# Final Report SolverPro

# Universidad EAFIT Computer and Systems Department

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Project Name: SolverPro

Website URL (GitHub repository)

https://github.com/MauricioCaO7/SolverPro.git

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# User's manual

The set of instructions and important information for the user to know and understand how to navigate and use the final product is available in the proper project, section "Tutorials" with the purpose of unification.

# Installation/setup guide

Attached on repo's Readme.

# Pseudocodes of methods

## 1. Bisection

## Algorithm 1 Bisection Method with Error Handling and Iteration Limit

```
0: Input: f, a, b, tol, N_max
0: Initialize: fa \leftarrow f(a), pm \leftarrow \frac{a+b}{2}, fpm \leftarrow f(pm)
0: E \leftarrow 1000, \, cont \leftarrow 1
0: while E > \text{tol} and cont < N_{\text{-}} max do
      if fa \times fpm < 0 then
         b \leftarrow pm
0:
      else
0:
0:
         a \leftarrow pm
         fa \leftarrow fpm  {Update fa when new asigned a}
0:
      end if
0:
      p_0 \leftarrow pm
      pm \leftarrow \frac{a+b}{2}fpm \leftarrow f(pm)
0:
0:
      E \leftarrow |pm - p_0|
      cont \leftarrow cont + 1
0: end while
0: Print: "Root found at", pm, "in", cont, "iterations with an error of:", E
0: Return: pm = 0
```

# 2. False Position

## Algorithm 2 False Position Method with Error Handling and Iteration Limit

```
0: Input: f, a, b, tol, N<sub>max</sub>
0: Initialize: fa \leftarrow f(a), fb \leftarrow f(b)

0: pm \leftarrow \frac{fb \cdot a - fa \cdot b}{fb - fa}

0: fpm \leftarrow f(pm), E \leftarrow 1000, cont \leftarrow 1
0: while E > \text{tol} and cont < N_{\text{-}} max do
        if fa \times fpm < 0 then
0:
             b \leftarrow pm
0:
             fb \leftarrow fpm
0:
         {f else}
0:
             a \leftarrow pm
             fa \leftarrow fpm
0:
0:
        end if
        p_0 \leftarrow pm
0:
        pm \leftarrow \frac{fb \cdot a - fa \cdot b}{fb - fa}fpm \leftarrow f(pm)
0:
0:
         E \leftarrow |pm - p_0|
0:
         cont \leftarrow cont + 1
0:
0: end while
0: Print: "Root found at", pm, "in", cont, "iterations with an error of", E
0: Return: pm = 0
```

# 4. Gaussian Elimination

## Algorithm 3 Gaussian Elimination with Back Substitution

```
0: Input: matrixA, matrixB, n
0: Initialize: M \leftarrow \text{new Array}(n)
0: for i = 0 to n - 1 do
       M[i] \leftarrow \text{new Array}(n+1)
       for j = 0 to n - 1 do
           M[i][j] \leftarrow A[i][j]
0:
       end for
       M[i][n] \leftarrow B[i]
0:
0: end for
0: for k = 0 to n - 2 do
       \begin{array}{l} \mathbf{for} \ i = k+1 \ \mathbf{to} \ n-1 \ \mathbf{do} \\ ratio \leftarrow \frac{M[i][k]}{M[k][k]} \\ \mathbf{for} \ j = k \ \mathbf{to} \ n \ \mathbf{do} \end{array}
0:
0:
              M[i][j] \leftarrow M[i][j] - ratio \cdot M[k][j]
0:
           end for
0:
       end for
0: end for
0: X \leftarrow \text{new Array}(n)
0: for i = n - 1 to 0 do
       sum \leftarrow 0
       for j = i + 1 to n - 1 do
0:
0:
           sum \leftarrow sum + M[i][j] \cdot X[j]
       X[i] \leftarrow \frac{M[i][n] - sum}{M[i][i]}
0:
0: end for
0: Return: X
   =0
```

