Numerical Method Programming

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1 Introduction

Numerical methods are a very effective way to solve problems with a high degree of complexity, which analytically would not have a solution. Also, thanks to the exponential growth in computing power in computers, it is possible for us to implement these numerical methods to obtain better results. As a final project of the numerical analysis subject, we have developed some numerical methods for solving different problems in the Python programming language.

2 How to use the program

The first step to use the program is to download the funete code from the repository. In order to run our program, you have to do the following process: Having Python installed, the version used is Python3, along with some additional libraries (Tkinter, numpy, sympy). What we will do is run the following command in the project directory:

Python ./Principal.py principal being the class with the initial interface.

The next thing we will see will be a screen with 3 categories of methods, we will see a list of buttons with the name of the corresponding method, what follows will be to click on the method we want to execute, a sub-window will appear which will have the necessary fields for calculate method and a button to calculate, the values will come out in the same interface.

Second option: the buttons of the main interface will be linked to their respective code on the Replit page, for later execution, each method will have input variables which we will change at will and execute the corresponding operations, the results appearing on the console.

3 Methods developed

3.1 Numerical solution of nonlinear equations

• Incremental searches

program that finds an interval where f(x) has a change of sign using the method of incremental searches.

Pseudocode:

```
input:
f, continuous function
x0, starting point
h, step
Nmax, maximum number of iterations
output:
a, left end of the interval
b, right end of interval
iter, number of iterations
//function
```

```
Incremental searches(f, x0, h, Nmax):
      // Initialization
      xant = x0
      fant = f (xant)
      xact = xant + h
      fact = f(xact)
      //Cycle
      for i = 1: Nmax
           if fant * fact <0
               break
           end if
           xant = xact
           fantastic = fact
           xact = xant + h
           fact = f(xact)
      end for
      // Delivery of results
      a = xant
      b = xact
      iter = i
• Bisection
 program that finds the solution to the equation f(x) = 0 in the interval [a, b] using the bisection
 method.
 Pseudocode:
 input:
 f, continuous function
 a, right end of the initial interval
 b, end end of end interval
 tol, tolerance
 Nmax, maximum number of iterations
 output:
 x, solution
 iter, number of iterations
 err, error
 //function
 bisection (f, a, b, tol, Nmax):
      // Initialization
      fa = f(a)
      pm = (a + b) / 2
      fpm = f (pm)
      E = 1000;
      cont = 1
      //Cycle
      while E> tol && cont <Nmax
         if fa * fpm <0
            b = pm;
```

else

a = pm

```
end if
  p0 = pm
  pm = (a + b) / 2
   fpm = f (pm)
  E = abs (pm-p0)
   cont = cont + 1
end while
// Delivery of results
x = pm
iter = cont
err = E
```

• False rule

program that finds the solution to the equation f(x) = 0 in the interval [a, b] using the false rule method.

```
Pseudocode:
Input:
f, continuous function
a, right end of the initial interval
b, end end of end interval
tol, tolerance
Nmax, maximum number of iterations
Output:
x, solution
iter, number of iterations
err, error
//function
false rule (f, a, b, tol, Nmax):
    // Initialization
    fa = f(a)
    fb = f(b)
    pm = (fb * a-fa * b) / (fb-fa)
    fpm = f (pm)
    E = 1000
    cont = 1
    //Cycle
    while E> tol && cont <Nmax
       if fa * fpm <0
          b = pm
       else
          a = pm
       end if
       p0 = pm
       pm = (f (b) * a-f (a) * b) / (f (b) -f (a))
       fpm = f (pm)
       E = abs (pm-p0)
       cont = cont + 1
    end while
    // Delivery of results
    x = pm
```

```
iter = cont
err = E
```

