

AI project documentation

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Project name:

2) A Sudoku Puzzle Solver using the Backtracking Algorithm AND a Genetic Algorithm.

Project link:

<https://github.com/divid7/AI-project>

Sudoku Puzzle Solver Project Documentation

1. Project Idea in Detail

The project involves developing a Sudoku Puzzle Solver that utilizes two algorithms:

1. Backtracking Algorithm: A systematic, recursive method for solving constraint satisfaction problems like Sudoku by exploring configurations incrementally.
2. Genetic Algorithm: An evolutionary approach inspired by natural selection, using operations like crossover and mutation to optimize solutions over generations.

The solver will feature:

- Input Flexibility: Users can input custom Sudoku puzzles or choose from pre-existing ones.
- Customization: Allow different grid sizes (e.g., 4x4, 6x6, 9x9).
- Visualization: Step-by-step explanation of how the solution is derived.
- Performance Metrics: Evaluation of efficiency (iterations, execution time) for both algorithms.

Sudoku Environment Type

The environment of Sudoku can be described as follows:

1. Agent Type:

- **Single Agent:** Only one solver (human or AI) interacts with the puzzle at a time, working independently to reach a solution.

2. Deterministic:

- The outcome of placing a number in a cell is predictable, with no random elements affecting the puzzle state.

3. Static:

- The puzzle is unchanging over time; the initial grid and constraints remain fixed throughout the solving process.

4. Fully Observable:

- All information about the puzzle is available at any time, including the state of every row, column.

5. Discrete:

- The puzzle has finite and distinct states, where each action (placing a number) leads to a specific and measurable outcome.

6. Episodic vs. Sequential:

- **Sequential:** Each action (placing a number) affects future decisions, as constraints are updated dynamically with each move.

This structured, single-agent, sequential environment makes Sudoku an excellent problem for exploring AI algorithms like Backtracking and Genetic Algorithm, as it requires systematically solving under fixed constraints.

2. Main Functionalities

1. Puzzle Input:

- Manual input of custom Sudoku puzzles.
- Selection from predefined puzzles.

2. Algorithm Selection:

- Choice between Backtracking and Genetic Algorithm.

3. Customization Options:

- Selection of grid size for varied complexity.

4. Puzzle Solving:

- Execution of the selected algorithm to solve the puzzle.

5. Visualization:

- Display of step-by-step solving process for user comprehension.

6. Performance Evaluation:

- Metrics including the number of iterations and execution time for comparison between algorithms.

3. Similar Applications in the Market

1. Sudoku.com: A popular Sudoku-solving platform offering pre-generated puzzles for play but lacks algorithmic visualization or custom inputs.

2. Microsoft Sudoku: Provides puzzles of varying difficulties but doesn't showcase the solving process or offer customization in grid size.

3. Sudoku Solver Apps: Many apps allow users to input puzzles and provide solutions but lack algorithmic flexibility and performance comparisons.

4. AI-Based Solvers (Web): Tools leveraging AI for Sudoku but typically don't explain the underlying steps or logic.

4. Literature Review

1. Sudoku as a Constraint Satisfaction Problem:

- Explores methods like backtracking and forward-checking to efficiently solve Sudoku puzzles.
- Reference: Russell, S., & Norvig, P. (2003). Artificial Intelligence: A Modern Approach.

2. Genetic Algorithms for Optimization Problems:

- Discusses the application of genetic algorithms for solving optimization tasks, including Sudoku puzzles.