# 4. Partial Pivoting

#### Algorithm 4 Gaussian Elimination with Partial Pivoting

```
0: Input: A, b, n {A is the coefficient matrix, b is the constant vector, n is the number of variables}
0: Combine A and b into an augmented matrix M
0: for k = 1 to n - 1 do
     Initialize: \max \leftarrow |M[k, k]|, \max \text{Row} \leftarrow k
     for i = k + 1 to n do
0:
        if |M[i,k]| > \max then
0:
0:
           \max \leftarrow |M[i,k]|
           \text{maxRow} \leftarrow i
0:
        end if
0:
     end for
0:
     if max \approx 0 then
0:
        Throw error: "Singular matrix"
0:
0:
     Swap: M[k, *] with M[\max Row, *] {Swap rows for partial pivoting}
0:
0:
     for i = k + 1 to n do
        \text{factor} \leftarrow M[i,k]/M[k,k]
0:
        for j = k to n + 1 do
0:
           M[i,j] \leftarrow M[i,j] - \text{factor} \cdot M[k,j]  {Elimination step}
0:
        end for
0:
     end for
0:
0: end for
0: Initialize: x \leftarrow new vector of size n
0: for i = n to 1 (descending) do
     sum \leftarrow 0
     for j = i + 1 to n do
0:
        \operatorname{sum} \leftarrow \operatorname{sum} + M[i,j] \cdot x[j]
0:
0:
     x[i] \leftarrow (M[i, n+1] - \text{sum})/M[i, i] \{\text{Back-substitution}\}\
0: end for
0: Return: x = 0
```

# 5. Total Pivoting

#### Algorithm 5 Gaussian Elimination with Total Pivoting

```
0: Input: A, b, n {A es la matriz de coeficientes, b es el vector de constantes, n es el número de
   variables}
0: Combine A y b en una matriz aumentada M
0: for k = 1 to n - 1 do
     Initialize: \max \leftarrow 0
     for i = k to n do
0:
0:
        for j = k to n do
          if |M[i,j]| > \max then
0:
0:
            \max \leftarrow |M[i,j]|
            \text{maxRow} \leftarrow i
0:
            \max \text{Col} \leftarrow j
0:
          end if
0:
        end for
0:
     end for
0:
0:
     if max \approx 0 then
        Throw error: "Singular matrix"
0:
     end if
0:
     Swap: M[k, *] with M[\max Row, *] {Intercambiar files}
0:
     Swap: M[*,k] with M[*, maxCol] {Intercambiar columnas}
0:
0:
     Record the column swap for reordering the final solution
     for i = k + 1 to n do
0:
        factor \leftarrow M[i,k]/M[k,k]
0:
        for j = k to n + 1 do
0:
0:
          M[i,j] \leftarrow M[i,j] - \text{factor} \cdot M[k,j]  {Paso de eliminación}
0:
     end for
0:
0: end for
0: Initialize: x \leftarrow new vector of size n
0: for i = n to 1 (descending) do
     sum \leftarrow 0
0:
     for j = i + 1 to n do
0:
       sum \leftarrow sum + M[i, j] \cdot x[j]
0:
0:
     x[i] \leftarrow (M[i, n+1] - \text{sum})/M[i, i] {Sustitución hacia atrás}
0: end for
0: Reorder x according to the recorded column swaps {Reordenar según los intercambios de
   columnas}
0: Return: x = 0
```

#### 6. Newton's Method

#### Algorithm 6 Newton's Method with Error Handling

```
0: Entrada: f, f', x_0, \text{tol}, \text{max\_iter}
0: Inicializar: x \leftarrow x_0, iter \leftarrow 0, error \leftarrow 1
0: Verificar: Si f(x_0) o f'(x_0) no están definidos en el dominio, lanzar error
0: Verificar: Si f'(x_0) = 0, lanzar error (no se puede dividir por 0)
0: while error ; tol y iter; max_iter do
      f(x) \leftarrow \text{evaluar } f \text{ en } x
     f'(x) \leftarrow \text{evaluar } f' \text{ en } x
0:
     x_{\text{next}} \leftarrow x - \frac{f(x)}{f'(x)}
0:
      error \leftarrow |x_{\text{next}} - x|
0:
0:
      if error < tol then
         Imprimir: "Solución encontrada en la iteración", iter
0:
        Retornar: x_{\text{next}}
0:
0:
      else if f(x_{\text{next}}) = 0 then
        Imprimir: "Raíz exacta encontrada en la iteración", iter
0:
0:
        Retornar: x_{\text{next}}
      else if iter = max_iter then
0:
        Imprimir: "No se encontró solución en el número máximo de iteraciones"
0:
         Retornar: None
0:
      end if
0:
0:
      x \leftarrow x_{\text{next}}
     iter \leftarrow iter + 1
0: end while
0: Imprimir: "It was impossible to find the root within", max_iter, "iterations"
0: Retornar: None =0
```

