An implementation of a python sudoku solver package and a complete review of the software development process involved.

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1 Introduction

2 Solution Design

2.1 Selection of Solution Algorithm

Why did I choose backtracking?

2.2 Prototyping

Prior to writing any code, we prototyped the solution. Prototyping prior to coding allowed us to

- identify possible bugs and complexity earlier on in the development process, such that we could consider them prior to coding, rather than discover them after the fact;
- identify non-trivial and edge cases that might need to be considered while writing code and unit tests;
- identify API interfaces of the package, allowing us to write the unit tests beforehand (more on this when we talk about test driven development in Section(3.3)).
- explore different implementations of the backtracking algorithm;

• identify python packages and resources that we may need to use in the implementation, and make decisions on which might be the most appropriate;

The implementation of the sudoku solver consisted of two main components: parsing the user input (and handling associated errors), and the backtracking algorithm itself. The prototyping flowcharts of these two components is shown in Fig.(1) and Fig.(2). Additionally, we wrote some *pseudo-python-code* as shown in Fig.(3), allowing us to define the interfaces of the functions and classes that we would write - which is required to write any unit tests before prior to coding. The tests derived from this solution design are discussed more in Section(4). Further iterations on these initial prototypes are discussed more in Section(3) where we discuss how experimentation and profiling changed the end implementation in efforts to optimise the code.

2.3 API Interface

An important decision to make early on is how the code will be structured into modules and what the external API interface will look like, as shown in Fig.(3). We decided to model the sudokusolver package after well known scikit-learn [2] package because scikit-learn's structure is extremely well thought out and simple to use. Each solver would be segmented into own module, containing a class which can be instantiated with the hyperparameters of the solver (in this case, is there multiple_solutions), and then a solve method which takes in the data (the unsolved_board) and returns the class instance with the solved board as an attribute. This seemed like a fitting way to model the package as our solvers are akin to scikit-learn's estimators, and the solve method is akin to scikit-learn's fit method. To extend the package we simply need to add more solvers in their own modules, and to extend a solver we simply need to add more hyperparameters to the class and modify the solve method to take these hyperparameters into account.

2.4 Key Conclusions

We decided to only implement the BacktrackingSolver with no multiple_solutions functionality in the initial v0.0.1 implementation of the package, The implementation of multiple_solutions was not thought through in the prototyping stage but given that we now understood how to syntactically and structurally extend the package and the solvers within, it would be trivial to add this functionality in a future version of the package, which demonstrates the importance of blueprinting out the API interface.

Another key decision which was implicit and not thoroughly discussed was to utilise numpy arrays to represent the sudoku board, this is because numpy is well known to be the most performant array manipulation package in python. Alternatives include using pandas or python's in-built list object, however, numpy is generally known to be more performant than both of these and numpy also includes numerous built-in functions that will likely be useful in the implementation of the backtracking algorithm.

3 Development, Experimentation and Profiling

Here we discuss several components of the development process, and reason why we chose to do things in certain ways

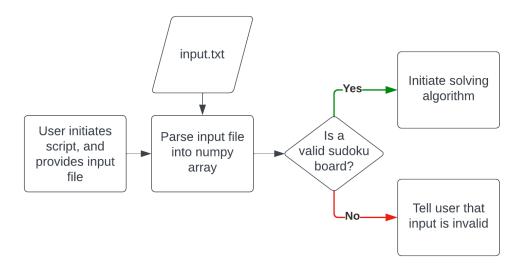


Figure 1. A flowchart for part 1 of solving a sudoku puzzle: parsing the user input.

3.1 Linting and Formatting - ruff

Linting and formatting is useful, especially in shared projects, as it allows for a consistent style across the codebase. There are a plethora of python linting and formatting tools available and for this project, we chose to use ruff. There were two primary reasons for this choice: speed and simplicity. ruff is faster than most other linting tools including flake8. In a test done by the developers of ruff, it managed to lint the CPython codebase 42x faster than flake8 [1]. Whilst speed of linting is not a primary concern for this project, it doesn't hurt to pick the faster option.

Additionally, ruff provides formatting functionality and as such it can also replace tools such as black. This makes the implementation of linting and formatting simpler, as we only need to use one tool. ruff's configuration capabilities allow it to lint and format to any standard we want to, and as such, it was configured to mimic black and flake8's default config in accordance with PEP8.

3.2 Git and Gitlab pipeline

Why branches? Branch naming convention. Creating issues for tasks. Why? Project management. JIRA. New branch and merge per issue. Labels for issues because easy to and can folder the branches with those labels (show PyCharm example).

3.3 Test Driven Development

Explain why I wrote the tests first.

3.4 Profiling and Optimisation

Where did i test different packages for speed? Why did I choose numpy? Show how I profiled my code to identify where the bottleneck was. How did I work around this? Cython?

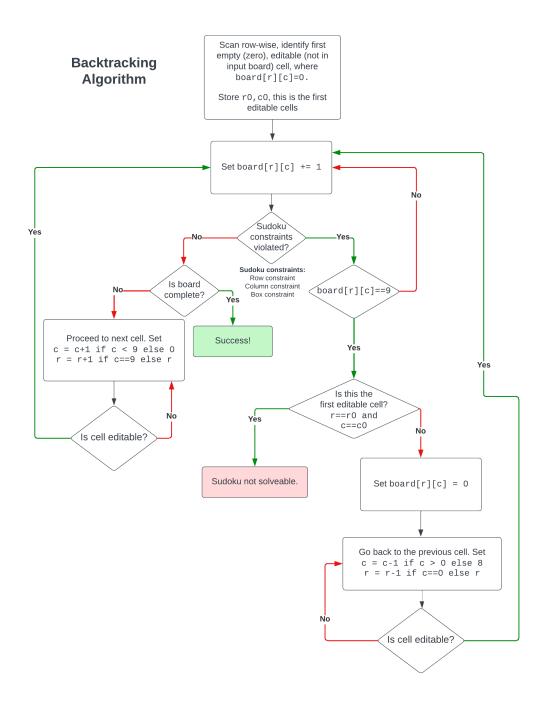


Figure 2. A flowchart for part 2 of solving a sudoku puzzle: the backtracking algorithm.

3.5 Coding Best Practises

Modularisation. typing. Exceptions. Error handling, try except. Never catch all exceptions.

```
class BacktrackingSolver
    def __init__(self, multiple_solutions: bool = False):
        self.board # will store solved board
        self.is_solvable # will store whether board is solvable
        self.is_solved # will store whether board is solved
    def solve(self, unsolved_board: numpy.NDArray) -> None:

def parse_input_file(input_file_path: str) -> numpy.NDArray

def save_board(
    board: numpy.NDArray,
    output_file_path: str
) -> None

# entry point for script, argv = sys.argv

def handler(argv) -> None:
```

Figure 3. *pseudo-python-code* representation of the interfaces of the functions and classes within our sudokusolver package.

4 Validation, Unit Tests and CI set up

Talk through why my unit tests are sufficient. How did I put this into the CI? With precommit.yaml

5 Documentation, Packaging and Usability

5.1 Documentation - sphinx

Why did I use sphinx over Doxygen? Sphinx is more popular, more support, more documentation, more features.

5.2 Packaging - Conda and Docker

What benefits does conda have over pip? Why did I choose conda? Why Docker?

6 Summary

References

- [1] Astral, ruff GitHub Repository. Available at: https://github.com/astral-sh/ruff [Accessed: 7-Dec-2023].
- [2] scikit-learn, schikit-learn GitHub Repository. Available at: https://github.com/scikit-learn/scikit-learn/tree/main [Accessed: 7-Dec-2023].