



REPORT : Gauss–Jordan Linear Equation Solver Using Python

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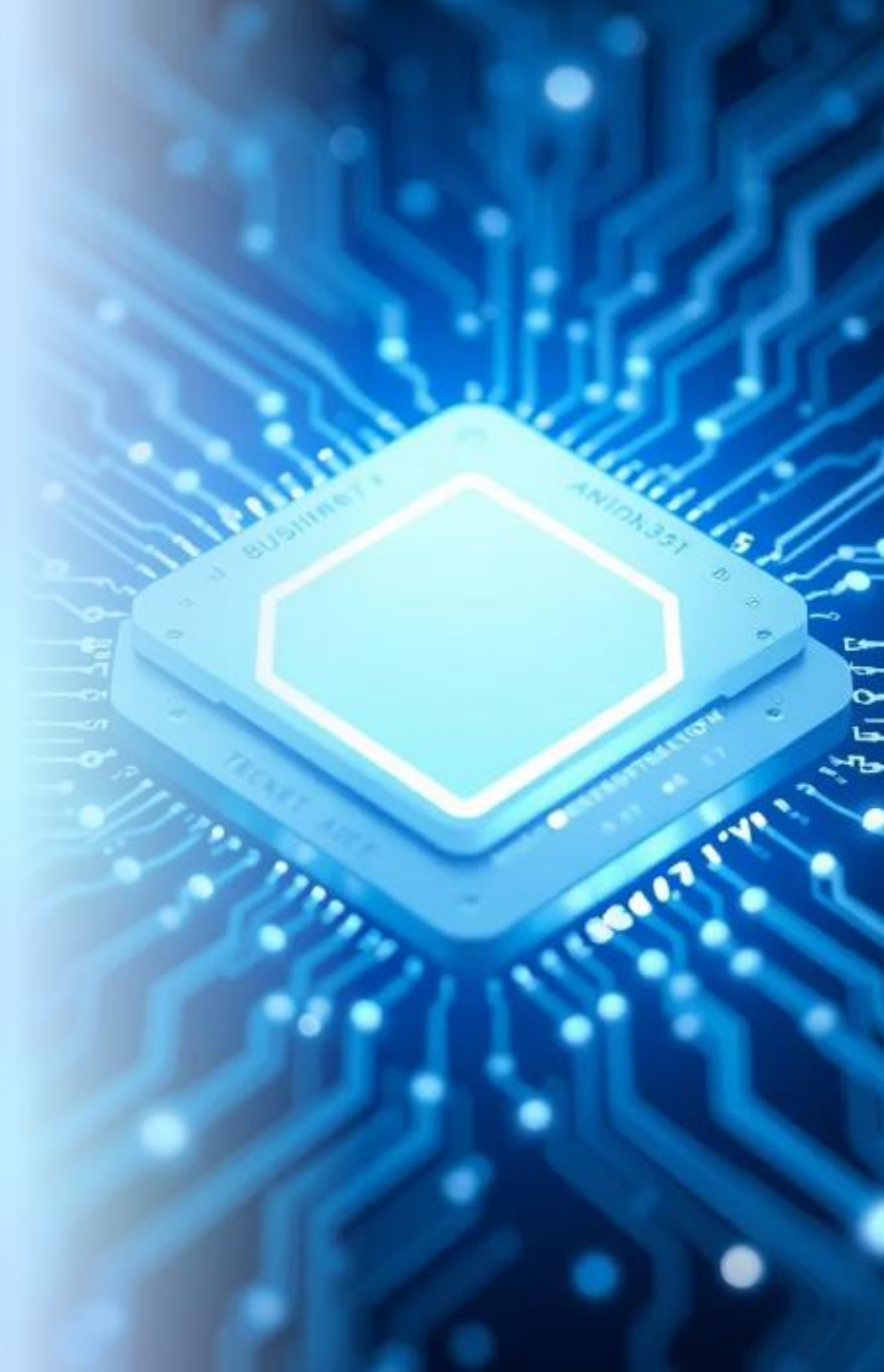
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1. Problem Statement

Solving systems of linear equations is a fundamental task in mathematics, engineering, and computer science. Manually solving large systems is time-consuming and prone to errors. Therefore, there is a need for a computerized solution that can take an augmented matrix as input and automatically compute the values of unknown variables accurately.

This project addresses the real-world requirement of solving simultaneous linear equations using an efficient numerical method — Gauss–Jordan Elimination.