#### 7. Fixed Point

#### Algorithm 7 Fixed Point Method with Error Handling and Iteration Limit

```
0: Entrada: f, g, x_0, \text{ tol, max\_iter}
0: Inicializar: x \leftarrow x_0, iter \leftarrow 0, error \leftarrow 1
0: Verificar: Si f(x_0) o g(x_0) no están definidos en el dominio, lanzar error
0: Verificar: Si f \equiv g, lanzar error (no pueden ser iguales)
0: \mathbf{while} error \mathbf{j} tol \mathbf{y} iter \mathbf{j} max_iter \mathbf{do}
0:
      x_{\text{next}} \leftarrow g(x)
      error \leftarrow |x_{\text{next}} - x|
0:
     if error < tol then
0:
        Imprimir: "Solución encontrada en la iteración", iter
0:
0:
        Retornar: x_{\text{next}}
     else if f(x_{\text{next}}) = 0 then
0:
        Imprimir: "Raíz exacta encontrada en la iteración", iter
0:
        Retornar: x_{\text{next}}
0:
      else if iter = max_iter then
0:
        Imprimir: "No se encontró solución en el número máximo de iteraciones"
0:
         Retornar: None
0:
     end if
0:
0:
     x \leftarrow x_{\text{next}}
     iter \leftarrow iter + 1
0:
0: end while
0: Imprimir: "It was impossible to find the root within", max_iter, "iterations"
0: Retornar: None =0
```

#### 8. Incremental searches

#### Algorithm 8 Incremental Search Method with Iteration Limit

```
0: Input: f, x_0, h, N_max
0: Initialize: x_{\text{inf}} \leftarrow x_0, x_{\text{sup}} \leftarrow x_{\text{inf}} + h
0: y_{\text{inf}} \leftarrow f(x_{\text{inf}}), y_{\text{sup}} \leftarrow f(x_{\text{sup}})
0: for i = 1 until N<sub>-</sub>max do
        if y_{\inf} \times y_{\sup} \leq 0 then
           Print: "Root at the interval between", x_{inf}, "y", x_{sup}, "in iteration", i
0:
0:
           Return: x_{inf}, x_{sup}
0:
        end if
0:
       x_{\text{inf}} \leftarrow x_{\text{sup}}
0:
       y_{\text{inf}} \leftarrow y_{\text{sup}}
       x_{\text{sup}} \leftarrow x_{\text{inf}} + h
0:
       y_{\sup} \leftarrow f(x_{\sup})
0:
0: end for
0: if i > N_{-}max then
        Print: "Could not find sign switch on maximum iterations"
        Return: None
0: \mathbf{end} \mathbf{if} = 0
```

# 9. Multiple Roots Method (Euler-Chebyshev Method)

# Algorithm 9 Multiple Roots Method

```
0: Input: f, f', f'', x_0, \epsilon, N_{max}
0: Initialize: n \leftarrow 0
0: if |f(x_0)| < \epsilon then
      Return: x_0 {x_0 is already a root}
0: while n < N_{-}max do
     f_x \leftarrow f(x_0)
     f'_x \leftarrow f'(x_0) 
 f''_x \leftarrow f''(x_0)
      x_1 \leftarrow x_0 - \frac{f_x \cdot f_x'}{(f_x')^2 - f_x \cdot f_x''} {Update x using the Euler-Chebyshev formula}
      if |x_1 - x_0| < \epsilon or |f(x_1)| < \epsilon then
         Return: x_1 {Convergence achieved}
0:
      end if
0:
      x_0 \leftarrow x_1
0:
      n \leftarrow n + 1
0: end while
0: Print: "Maximum iterations reached without convergence"
0: Return: None =0
```

