

Solve a System of Equations in R (3 Examples)

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November 2, 2025

RECOMMENDED CITATION

Mohammed looti (2025). *Solve a System of Equations in R (3 Examples)*.
PSYCHOLOGICAL STATISTICS. Retrieved from
<https://statistics.arabpsychology.com/?p=8384>

The Power of R for [System of Equations](#)

Solving a [system of linear equations](#) is a fundamental task in mathematics, engineering, and data science. When dealing with multiple variables and complex systems, computational tools become essential. Within the statistical programming environment of [R](#), we can utilize the highly efficient built-in function, `solve()`, designed specifically for matrix algebra operations.

This function leverages principles of [linear algebra](#) to quickly determine the unknown variables. The following guide provides a detailed breakdown of how to structure your equations as matrices and apply the `solve()` function across systems of increasing complexity, from two variables up to four.

Mathematical Foundation: Representing Linear Systems as Matrices

Before executing the solution in R, it is critical to understand the standard mathematical representation of a linear system. Any set of linear equations can be expressed in the form $\mathbf{AX} = \mathbf{B}$, where:

A is the coefficient [matrix](#) (containing the coefficients of the variables).

X is the vector of unknown variables (the solution we are seeking).

B is the constant vector (containing the values on the right-hand side of the equations).

The `solve()` function in R requires us to explicitly define the **A** matrix and the **B** vector as R objects. The solution **X** is then calculated by multiplying the inverse of **A** by **B** (i.e., $X = A^{-1}B$).

Example 1: Solving a System of Equations with Two Variables

We begin with a simple system involving two variables, x and y . This process demonstrates the core steps required for setting up and solving any linear system in [R](#).

Consider the following [system of equations](#):

$$5x + 4y = 35$$

$$2x + 6y = 36$$

To prepare this for the `solve()` function, we must extract the coefficients into the **A** matrix and the constants into the **B** vector. Notice how the coefficients are entered column-wise into the `matrix()` function in R.

```
#define left-hand side of equations
```

```
left_matrix <- matrix(c(5, 2, 4, 6), nrow=2)
```

```

left_matrix

5 4
2 6

#define right-hand side of equations
right_matrix <- matrix(c(35, 36), nrow=2)

right_matrix

35
36

#solve for x and y
solve(left_matrix, right_matrix)

3
5

```

The resulting output provides the values for the variables in the order they were defined in the coefficient matrix (A). This tells us that the value for x is **3** and the value for y is **5**. We can verify this solution by substituting these values back into the original equations.

Example 2: Expanding to a System of Equations with Three Variables

The methodology remains consistent when scaling up to higher dimensions. For a system involving three variables (x , y , and z), we will construct a 3×3 coefficient matrix (A) and a 3×1 constant vector (B).

Suppose we have the following system:

$$4x + 2y + 1z = 34$$

$$3x + 5y - 2z = 41$$

$$2x + 2y + 4z = 30$$

We define the coefficient matrix (A) by collecting the coefficients for x (column 1), y (column 2), and z (column 3). Note that coefficients of 1 or -1 must be explicitly included in the matrix definition within [R](#).

```

#define left-hand side of equations
left_matrix <- matrix(c(4, 3, 2, 2, 5, 2, 1, -2, 4), nrow=3)

```

```
left_matrix

4 2 1
3 5 -2
2 2 4

#define right-hand side of equations
right_matrix <- matrix(c(34, 41, 30), nrow=3)

right_matrix

34
41
30

#solve for x, y, and z
solve(left_matrix, right_matrix)

5
6
2
```

The computation yields the solution vector. This tells us that the value for x is **5**, the value for y is **6**, and the value for z is **2**. This example clearly illustrates the power of **solve()** in handling multivariate [linear algebra](#) problems efficiently.

Example 3: Solving a System of Equations with Four Variables

For systems involving four or more variables, manual calculation becomes highly impractical. This is where computational tools like [R](#) truly shine. In this example, we solve for w , x , y , and z , requiring a 4×4 coefficient matrix.

Suppose we have the following complex [system of equations](#):

$$6w + 2x + 2y + 1z = 37$$

$$2w + 1x + 1y + 0z = 14$$

$$3w + 2x + 2y + 4z = 28$$

$$2w + 0x + 5y + 5z = 28$$

When constructing the coefficient matrix (A), ensure that coefficients of zero (as seen for z in the

second equation and x in the fourth equation) are included as the number 0 in the vector passed to the `matrix()` function. This ensures the correct structure for the 4×4 matrix.

#define left-hand side of equations

```
left_matrix <- matrix(c(6, 2, 3, 2, 2, 1, 2, 0, 2, 1, 2, 5, 1, 0, 4, 5), nrow=4)
```

```
left_matrix
```

```
6 2 2 1
```

```
2 1 1 0
```

```
3 2 2 4
```

```
2 0 5 5
```

#define right-hand side of equations

```
right_matrix <- matrix(c(37, 14, 28, 28), nrow=4)
```

```
right_matrix
```

```
37
```

```
14
```

```
28
```

```
28
```

#solve for w, x, y and z

```
solve(left_matrix, right_matrix)
```

```
4
```

```
3
```

```
3
```

```
1
```

The resulting matrix provides the solution vector for the four variables. This tells us that the value for w is **4**, x is **3**, y is **3**, and z is **1**.

Summary and Additional Resources

The `solve()` function in [R](#) provides a robust and efficient mechanism for finding the unique solution to systems of linear equations. By correctly formatting the coefficient matrix (A) and the constant vector (B) using the `matrix()` function, users can quickly obtain solutions, regardless of the number of variables involved (as long as the system is non-singular).

To further enhance your skills in numerical computation and data manipulation using R, consider

exploring these related topics and tutorials:

How to perform matrix multiplication in R.

Understanding eigenvalue decomposition using R functions.

Techniques for handling sparse matrices in large datasets.