FYS 3150 Project 1

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Github browser-link to this repository: https://github.com/zmbdr/FYS3150/tree/main/project1

Problem 1

Given

$$u(x) = 1 - (1 - e^{-10})x - e^{-10x}$$

we find that

$$u'(x) = (1 - e^{-10}) - (-10)e^{-10x},$$

and

$$u''(x) = -100e^{-10x}.$$

With the supplied source term f(c) given by

$$f(x) = 100e^{-10x},$$

we get

$$u''(x) = -f(x).$$

The above is valid for $x \in [0, 1]$, and the boundary conditions u(0) = u(1) = 0 are satisfied. Hence, u(x) is an exact solution to the given problem.

Problem 2

A simple c++ script to dump $u(x_i)$ for x=ih is supplied in the file main_2.cpp. The plot in Figure 1 is generated by the file plot_2.py.

Problem 3

It is well established that an approximation to u''(x) is given as follows:

$$u''(x) \approx = \frac{u(x+h) - 2u(x) + u(x-h)}{h^2},$$

where h is an appropriately chosen small step-size. For the problem at hand, we are interested in approximating u'' over the unit interval. We define $x_i = i/N$, discretizing the unit interval into N non-overlapping intervals, and let i = 0 ... N. Furthermore, we introduce $u(x_i) = u_i$, and $f(x_i) = f_i$.

Thus, from the original Poisson equation, we can write

$$-\frac{u_{i+1}-2u_i+u_{i-1}}{h^2} + \text{higher order terms} = f_i, \quad i=1,\ldots N-1.$$

This leads to the simple discrete approximation v_i :

$$-(v_{i+1}-2v_i+v_{i-1})=h^2f_i, \quad i=1,\ldots N-1.$$

Here, $v_0 = u_0 = u(0) = 0$, $v_N = u_N = u(1) = 0$, and h = 1/N. This is the discrete version of the original Poisson equation, and if this is solvable (it is!), we have found approximation to u(x) on the discrete grid x_i .

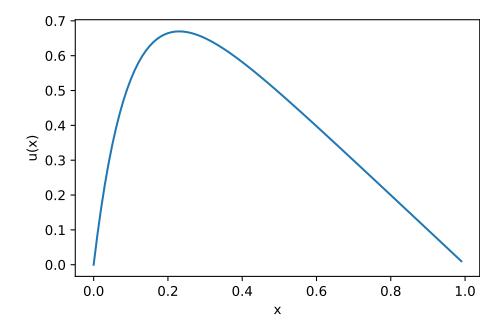


FIG. 1: The exact solution u(x) to the Poisson equation with the given source term over its domain.

Problem 4

The discrete equation above can be written as follows for i = 2, ..., N-2:

$$\begin{bmatrix} -1 & 2 & -1 \end{bmatrix} \begin{bmatrix} v_{i-1} \\ v_i \\ v_{i+1} \end{bmatrix} = h^2 f_i.$$

However, we need to take care for the cases i = 1 and i = N - 1:

$$\begin{bmatrix} 2 & -1 \end{bmatrix} \begin{bmatrix} v_i \\ v_{i+1} \end{bmatrix} - v_{i-1} = h^2 f_i, \text{ for } i = 1,$$

and,

$$\begin{bmatrix} -1 & 2 \end{bmatrix} \begin{bmatrix} v_{i-1} \\ v_i \end{bmatrix} - v_{N+1} = h^2 f_i, \text{ for } i = N-1.$$

Stacking these three equations in matrix form, we get

$$\begin{bmatrix} 2 & -1 & \dots & 0 & 0 \\ -1 & 2 & -1 & \dots & 0 \\ 0 & -1 & \ddots & \ddots & 0 \\ \vdots & \ddots & \ddots & \ddots & 0 \\ 0 & \dots & \ddots & -1 & 2 \end{bmatrix}_{(N-1)\times(N-1)} \begin{bmatrix} v_1 \\ \vdots \\ \vdots \\ v_{N-1} \end{bmatrix}_{(N-1)\times 1} - \begin{bmatrix} v_0 \\ 0 \\ \vdots \\ 0 \\ v_N \end{bmatrix}_{(N-1)\times 1} = h^2 \begin{bmatrix} f_1 \\ \vdots \\ \vdots \\ f_{N-1} \end{bmatrix}_{(N-1)\times 1}.$$

This is what we want; we get

$$\mathbf{A}\vec{v} = \vec{g},\tag{1}$$

where

$$g_i = \begin{cases} h^2 f_i + v_0 & \text{for } i = 1\\ h^2 f_i & \text{for } i = 2 \dots N - 2\\ h^2 f_i + v_N & \text{for } i = N - 1 \end{cases}$$
 (2)

Problem 5

A. Relation n to m

Now, we let $\vec{v^*}$ be a complete solution of the discretized Poisson equation. If we employ N intervals, the full solution need to be specified for N+1 values. Hence, if $\vec{v^*}$ has length m, it is clear that m=N+1. As described under Problem 4, the matrix \mathbf{A} has dimensions $(N-1)\times N-1$, which is $n\times n$. Thus, n=m+2.

B. Relation $\vec{v^*}$ to \vec{v}

From the discussion above, it appears that solving 1 gives the *interior solution* of the discrete Poisson equation, i.e. not including the boundary conditions.

Problem 6

OK, now **A** is an $n \times n$ tridiagonal matrix, and we are interested in solving $\mathbf{A}\vec{v} = \vec{g}$ for \vec{v} .

a) General algorithm

A general algorithm follows from standard Gaussian elimination forward for eliminating the subdiagonal a, and then solving backwards to find v. This is summarized here for algorithm 1.

Algorithm 1 Thomas algorithm (Gaussian elimination applied to tridiagonal matrices)

```
Require: a = [a_2, \dots a_n], b = [b_1, \dots b_n], c = [c_1, \dots c_{n-1}], g = [g_1, \dots g_n]

Allocate memory \tilde{b} = \begin{bmatrix} \tilde{b}_1, \dots \tilde{b}_n \end{bmatrix}, \tilde{g} = [\tilde{g}_1, \dots \tilde{g}_n], v = [v_1, \dots v_n]

Initialize new diagonal: \tilde{b}_1 = b_1

for i = 2, \dots, n do

\tilde{b}_i = b_i - \frac{a_i}{\tilde{b}_{i-1}} c_i
\tilde{g}_i = g_i - \frac{a_i}{\tilde{b}_{i-1}} \tilde{g}_i

v_n = \frac{\tilde{g}_n}{\tilde{b}_n}

for i = n - 1, \dots, 1 do

v_i = \frac{\tilde{g}_1 - c_i v_{i+1}}{\tilde{b}_i}

return v
```

b) Number of FLOPs

From the algorithm above, we see that the number of FLOPS equals $(n-1) \times (4+4) + 1 + (n-1) \times 4 = 12 \times (n-1) + 1$. In other words, solving a triangular system of n linear equations is requires $\approx 12n$ FLOPs.