdot f) \cdot \text{mcp} \cdot \text{pruneBy } f \end{aligned}$$

The definition of *pruneBy* is

$$\begin{aligned} \text{pruneBy} &:: (\text{MatrixChoices} \rightarrow \text{MatrixChoices}) \rightarrow \\ &\quad (\text{MatrixChoices} \rightarrow \text{MatrixChoices}) \\ \text{pruneBy } f &= f \cdot \text{map reduce} \cdot f \end{aligned}$$

We have shown that, provided *f* is one of *rows*, *cols* or *boxs*,

$$\text{filter } (\text{all nodups} \cdot f) \cdot \text{mcp} = \text{filter } (\text{all nodups} \cdot f) \cdot \text{mcp} \cdot \text{pruneBy } f$$

For the final step we need one more law, the fact that we can interchange the order of two *filter* operations:

$$\text{filter } p \cdot \text{filter } q = \text{filter } q \cdot \text{filter } p$$

This law is not generally valid in Haskell without qualification on the boolean functions *p* and *q*, but provided *p* and *q* are total functions, as is the case here, the law is OK. Indeed we implicitly made use of it when claiming that the order of the component filters in the expansion of *filter correct* was unimportant.

Now we can calculate, abbreviating *nodups* to *nd* to keep the expressions short:

$$\begin{aligned} &\text{filter correct} \cdot \text{mcp} \\ = &\quad \{\text{rewriting } \text{filter correct} \text{ as three filters}\} \\ &\text{filter } (\text{all nd} \cdot \text{boxs}) \cdot \text{filter } (\text{all nd} \cdot \text{cols}) \cdot \text{filter } (\text{all nd} \cdot \text{rows}) \cdot \text{mcp} \\ = &\quad \{\text{calculation above}\} \\ &\text{filter } (\text{all nd} \cdot \text{boxs}) \cdot \text{filter } (\text{all nd} \cdot \text{cols}) \cdot \text{filter } (\text{all nd} \cdot \text{rows}) \cdot \text{mcp} \cdot \\ &\quad \text{pruneBy rows} \\ = &\quad \{\text{interchanging the order of the filters}\} \\ &\text{filter } (\text{all nd} \cdot \text{rows}) \cdot \text{filter } (\text{all nd} \cdot \text{boxs}) \cdot \text{filter } (\text{all nd} \cdot \text{cols}) \cdot \text{mcp} \cdot \\ &\quad \text{pruneBy rows} \\ = &\quad \{\text{using the calculation above again}\} \\ &\text{filter } (\text{all nd} \cdot \text{rows}) \cdot \text{filter } (\text{all nd} \cdot \text{boxs}) \cdot \text{filter } (\text{all nd} \cdot \text{cols}) \cdot \text{mcp} \cdot \\ &\quad \text{pruneBy cols} \cdot \text{pruneBy rows} \\ = &\quad \{\text{repeating the last two steps one more time}\} \\ &\text{filter } (\text{all nd} \cdot \text{rows}) \cdot \text{filter } (\text{all nd} \cdot \text{boxs}) \cdot \text{filter } (\text{all nd} \cdot \text{cols}) \cdot \text{mcp} \cdot \\ &\quad \text{pruneBy boxs} \cdot \text{pruneBy cols} \cdot \text{pruneBy rows} \\ = &\quad \{\text{definition of } \text{filter correct}\} \\ &\text{filter correct} \cdot \text{mcp} \cdot \text{pruneBy boxs} \cdot \text{pruneBy cols} \cdot \text{pruneBy rows} \end{aligned}$$

Hence, we can define *prune* by

$$\begin{aligned} \text{prune} &:: \text{MatrixChoices} \rightarrow \text{MatrixChoices} \\ \text{prune} &= \text{pruneBy boxs} \cdot \text{pruneBy cols} \cdot \text{pruneBy rows} \end{aligned}$$

Readers who gave this solution (or a similar one in which the three components appear in any other order) can award themselves full marks.

I marked the prune as `prune_level_1` in my code since I also implemented `prune_level_2` and `prune_level_3` as well. These three functions are similar to some extent.

2) Naked Pairs (Solver - LEVEL 2)

The first is how to get the same pairs, I defined a function to samepair with QuickCheck property. It can find the repeating pairs in the nested list and return a list with 2 elements, that is, the duplicate pair. Reduce samepair takes a list of choice list and prune the “naked pairs”.

I imported `isInfixOf` to check the sub-list, but there still exists redundancy since I did not remove the single value from other related cells. In other words, I just deal with this situation:

Assume cell 1,2,3 are in the same block.

Choices for cell 1: [Just 2, Just 3]

Choices for cell 2: [Just 2, Just 3]

Choices for cell 3: [Just 2, Just 3, Just 4]

...

Then the choices for cell 3 will be pruned to [Just 4]

However, when it comes to another situation:

Assume cell 1,2,3,4 are in the same block.

Choices for cell 1: [Just 2, Just 3]

Choices for cell 2: [Just 2, Just 3]

Choices for cell 3: [Just 2, Just 4]

Choices for cell 4: [Just 3, Just 5, Just 6]

...

If this algorithm is implemented well, it will find dup_pair = [Just 2, Just 3], and the choice for cell 3 will be updated by [Just 4] (removing Just2); similarly, the choice for cell 4 will be updated by [Just 5, Just 6] (removing Just 2).