• Fixed point

program that finds the solution to the equation f(x) = 0 by solving the analogous problem x = g(x) using the fixed point method. Pseudocode:

```
Input:
f, continuous function
g, continuous function
x0, initial approximation
tol, tolerance
Nmax, maximum number of iterations
output:
x, solution
iter, number of iterations
err, error
fixedpoint (g, x0, tol, Nmax):
    // Initialization
    xant = x0
   E = 1000
    cont = 0
    //Cycle
    while E> tol && cont <Nmax
       xact = g (xant)
       E = abs (xact-xant)
       cont = cont + 1
       xant = xact
    end while
    // Delivery of results
    x = xact
    iter = cont
    err = E
```

• Newton

err, error

program that finds the solution to the equation f(x) = 0 using Newton's method. Pseudocode:

```
Input:
f, continuous function
f ', continuous function
x0, initial approximation
tol, tolerance
Nmax, maximum number of iterations
output:
x, solution
iter, number of iterations
```

```
//function
 newton (f, df, x0, tol, Nmax):
     // Initialization
     xant = x0
     fant = f (xant)
     E = 1000
     cont = 0
     //Cycle
     while E> tol && cont <Nmax
        xact = xant-fant / (df (xant))
        fact = f (xact)
        E = abs (xact-xant)
        cont = cont + 1
        xant = xact
        fantastic = fact
     end while
     // Delivery of results
     x = xact
     iter = cont
     err = E
• Secant
 program that finds the solution to the equation f(x) = 0 using the secant method.
 Pseudocode:
 Input:
 f, continuous function
 x0, initial approximation
 x1, initial approximation
 tol, tolerance
 Nmax, maximum number of iterations
 Output:
 x, solution
 iter, number of iterations
 err, error
 secant (f, x0, x1, tol, Nmax):
     // Initialization
     f0 = f(x0)
     f1 = f(x1)
     E = 1000
     cont = 1
     //Cycle
     while E> tol && cont <Nmax
        xact = x1-f1 * (x1-x0) / (f1-f0)
        fact = f(xact)
        E = abs (xact-x1)
         cont = cont + 1
        x0 = x1
        f0 = f1
        x1 = xact
```

```
f1 = fact
end while

// Delivery of results
x = xact
iter = cont
err = E
```

• Multiple roots

program that finds the solution to the equation f(x) = 0 using the multiple roots method. Pseudocode:

```
Input:
f, continuous function
f ', continuous function
f '', continuous function
x0, initial approximation
tol, tolerance
Nmax, maximum number of iterations
Output:
x, solution
iter, number of iterations
err, error
multiple roots (f, df, d2f, x0, tol, Nmax):
    // Initialization
    xant = x0
    fant = f(xant)
    E = 1000
    cont = 0
    //Cycle
    while E> tol && cont <Nmax
       xact = xant-fant * df (xant) / ((df (xant)) ^ 2-fant * d2f (xant))
       fact = f(xact)
       E = abs (xact-xant)
       cont = cont + 1
       xant = xact
       fantastic = fact
    end while
    // Delivery of results
    x = xact
    iter = cont
    err = E
```

3.2 Solution of systems of linear equations

• Gauss

program that finds the solution to the system Ax = b using the Gaussian elimination method. Pseudocode:

```
Input:
```

A, invertible matrix

```
b, constant vector
 Output:
 x, solution
 //function
 gausspl (A, b):
     // Initialization
     n = size (A, 1)
     M = [A b]
     We reduce the system
     for i = 1: n-1
           for j = i + 1: n
               if M (j, i) = 0
                  M(j, i: n + 1) = M(j, i: n + 1) - (M(j, i) / M(i, i)) * M(i, i: n + 1)
           end for
     end for
     // Delivery of results
     x = subst (M); // Backward substitution
• LU factorization
 Program that finds the solution to the system Ax = b and the LU factorization of A using the
 LU factorization method with simple Gaussian elimination.
 Pseudocode:
 Input:
 A, invertible matrix
 b, constant vector
 Output:
 x, solution
 L, factorization matrix L
 U, U matrix of factorization
 lu (A, b):
     // Initialization
     n = size (A, 1)
     L = eye (n)
     U = zeros (n)
     M = A
     //Factoring
     for i = 1: n-1
           for j = i + 1: n
               if M (j, i) ~= 0
                  L(j, i) = M(j, i) / M(i, i)
                  M(j, i: n) = M(j, i: n) - (M(j, i) / M(i, i)) * M(i, i: n)
               end if
           end for
           U (i, i: n) = M (i, i: n)
           U (i + 1, i + 1: n) = M (i + 1, i + 1: n)
      end for
```

```
U (n, n) = M (n, n)

// Delivery of results
z = subs ([L b])
x = subs ([U z])
```

Doolitle

Program that finds the solution to the system Ax = b and the LU factorization of A using the Doolitle method.

