

# **Exercise No. 5**

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## 2 - Tridiagonal matrices

1)

Due to the high amount of zeros in the matrix the Gaußian elimination can be simplified to

$$\text{for } i = 2 \dots n \begin{cases} a_i = \frac{a_i}{b_{i-1}} \\ b_i = b_i - a_i \cdot c_{i-1} \\ r_i = r_i - a_i \cdot r_{i-1} \end{cases} \quad (1)$$

However,  $b_1$  and  $c_1$  will not be changed.

2)

Now we can use the new found values of  $b_i$  and  $r_i$  to do a backward substitution. We start the process with the last row, by which the entry of the solution vector is

$$x_n = \frac{r_n}{b_n} \quad (2)$$

For the other entries we obtain with a recursion formula

$$x_i = \frac{r_i - c_i \cdot x_{i+1}}{b_i} \quad (3)$$

3)

To write the code we use the formula above and implement it as a numerical subroutine.

Listing 1: Exercise05.py

```
1 import numpy as np
2
3
4 def solve_tridiagonal(a, b, c, r):
5     #define length of a as equations we need to calculate
6     numberEquation = len(a)
7
8     #loop for gaussian elimination
9     for i in range(1, numberEquation):
10         a[i] = a[i] / b[i - 1]
11         b[i] = b[i] - a[i] * c[i - 1]
12         r[i] = r[i] - a[i] * r[i - 1]
13
14     x = np.zeros(len(b))
15
16     #backward substitution
17     x[numberEquation - 1] = r[numberEquation - 1] / b[numberEquation - 1]
18     for i in range(numberEquation - 2, -1, -1):
19         x[i] = (r[i] - c[i] * x[i + 1]) / b[i]
20
21     return x
```

4)

To test the algorithm we set a matrix with the values to all  $a = -1$ , all  $b = 2$ , all  $c = -1$  and all  $r = 0.1$ .

Listing 2: Exercise05.py

```

24 #define the matrix as give in the sheet
25 a = np.zeros(9)
26 b = np.zeros(10)
27 c = np.zeros(9)
28 r = np.zeros(10)
29
30 for i in range(10):
31     b[i] = 2
32     r[i] = 0.1
33 for i in range(9):
34     a[i] = -1
35     c[i] = -1
36
37 aa = np.append([0], [a])
38 ca = np.append([c], [0])
39
40 #copy array values
41 ac, bc, cc, rc = map(np.array, (aa, b, ca, r))
42
43 x = solve_tridiagonal(ac, bc, cc, rc)
44 print(x)

```

For the vector  $x$  we obtain

$$x = \begin{pmatrix} 0.5 \\ 0.9 \\ 1.2 \\ 1.4 \\ 1.5 \\ 1.5 \\ 1.4 \\ 1.2 \\ 0.9 \\ 0.5 \end{pmatrix}$$

0.1 5)

We put the solution back into the original matrix equation to check if the algorithm works correctly. Therefore we use the equations below for a numerical subroutine.

$$a_1 \cdot x_n + b_1 \cdot x_1 + c_1 \cdot x_2 = r_1, \quad (4)$$

$$i = 2, \dots, n-1 \quad a_i \cdot x_{i-1} + b_i \cdot x_i + c_i \cdot x_{i+1} = r_i, \quad (5)$$

$$a_n \cdot x_{n-1} + b_n \cdot x_n + c_n \cdot x_1 = r_n \quad (6)$$

Listing 3: Exercise05.py

```

46 #check if the solution is correct
47 R = np.zeros(10)
48 R[0] = b[0] * x[0] + c[0] * x[1]
49 R[9] = a[-1] * x[-2] + b[9] * x[9]
50
51 for i in range(1, 9):
52     R[i] = a[i] * x[i - 1] + b[i] * x[i] + c[i] * x[i + 1]
53 print(R)

```

For the vector  $r$  we obtain

$$R = \begin{pmatrix} 0.1 \\ 0.1 \\ 0.1 \\ 0.1 \\ 0.1 \\ 0.1 \\ 0.1 \\ 0.1 \\ 0.1 \\ 0.1 \end{pmatrix}$$

We get the same solution we put into our algorithm,  $R = r$ , therefore is no deviation visible.