- Reference: Goldberg, D. E. (1989). Genetic Algorithms in Search, Optimization, and Machine Learning.
- 3. Hybrid Methods for Solving Sudoku:
 - Combines deterministic and stochastic approaches for enhanced efficiency in solving complex puzzles.
 - Reference: Yato, T., & Seta, T. (2003). Complexity and Completeness of Finding Another Solution and Its Application to Puzzles.
- 4. Visualization Techniques for Algorithmic Processes:
 - Examines methods for making algorithms transparent through visualization tools.
 - Reference: Brown, M. H. (1988). Algorithm Animation.
- 5. Comparative Study of AI Algorithms in Constraint Solving:
 - Reviews the strengths and weaknesses of various algorithms, including backtracking and genetic methods.
 - Reference: Dechter, R. (2003). Constraint Processing.
- 6. Performance Metrics for Optimization Algorithms:
 - Discusses frameworks for evaluating algorithmic performance in real-time systems.
 - Reference: Hoos, H. H., & Stützle, T. (2004). Stochastic Local Search: Foundations & Applications.

5. Details of the Algorithms /Approach Used

Why Backtracking?

- Offers a deterministic solution for standard Sudoku puzzles.
- Efficient for small-to-medium grids.

Why Genetic Algorithm?

- Ideal for exploring larger, more complex grids.
- Provides insights into evolutionary optimization.

Backtracking Algorithm:

- Approach: Recursive depth-first search technique. Fills each empty cell sequentially and backtracks when encountering conflicts.
- Implementation Details: Cell-by-cell assignment, checks row, column, and 3x3 subgrid constraints.
- Results: Efficient for smaller grids but scales poorly with increased complexity.

Genetic Algorithm:

- Approach: Starts with a population of random grids, applies crossover and mutation to evolve the population, uses a fitness function based on adherence to Sudoku rules.
- Implementation Details: Population size, mutation rates, and stopping criteria are configurable.
- Results: Performs better on larger puzzles but may require fine-tuning to avoid convergence issues.

Comparison Results:

| Algorithm | Iterations | Execution Time | Complexity Handling | |
|-------------------|------------|----------------|---------------------|--|
| ----- | ----- | ----- | ----- | |
| Backtracking | Low | Fast | Moderate | |
| Genetic Algorithm | High | Moderate | High | |

7. Development Platform

Programming Language:

- Python, leveraging libraries like NumPy and Matplotlib for computation and visualization.

Development Tools:

- IDE: Visual Studio Code.

Libraries/Frameworks:

-Pandas and Numpy

- For Backtracking: Native Python constructs.