Pseudocode:

```
Input:
A, invertible matrix
b, constant vector
Output:
x, solution
L, factorization matrix L
U, U matrix of factorization
//function
Doolittle (A, b):
    // Initialization
   n = size (A, 1)
   L = eye (n)
    U = eye (n)
    //Factoring
    for i = 1: n-1
         for j = i: n
            U(i, j) = A(i, j) - dot(L(i, 1: i-1), U(1: i-1, j)')
         end for
         for j = i + 1: n
            L(j, i) = (A(j, i) - dot(L(j, 1: i-1), U(1: i-1, i))) / U(i, i)
         end for
    U(n, n) = A(n, n) - dot(L(n, 1: n-1), U(1: n-1, n))
    // Delivery of results
    z = nounprgr ([L b])
    x = subregr ([U z])
```

• Crout

Program that finds the solution to the system Ax = b and the LU factorization of A using the Crout method.

Pseudocode:

```
Input:
```

A, invertible matrix b, constant vector

Output:

x, solution

L, factorization matrix L

U, U matrix of factorization

```
Crout (A, b):
     // Initialization
     n = size (A, 1)
     L = eye (n)
     U = eye (n)
     //Factoring
     for i = 1: n-1
           for j = i: n
              L(j, i) = A(j, i) - dot(L(j, 1: i-1), U(1: i-1, i))
           end for
           for j = i + 1: n
              U(i, j) = (A(i, j) - dot(L(i, 1: i-1), U(1: i-1, j))) / L(i, i)
           end for
      end for
     L(n, n) = A(n, n) - dot(L(n, 1: n-1), U(1: n-1, n))
     // Delivery of results
     z = nounprgr ([L b])
     x = subregr ([U z])

    Cholesky

  Program that finds the solution to the system Ax = b and the LU factorization of A using the
  Cholesky method.
 Pseudocode:
 Input:
  A, invertible matrix
 b, constant vector
 Output:
 x, solution
 L, factorization matrix L
 U, U matrix of factorization
  //function
 Cholesky (A, b):
     // Initialization
     n = size (A, 1)
     L = eye (n)
     U = eye (n)
     //Factoring
     for i = 1: n-1
          L(i, i) = sqrt(A(i, i) - dot(L(i, 1: i-1), U(1: i-1, i)))
          U(i, i) = L(i, i)
          for j = i + 1: n
              L(j, i) = (A(j, i) - dot(L(j, 1: i-1), U(1: i-1, i))) / U(i, i)
           end for
           for j = i + 1: n
              U(i, j) = (A(i, j) - dot(L(i, 1: i-1), U(1: i-1, j))) / L(i, i)
           end for
```

//function

end for

```
L (n, n) = sqrt (A (n, n) -dot (L (n, 1: n-1), U (1: n-1, n) '))
U (n, n) = L (n, n)

// Delivery of results
z = nounprgr ([L b])
x = subregr ([U z])
```

• Jacobi

Program that finds the solution to the system Ax = b using the Jacobi method. Pseudocode:

```
Input:
A, invertible matrix
b, constant vector
x0, initial approximation
tol, tolerance
Nmax, maximum number of iterations
Output:
x, solution
iter, number of iterations
err, error
jacobi (A, b, x0, tol, Nmax):
    // Initialization
    D = diag (diag (A))
    L = -tril(A) + D
    U = -triu (A) + D
    T = inv (D) * (L + U)
    C = inv (D) * b
    xant = x0
    E = 1000
    cont = 0
    //Cycle
    while E> tol && cont <Nmax
         xact = T * xant + C
         E = norm (xant-xact)
         xant = xact
         cont = cont + 1
    end while
    // Delivery of results
    x = xact
    iter = cont
    err = E
```

• Gauss-Seidel Program that finds the solution to the system Ax = b using the Gauss-Seidel method.