#### 10. Crout's method

#### Algorithm 10 Crout's Method for LU Factorization

```
0: Input: Matrix A, vector b
0: Initialize: n \leftarrow \text{rows of } A
0: L \leftarrow I_n, U \leftarrow I_n {Identity matrices of size n}
0: Steps: Empty list to store progress
0: for i = 0 to n - 2 do
     for j = i to n - 1 do
0:
        L[j][i] \leftarrow A[j][i]
0:
        for k = 0 to i - 1 do
0:
          L[j][i] \leftarrow L[j][i] - L[j][k] \cdot U[k][i]
0:
        end for
0:
     end for
0:
     Store Step: "Matrix L updated after column i + 1"
0:
0:
     for j = i + 1 to n - 1 do
        U[i][j] \leftarrow A[i][j]
0:
        for k = 0 to i - 1 do
0:
          U[i][j] \leftarrow U[i][j] - L[i][k] \cdot U[k][j]
0:
        end for
0:
        U[i][j] \leftarrow U[i][j]/L[i][i]
0:
0:
     end for
     Store Step: "Matrix U updated after row i + 1"
0:
0: end for
0: L[n-1][n-1] \leftarrow A[n-1][n-1]
0: for k = 0 to n - 2 do
     L[n-1][n-1] \leftarrow L[n-1][n-1] - L[n-1][k] \cdot U[k][n-1]
0: end for
0: Store Step: "Final L matrix computed"
0: Solve: Ly = b using forward substitution
0: Store Step: "Intermediate solution vector y computed"
0: Solve: Ux = y using backward substitution
0: Store Step: "Solution vector x computed"
0: Return: x, L, U = 0
```

# 12. Jacobi

## Algorithm 11 Jacobi Method

Require: Matrix A of size  $n \times n$ , vector b of size n, initial guess  $x^{(0)}$ , tolerance tol, maximum

```
iterations \maxIter
Ensure: Approximate solution vector x
 1: Set k \leftarrow 0
 2: Set x^{(k)} \leftarrow x^{(0)}
 3: repeat
        Set x^{(k+1)} \leftarrow x^{(k)}
        for i = 1 to n do
 5:
            \mathrm{sum} \leftarrow 0
 6:
            for j = 1 to n do
 7:
               if j \neq i then
 8:
                  \operatorname{sum} \leftarrow \operatorname{sum} + A_{ij} \cdot x_j^{(k)}
 9:
10:
               end if
           end for x_i^{(k+1)} \leftarrow \frac{b_i - \text{sum}}{A_{ii}}
11:
12:
13:
        Compute error \leftarrow \|x^{(k+1)} - x^{(k)}\|
14:
        k \leftarrow k+1
16: until error < tol or k \ge \maxIter
17: return x^{(k)} = 0
```

## 13. Newton's Divided Differences

# Algorithm 12 Newton's Divided Differences Method

```
Require: Data points (x_0, y_0), (x_1, y_1), \dots, (x_n, y_n)
Ensure: Newton's interpolating polynomial P(x)
 1: Compute divided differences:
 2: for i = 0 to n do
       f[x_i] \leftarrow y_i
 4: end for
 5: for j = 1 to n do
       for i = n down to j do
          f[x_i, x_{i-1}, \dots, x_{i-j}] \leftarrow \frac{f[x_i, x_{i-1}, \dots, x_{i-j+1}] - f[x_{i-1}, x_{i-2}, \dots, x_{i-j}]}{x_i - x_{i-j}}
 7:
       end for
 8:
 9: end for
10: Construct the interpolating polynomial:
11: P(x) \leftarrow f[x_0]
12: for k = 1 to n do
       \text{term} \leftarrow f[x_k, x_{k-1}, \dots, x_0]
       for i = 0 to k - 1 do
          term \leftarrow term \times (x - x_i)
15:
       end for
16:
       P(x) \leftarrow P(x) + \text{term}
17:
18: end for
19: return P(x) = 0
```

#### 13. Secant

## Algorithm 13 Secant Method

```
Require: Function f(x), initial guesses x_0 and x_1, tolerance tol, maximum iterations maxIter
Ensure: Approximate root x
 1: Initialize k \leftarrow 0
 2: while k < \maxIter do
       f(x_0) \leftarrow f(x_0)
       f(x_1) \leftarrow f(x_1)
 4:
      if f(x_1) = 0 then
 5:
         return x_1
 6:
 7:
       end if
       if f(x_1) - f(x_0) = 0 then
 8:
         return Error: Division by zero
 9:
10:
      x_2 \leftarrow x_1 - f(x_1) \times \frac{(x_1 - x_0)}{f(x_1) - f(x_0)}
11:
       error \leftarrow |x_2 - x_1|
12:
       if \ \mathrm{error} < \mathrm{tol} \ \mathbf{then}
13:
14:
         return x_2
       end if
15:
16:
       x_0 \leftarrow x_1
       x_1 \leftarrow x_2
17:
       k \leftarrow k+1
18:
19: end while
20: return No root found within the maximum number of iterations =0
```