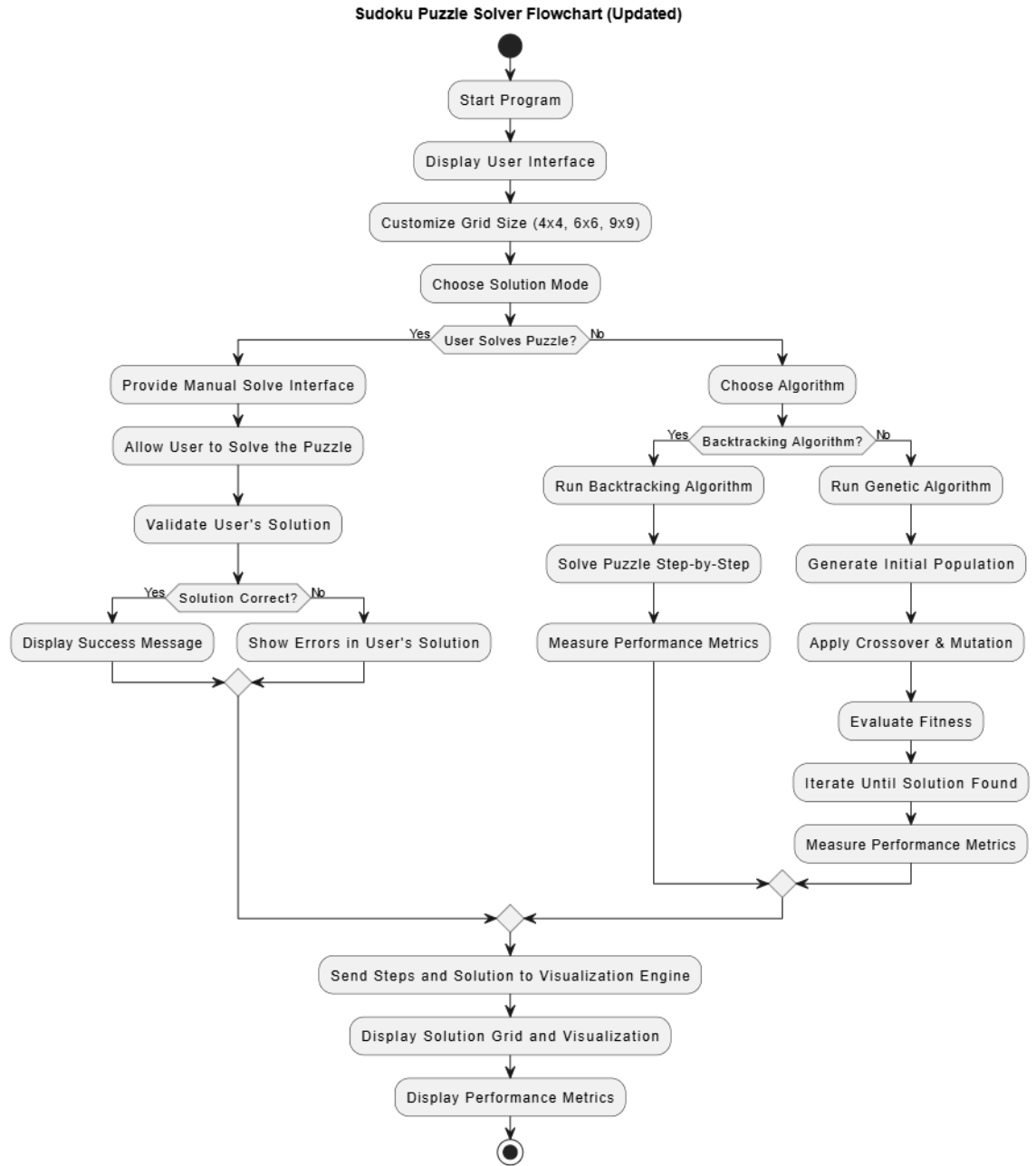
- For Genetic Algorithm: DEAP (Distributed Evolutionary Algorithms in Python).

- For Visualization: Matplotlib and PyQt5 for graphical interfaces.