I once wrote code like:

```
reduce_samepair x = [if (dup_pair == xs) then xs else xs\\dup_pair | xs <- x]
```

it works well for `easy.txt` with 0.6s by `expandFirst` & `prune_level_2` but when it comes to `hard.txt`, nothing is printed. Probably it did not check the validity then lead to bugs.

Therefore, I changed the code to:

```
[if (isInfixOf dup_pair xs)&&(dup_pair/=xs) then xs\\dup_pair else xs | xs <- x]
```

`isInfixOf` is a built-in function to check the sub_list and returns a Boolean type.

This condition statement is much more restricted and it can be run successful with a 3m58.542s using `expandFirst` and level 2.

3) Naked Triples (Solver - LEVEL 3)

The first is how to get the same triples, I defined a function to sametriple with QuickCheck property. It can find the triples occurring 3 times in the nested list and return a list with 3 elements, that is, the 3-times triples. I constructed the count function to find the frequency of an element in a list. And the anonymous function and filter help to write concise code.

Also I changed my previous code:

reduce_sametriple x = [if (dup_triple/=xs) then xs\\dup_triple else xs | xs <- x]
to:

[if (isInfixOf dup_triple xs)&&(dup_triple/=xs) then xs\\dup_triple else xs | xs <- x]
except the count function and filter 3-times triples which are slightly different from level 2, other functions are quite similar, prune dup triple, reduce same triple as well as the concise version of pruneBy function.

Besides, helper functions were constructed: is_single, is_pairs and is_triple as well as getDups.

4) Enforce Consistency Property

After implementing 4 levels (including the level 0: brute force), we have to set the winner condition: complete. The puzzle is completed when all cells' choice list only contains one valid value. If any position in the puzzle has no choice, this Sudoku is unsolvable. Consistent constrains the "uniqueness" game rule and satisfy apply consistent to each row, column and box.

4. Generating Sudoku (Extension 3)

- based on a paper Named "Sudoku Puzzles Generating: From Easy To Evil"

Test & Analysis (Extension 4)

Easy.txt

	Solver - Level 0	Solver - Level 1	Solver - Level 2	Solver - Level 3
	Brute Force	Hidden Singles	Naked Pairs	Naked Triples
expand Note: (Level 0 doesn't use)	real 5m45.574s user 5m42.720s sys 0m4.686s	real 5m53.522s user 5m49.372s sys 0m4.010s	NULL	NULL
expandFirst	NULL	real 0m1.175s user 0m1.184s sys 0m0.284s	real 0m0.695s user 0m0.720s sys 0m0.255s	real 0m0.789s user 0m0.814s sys 0m0.266s

Hard.txt

	Solver – Level 0 Brute Force	Solver – Level 1 Hidden Singles	Solver – Level 2 Naked Pairs	Solver – Level 3 Naked Triples
expand	Killed	24min27s	NULL	NULL
expandFirst	NULL	real 4m13.680s user 4m11.827s sys 0m3.315s	real 3m58.542s user 3m56.831s sys 0m3.264s	real 5m13.327s user 5m10.379s sys 0m2.624s

Since “Naked triple” is not very general so Solver-level 3 performs a lot of invalid checks that leads to the relatively slower than Solver_level 2.

But we can obviously find the dramatic decline in time if we replace the naïve expand by expandFirst. The guessing order makes a huge difference indeed!

Sudoku17.txt

To deal with this huge file, I found a very fast solver online ([https://wiki.haskell.org/Sudoku#Very fast Solver](https://wiki.haskell.org/Sudoku#Very_fast_Solver)) and I tested it for the Sudoku17.txt, only taking 43.898s to print all the output, approximately 50000 Sudoku!). Data.vector is imported in this solution and this library has a wide range of built-in functions and the complexity of the majority is merely $O(1)$! Indexing, update, imap, ifilter are all built-in. It is so awesome that this Data.vector library seems to be born to solve Sudoku. Besides, the solution provided applies the Monad in an extremely skilled manner, which also enables to boost the solver.

Acknowledgement

I really appreciate those people who shared their thoughts online and the open source platform online. Sudoku is so attractive for Haskell programmers that they even created a wiki Haskell page (<https://wiki.haskell.org/Sudoku>) especially for Sudoku, which I really benefited from.

Reference:

BIRD, RICHARD. "FUNCTIONAL PEARL: A Program To Solve Sudoku". Journal of Functional Programming 16.06 (2006): 671. Web. 7 July 2006.

<http://www.cs.tufts.edu/~nr/cs257/archive/richard-bird/sudoku.pdf>

Functions for solving a Sudoku <https://hackage.haskell.org/package/hsudoku-0.1.1.0/src/src/Sudoku/Solver.hs>

[Haskell-cafe] Norvig's Sudoku Solver in Haskell <https://mail.haskell.org/pipermail/haskell-cafe/2007-August/031049.html>

Human-like algorithms for solving Sudoku <http://www.programming-algorithms.net/article/42521/Sudoku>

Jiang, Biaobin. "Sudoku Puzzles Generating: From Easy To Evil". Chinese Journal of Mathematics in Practice and Theory 39.21 (2009): 1-7. Web. 26 May 2017.

http://zhangroup.aporc.org/images/files/Paper_3485.pdf

Open source project on github: <https://github.com/Freezard/haskell-sudoku/blob/master/Sudoku.hs>

Problem 97 in 99 Haskell Problems <http://mfukar.github.io/2015/12/11/haskell-xvi.html>

Wiki Haskell: A very fast solver https://wiki.haskell.org/Sudoku#Very_fast_Solver