Pseudocode:

```
Input:
```

```
A, invertible matrix
b, constant vector
xO, initial approximation
```

```
tol, tolerance
Nmax, maximum number of iterations
Output:
x, solution
iter, number of iterations
err, error
gseidel (A, b, x0, tol, Nmax):
    // Initialization
    D = diag (diag (A))
   L = -tril(A) + D
    U = -triu(A) + D
    T = inv (D-L) * U
    C = inv (D-L) * b
    xant = x0
    E = 1000
    cont = 0
    //Cycle
    while E> tol && cont <Nmax
         xact = T * xant + C
         E = norm (xant-xact)
         xant = xact
         cont = cont + 1
    end while
    // Delivery of results
    x = xact
    iter = cont
    err = E
```

3.3 Interpolation

• Vandermonde

Program that finds the interpolating polynomial of the given data using the Vandermonde method.

Pseudocode:

```
Input:
X, abscissa
And, ordered

Output:
Coef, coefficients of the polynomial
vandermonde (X, Y):

    // Initialization
    n = length (X)
    A = zeros (n)

    //Cycle
    for i = 1: n
        A (:, i) = (X. ^ (N-i)) '
    end for
```

```
// Delivery of results
Coef = A \ Y '
```

• Divided differences

Program that finds the interpolating polynomial of the given data using the Divided differences method.

Pseudocode:

```
Input:
X, abscissa
And, ordered
Output:
Coef, coefficients of Newton's polynomial
differences (X, Y):
    // Initialization
   n = length(X)
    A = zeros (n)
    //Cycle
   D (:, 1) = Y ,
    for i = 2: n
         aux0 = D (i-1: n, i-1)
         aux = diff (aux0)
         aux2 = X (i: n) -X (1: n-i + 1)
         D (i: n, i) = aux / aux2
    end for
    // Delivery of results
    Coef = diag (D) '
```

• Lagrange

Program that finds the interpolating polynomial of the given data using the Lagrange method. Pseudocode:

```
Input:
X, abscissa
And, ordered

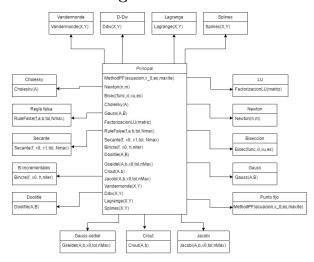
Output:
L, Lagrange polynomials
Coef, coefficients of the interpolation polynomial

lagrange (X, Y):
    // Initialization
    n = length (X)
    L = zeros (n)

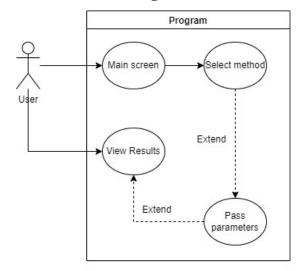
    //Cycle
    for i = 1: n
        aux0 = setdiff (X, X (i))
        aux = [1 -aux0 (1)]
```

4 Diagrams

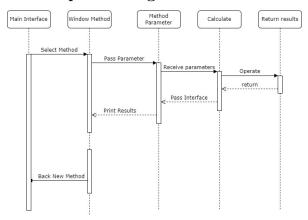
4.1 class diagram



4.2 use case diagram



4.3 sequence diagram



5 Conclusions

To carry out this project we have chosen the Python programming language, since it has different benefits compared to other programming languages. the first one is the simplicity in some aspects of the code, since this is a language interpreted with dynamic types. the second is the ease of working with mathematical functions using specialized libraries. and the third reason is the large amount of documentation that exists about the different numerical methods implemented in Python. the limitations or problems we had were related to the graphical interface, since there was no experience in the team to work with it and an accelerated learning curve was required to carry out the project.