2. Real-World Problem / Need Identified

Many scientific, engineering, and economic problems require solving linear systems, such as:

Network flow
analysis

Electrical circuit
analysis

Statistical
regression

Computer graphics
transformations

Physics and mechanics
problems

A digital solver helps automate these calculations and eliminates human error.

3. Objective / Expected Outcomes

The objectives of this project are:

1

To design a Python program that accepts a user-defined augmented matrix.

2

To solve the system using Gauss–Jordan Elimination.

3

To provide accurate values of the unknown variables.

4

To demonstrate programming concepts learned in the course.

5

To understand matrix transformations and numerical computation.

6

To give users a clean interface, error handling, and clear output.

Expected outcomes:

1

A correct Reduced Row Echelon Form (RREF) matrix.

2

Correct solution for each variable.

3

Smooth user interaction and logically structured code.

4. Concepts Applied from the Course

This project uses several core programming concepts:

Input Handling

Accepting user input for rows, columns, and matrix values.

Typecasting

Converting input strings to int or float.

Loops

- for loops for iterating through matrix rows and columns.
- while loops for repeated input validation.

Nested Loops

Used to build the matrix row-by-row and column-by-column.

Conditional Statements (if-else)

Checking pivot values, handling row swapping, and validating input.

Functions

Modular approach using:

- `get_matrix()`
- `gauss_jordan()`
- `print_matrix()`
- `print_solution()`

Arrays / NumPy

Using NumPy arrays for mathematical operations and matrix manipulation.

Exception Handling

Using try-except to catch invalid user inputs.

Mathematical Algorithm Implementation

Normalized row operations, pivoting, and elimination logic.



5. Appropriate Tools, Libraries, and Techniques

1

Python

selected for simplicity and readability.

2

NumPy

chosen for efficient matrix operations and convenience.

3

Gauss–Jordan Method

reliable and widely used in numerical computation.

4

Structured Programming Techniques

functions, modular design, loops, conditions.



6. Structured Development Process

The development followed a systematic and professional approach.



Problem Definition

To create a program that:

1

Accepts an augmented matrix of size $(n \times (n+1))$.

2

Performs Gauss–Jordan elimination.

3

Outputs the final solution in clear numerical form.

Requirement Analysis

Functional Requirements

- User inputs number of equations.
- User inputs all matrix values.
- Program validates input.
- Program performs row operations.
- Program outputs solution.

Non-Functional Requirements

- Accuracy of computation.
- Efficiency (handled using NumPy).
- Proper error handling.
- Clear and formatted output.



Top-Down Design / Modularization

The program is divided into smaller modules:

1 Main Program

Calls functions in correct order.

3 Module 2

`print_matrix()`: Print matrix neatly.

5 Module 4

`print_solution()`: Display values of unknowns.

2 Module 1

`get_matrix()`: Input rows, compute columns, input matrix values, typecasting and validation.

4 Module 3

`gauss_jordan()`: Perform elimination, handle pivot = 0, normalize rows, eliminate other rows.

This modular design improves readability, maintainability, and debugging.

Algorithm Development (Gauss–Jordan Method)

Step 1

Input number of unknowns.

Step 2

Input augmented matrix values.

Step 3

For each pivot row:

- If pivot is zero \rightarrow swap with a lower row.
- Divide the entire row by pivot to make it 1.
- For all other rows, make pivot column = 0.

Step 4

After matrix becomes RREF, last column gives solutions.

Step 5

Print variable values:

- $x_1 = A[0][-1]$.
- $x_2 = A[1][-1]$, etc.

Testing and Refinement

The program was tested with multiple input scenarios:

1

Test 1

Normal solvable system - Works correctly; gives expected solutions.

3

Test 3

Invalid numeric input - Program catches errors using try-except.

5

Test 5

Large values - Works well due to NumPy handling.

2

Test 2

Pivot zero case - Program swaps rows automatically.

4

Test 4

Floating-point numbers - Outputs correct results.

Refinements made:

1

Improved error messages.

2

Added row swapping logic.

3

Ensured pivot normalization.

7. Conclusion

This project successfully implements a robust Gauss–Jordan Linear Equation Solver. Through this project, key programming principles were applied:

Input validation

Typecasting

Conditional logic

Nested loops

Functions and modular design

Use of Python libraries

Implementation of a mathematical algorithm

The final program is efficient, modular, and user-friendly. It demonstrates both theoretical understanding and practical programming skills, meeting all academic and functional requirements.

Thank You