# 14. LU no pivoting

#### Algorithm 14 LU Decomposition without Pivoting

```
Require: Matrix A of size n \times n, vector b of size n
Ensure: Lower triangular matrix L, upper triangular matrix U, solution vector x
 1: Initialize L as an n \times n zero matrix
 2: Initialize U as an n \times n zero matrix
 3: for i = 1 to n do
       for k = i to n do
 5:
          sum \leftarrow 0
          for j = 1 \text{ to } i - 1 do
 6:
 7:
             \operatorname{sum} \leftarrow \operatorname{sum} + L_{ij} \times U_{jk}
          end for
 8:
          U_{ik} \leftarrow A_{ik} - \text{sum}
 9:
       end for
10:
11:
       for k = i to n do
          if i = k then
12:
13:
             L_{ik} \leftarrow 1
          else
14:
             sum \leftarrow 0
15:
16:
             for j = 1 \text{ to } i - 1 do
                \operatorname{sum} \leftarrow \operatorname{sum} + L_{kj} \times U_{ji}
17:
18:
19:
             if U_{ii} = 0 then
                return Error: Zero pivot encountered
            end if L_{ki} \leftarrow \frac{A_{ki} - \text{sum}}{U_{ii}}
21:
22:
          end if
23:
       end for
24:
25: end for
26: Forward Substitution: Solve Ly = b
27: Initialize vector y of size n
28: for i = 1 to n do
29:
       sum \leftarrow 0
       for j = 1 \text{ to } i - 1 do
30:
31:
          sum \leftarrow sum + L_{ij} \times y_j
       end for
32:
       y_i \leftarrow b_i - \text{sum}
34: end for
35: Back Substitution: Solve Ux = y
36: Initialize vector x of size n
37: for i = n downto 1 do
38:
       sum \leftarrow 0
       for j = i + 1 to n do
39:
40:
          \operatorname{sum} \leftarrow \operatorname{sum} + U_{ij} \times x_j
       end for
41:
       if U_{ii} = 0 then
42:
          return Error: Zero pivot encountered14
43:
44:
45:
                   \overline{U_{ii}}
46: end for
47: return L, U, x = 0
```

# 0.1 15. Cholesky

#### Algorithm 15 Cholesky Decomposition

```
Require: Symmetric positive definite matrix A of size n \times n, vector b of size n
Ensure: Lower triangular matrix L, solution vector x
 1: Initialize L as an n \times n zero matrix
 2: for i = 1 to n do
       for j = 1 to i do
           sum \leftarrow 0
 4:
           for k = 1 to j - 1 do
 5:
             sum \leftarrow sum + L_{ik} \times L_{jk}
 6:
 7:
           end for
          \quad \textbf{if} \quad i=j \quad \textbf{then} \quad
 8:
             if A_{ii} - \text{sum} \leq 0 then
                return Error: Matrix is not positive definite
10:
11:
             L_{ij} \leftarrow \sqrt{A_{ii} - \text{sum}}
12:
13:
             L_{ij} \leftarrow \frac{1}{L_{jj}} \times (A_{ij} - \text{sum})
14:
15:
       end for
16:
17: end for
18: Forward Substitution: Solve Ly = b
19: Initialize vector y of size n
20: for i = 1 \text{ to } n do
21:
       \mathrm{sum} \leftarrow 0
       for j = 1 \text{ to } i - 1 do
22:
          sum \leftarrow sum + L_{ij} \times y_j
23:
24:
25:
26: end for
27: Back Substitution: Solve L^T x = y
28: Initialize vector x of size n
29: for i = n downto 1 do
30:
       sum \leftarrow 0
       for j = i + 1 to n do
31:
          \operatorname{sum} \leftarrow \operatorname{sum} + L_{ii} \times x_i
32:
       end for x_i \leftarrow \frac{y_i - \text{sum}}{I}
33:
34:
35: end for
36: return L, x = 0
```