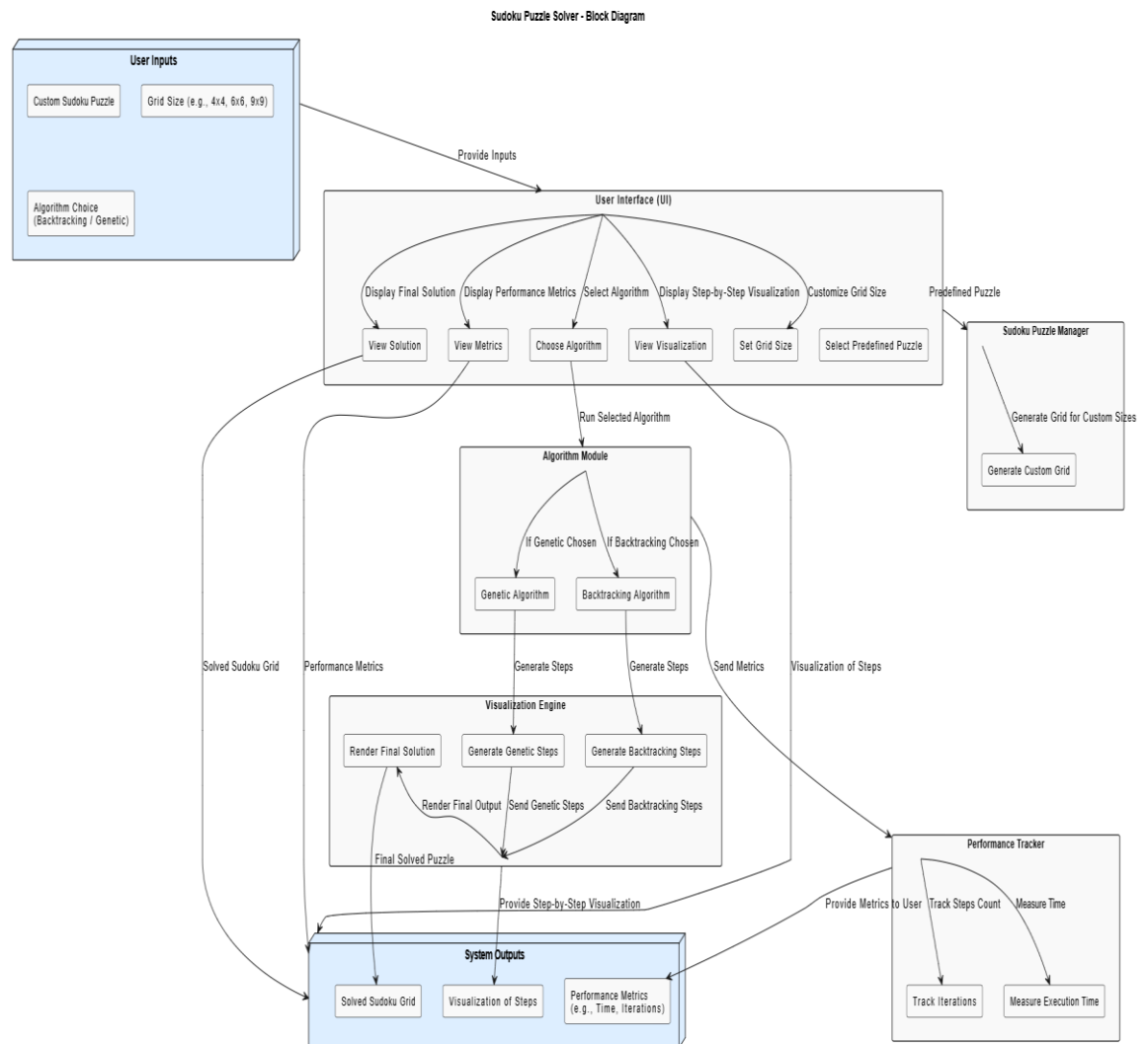
Execution Environment:

- Local systems (Windows) with Python 3.x installed.

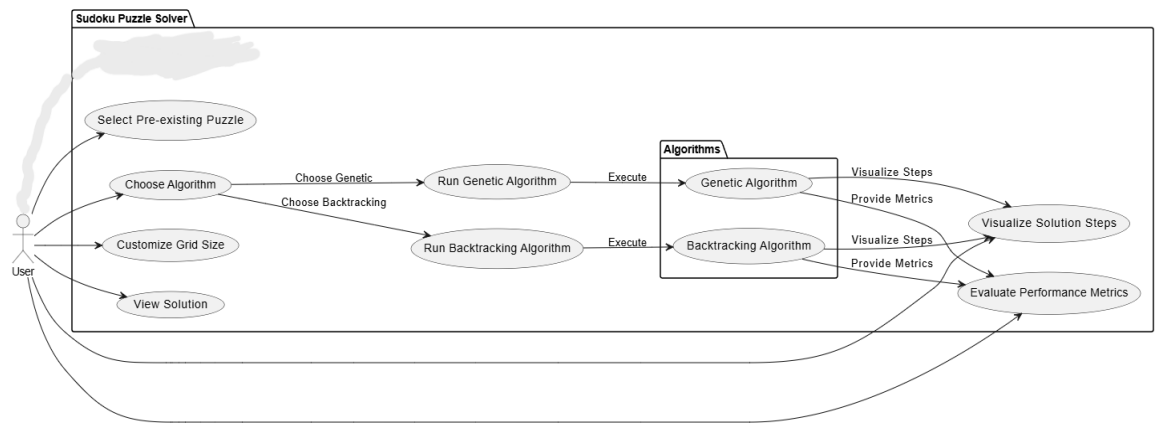
Flow-Chart:



Block Diagram:



Use Case Diagram:



Plot:

