What should be included: Answers to all questions, as well as the proof that Jacobi converges for this problem, the Jacobi-method code, and the plot showing the unsatisfactory solution for few iterations with N=5.

Lab 5 - Part 3

Implement a python function for the Jacobi-method in your code from Lab 5, part 2. The function should take the matrix A the right-hand side, an integer number being the number of iterations (note, we are not using a more advanced stopping criteria here), and an initial guess L_0 and return the solution L_k . Hints: You can create the diagonal matrix D like this D = np.diagflat(np.diag(A)). Other useful python functions are numpy.tril(), numpy.triu() and numpy.matmul(). Note that numpy.tril() and numpy.triu() includes the diagnoal of the original matrix, so for your purposes you need to subtract D. Include the code in your hand in.

Use your function for solving the ODE-BVP from Lab 5 part 2, for N=5 and a maximum number of iterations equal to 100. Make sure that the result is close to the analytical solution (e.g. by looking at your plot).

Decrease the number of Jacobi iterations until you can see in the plot that the solution is less accurate. How few iterations did you use? Include the plot in your hand in.

Set N=10 instead. Do you need more iterations for the solution to be acceptable? How many? Note: It is possible to prove that the number of iterations depend on ${\cal N}$

We have according to the method:

- ullet M=D where D is the diagonal matrix of A, so simply fetch the diagonal of A
- ullet K=L+U where L is the lower triangular part of A and U is the upper triangular part of A
- ullet -L,-U strictly lower and upper triangular parts of A respectively Can then write

$$x_k = D^{-1}(L+U)x_{k-1} + D^{-1}b$$

The function will take A, b, N and L_0 as input and return the solution L_k after N iterations.

We have

08/05/2025, 15:49 p

```
A = egin{bmatrix} 2 & -1 & 1 \ 1 & -2 & 1 \ 2 & 1 & -4 \end{bmatrix}
```

```
In [177... import numpy as np
          import matplotlib.pyplot as plt
          from math import pi
          import time
          from scipy.linalg import lu_factor, lu_solve
          def assemble_A(N):
              h = (pi / 2) / (N + 1)
              A = np.zeros((N, N)) # Coefficient matrix of NxN dimension
              # main diag
              np.fill_diagonal(A, -2 + h**2)
              # sub diag, super diag
              \# A[i, i-1] = 1 \text{ for } i = 1 \text{ to } N-1
              A[1:, :-1][np.eye(N-1, dtype=bool)] = 1
              \# A[i, i+1] = 1 \text{ for } i = 0 \text{ to } N-2
              A[:-1, 1:][np.eye(N-1, dtype=bool)] = 1
              return A
          def assemble_F(N, leftbc, rightbc):
              F = np.zeros(N)
              F[0] = -leftbc
              F[N - 1] = -rightbc
              return F
In [178... def jacobi_method(A, b, L0, max_iterations):
              D = np.diagflat(np.diag(A))
              D_inv = np.linalg.inv(D)
              L = np.tril(D - A) # need to get rid of main diagonal
              U = np.triu(D - A) # same thing
              for i in range(max_iterations):
                  L_next = D_inv @ (L + U) @ L0 + D_inv @ b
                  L = L_next
              return L
In [179... def ComputeAnalyticalSolution(N, leftbc, rightbc):
              x = np.linspace(0, (pi / 2), N + 2)
              y = np.cos(x) + 2* np.sin(x)
              return x, y
In [180... def ComputeNumericalSolution(N, leftbc, rightbc):
              h = (pi / 2) / (N + 1)
              x = np.linspace(0, (pi / 2), N + 2)
              A = np.zeros((N, N)) # Coefficient matrix of nxn dimension
```

08/05/2025, 15:49 part3

```
F = np.zeros(N) # The result vector such that AU = F of size n
# make initial a matrix of ones
L0 = np.ones(N)
A[0, 0] = -2 + h**2
A[0, 1] = 1
F[0] = -leftbc
for i in range(1, N - 1):
    A[i, i-1] = 1
    A[i, i] = -2 + h**2
    A[i, i + 1] = 1
    F[i] = 0
A[N - 1, N - 2] = 1
A[N - 1, N - 1] = -2 + h**2
F[N - 1] = -rightbc
y_h_{int} = jacobi_method(A, F, L0, 100)
y_h = np.zeros(N + 2)
y_h[0] = leftbc
y_h[1:N + 1] = y_h_int
y_h[-1] = rightbc
return x, y_h, h
```

```
In []: N = 5
    x = np.linspace(0, (pi / 2), N + 2)
    leftbc = 1
    rightbc = 2
    y = np.cos(x) + 2* np.sin(x)

    x_h, y_h, h = ComputeNumericalSolution(N, leftbc, rightbc)

# plot analytical and numerical
    plt.plot(x, y, label='Analytical Solution', color='blue')
    plt.plot(x_h, y_h, label='Numerical Solution', color='red')
    plt.title('Numerical (Jacobi) Solution vs Analytical Solution')
    plt.xlabel('x')
    plt.ylabel('y')
    plt.grid()
    plt.legend()
    plt.show()
```

08/05/2025, 15:49 part3