# 16. LU with partial pivoting

#### Algorithm 16 LU Decomposition with Partial Pivoting

```
0: Input: Matrix A, vector b
0: Initialize: n \leftarrow \text{rows of } A, L \leftarrow 0_{n \times n}
0: U \leftarrow A, P \leftarrow [0, 1, ..., n-1]
0: Steps: Empty list to store progress
0: for k = 0 to n - 1 do
     Find pivot: pivot \leftarrow k
     for i = k + 1 to n - 1 do
0:
        if |U[i][k]| > |U[pivot][k]| then
0:
          pivot \leftarrow i
0:
       end if
0:
     end for
0:
     if pivot \neq k then
0:
0:
       Swap rows k and pivot in U
        Swap P[k] and P[pivot]
0:
        Store Step: "Pivoting: Rows k and pivot swapped"
0:
0:
     end if
     for i = k + 1 to n - 1 do
0:
       m \leftarrow U[i][k]/U[k][k]
0:
        L[i][k] \leftarrow m
0:
        for j = k to n - 1 do
0:
0:
          U[i][j] \leftarrow U[i][j] - m \cdot U[k][j]
        end for
0:
     end for
0:
     Store Step: "Matrix U updated, and L column k computed"
0:
0: Solve: Ly = Pb using forward substitution
0: Store Step: "Solution vector y computed"
0: Solve: Ux = y using backward substitution
0: Store Step: "Solution vector x computed"
0: Return: x, L, U, P = 0
```

#### 17. Gauss-Seidel

#### Algorithm 17 Gauss-Seidel Method

```
0: Input: Matrix A, vector b, initial guess x_0, tolerance tol, maximum iterations maxIter
0: Initialize: n \leftarrow \text{rows of } A, x \leftarrow x_0, k \leftarrow 0, error \leftarrow \infty
0: Iterations: Empty list to store progress
0: while k < maxIter and error > tol do
     x_{old} \leftarrow x {Save previous iteration values}
     for i = 0 to n - 1 do
0:
        sum \leftarrow 0
0:
        for j = 0 to n - 1 do
0:
0:
          if j \neq i then
             sum \leftarrow sum + A[i][j] \cdot x[j]
0:
           end if
0:
        end for
0:
0:
        if A[i][i] = 0 then
           Throw Error: "Zero on diagonal at row i + 1. Cannot proceed."
0:
0:
        x[i] \leftarrow (b[i] - sum)/A[i][i]
0:
0:
     end for
     error \leftarrow \sqrt{\sum_{i=0}^{n-1} (x[i] - x_{old}[i])^2}
0:
     Store Iteration: "x^{(k+1)} = [x_1, \dots, x_n], Error = error"
0:
     k \leftarrow k + 1
0: end while
0: if k = maxIter and error > tol then
     Throw Error: "Maximum number of iterations reached without convergence."
0: end if
0: Return: Solution x, Iterations =0
```

#### 18. Vandermonde Method (Interpolation)

## Algorithm 18 Vandermonde Method (Interpolation)

```
0: Input: List of points (x_1, y_1), (x_2, y_2), ..., (x_n, y_n)

0: Initialize: Matrix A of size n \times n, vector b of size n

0: for i = 1 to n do

0: for j = 1 to n do

0: A[i][j] \leftarrow x_i^{n-j}

0: end for

0: b[i] \leftarrow y_i

0: end for

0: Solve: Solve the system A \cdot \text{coef} = b to get the coefficients of the polynomial

0: Return: Polynomial coefficients =0
```

# 19. Doolittle Method (LU Factorization)

#### Algorithm 19 Doolittle Method (LU Factorization)

```
0: Input: Matrix A of size n \times n
0: Initialize: Matrices L and U of size n \times n filled with zeros
0: for i = 1 to n do
0: for j = i to n do
0: U[i][j] \leftarrow A[i][j] - \sum_{k=1}^{i-1} L[i][k] \cdot U[k][j]
0: end for
0: for k = i + 1 to n do
0: L[k][i] \leftarrow \frac{A[k][i] - \sum_{j=1}^{i-1} L[k][j] \cdot U[j][i]}{U[i][i]}}{U[i][i]}
0: end for
0: end for
0: Return: Matrices L and U = 0
```

# 20. TrazCub Method (Cubic Spline Interpolation)

# Algorithm 20 TrazCub Method (Cubic Spline Interpolation)

```
0: Input: List of points (x<sub>1</sub>, y<sub>1</sub>), (x<sub>2</sub>, y<sub>2</sub>), ..., (x<sub>n</sub>, y<sub>n</sub>)
0: Initialize: Number of intervals m ← n − 1
0: Build: Construct the system for the second derivatives using boundary conditions
0: Solve: Solve the system to find the second derivatives at each point
0: for i = 1 to m do
0: Construct: For each interval [x<sub>i</sub>, x<sub>i+1</sub>], find the cubic polynomial using the second derivatives
0: end for
0: Return: Cubic spline function that interpolates the points =0
```

#### 21. Composite Trapezoidal Rule

#### Algorithm 21 Trapecio Compuesto (Composite Trapezoidal Rule)

```
0: Input: Function f(x), lower limit a, upper limit b, number of subintervals n
0: h \leftarrow \frac{b-a}{n}
0: sum \leftarrow f(a) + f(b)
0: for \ i = 1 \ to \ n - 1 \ do
0: x \leftarrow a + i \cdot h
0: sum \leftarrow sum + 2 \cdot f(x)
0: end \ for
0: Output: Integral approximation Integral \leftarrow \frac{h}{2} \cdot sum
0: Return: Integral approximation =0
```

# 22. Lagrange

#### Algorithm 22 Lagrange Interpolation Method

```
0: Input: Data points X = \{x_1, x_2, \dots, x_n\}, Y = \{y_1, y_2, \dots, y_n\}
0: Output: Coefficients of the Lagrange polynomial P(x)
0: Initialize: L \leftarrow [] (to store Lagrange basis polynomials)
0: function POLYVAL(coefficients, x)
      Return: \sum_{i=0}^{m} \text{coefficients}[i] \cdot x^{m-i}, where m is the degree of the polynomial
0: end function
0: function CONV(poly1, poly2)
      Initialize: result \leftarrow [0] of size (length of poly1 + length of poly2 - 1)
0:
0:
      for i \leftarrow 0 to length of poly1 – 1 do
         for j \leftarrow 0 to length of poly2 – 1 do
0:
            \operatorname{result}[i+j] \leftarrow \operatorname{result}[i+j] + \operatorname{poly1}[i] \cdot \operatorname{poly2}[j]
0:
         end for
0:
      end for
0:
      Return: result
0:
0: end function
0: for i \leftarrow 1 to n do
      \text{aux}0 \leftarrow X \text{ excluding } x_i
      aux \leftarrow [1, -aux0[1]] (initial polynomial for L_i(x))
      for j \leftarrow 2 to length of aux0 do
0:
         aux \leftarrow CONV(aux, [1, -aux0[j]])
0:
      end for
0:
      normalizer \leftarrow POLYVAL(aux, x_i)
      L[i] \leftarrow \text{aux/normalizer (normalize } L_i(x) \text{ so } L_i(x_i) = 1)
0: end for
0: Initialize: Coef \leftarrow [0] of size length(L[0])
0: for col \leftarrow 0 to length(L[0]) - 1 do
      Coef[col] \leftarrow \sum_{i=1}^{n} L[i][col] \cdot Y[i]
0: end for
0: Return: L (Lagrange basis polynomials), Coef (Lagrange polynomial coefficients) =0
```

# 23. Simpson

# Algorithm 23 Simpson's Rule for Numerical Integration

```
0: Input: Function f(x), lower limit a, upper limit b, number of subintervals n
0: Output: Approximation of the integral \int_a^b f(x) dx
0: Validate Inputs:
0: if n \leq 0 or n is odd then
     Error: "Number of subintervals (n) must be a positive even integer."
0:
     Exit.
0: end if
0: if a \ge b then
     Error: "The lower limit must be less than the upper limit."
     Exit.
0:
0: end if
0: Compute Step Size: h = \frac{b-a}{n}
0: Initialize: S = f(a) + f(b)
0: Iterate Over Subintervals:
0: for i = 1 to n - 1 do
     x_i = a + i \cdot h
     if i is odd then
        S \leftarrow S + 4 \cdot f(x_i)
0:
        S \leftarrow S + 2 \cdot f(x_i)
0:
     end if
0:
0: end for
0: Compute Integral: I = \frac{h}{3} \cdot S
0: Return: I = 0
```