

Zadanie 2 – rozwiązanie równania $\mathbf{Ax} = \mathbf{e}$ metodami Gaussa-Seidela i gradientów sprzężonych.

$$\mathbf{A} = \begin{bmatrix} 4 & 1 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & \dots \\ 1 & 4 & 1 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & \dots \\ 0 & 1 & 4 & 1 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & \dots \\ 0 & 0 & 1 & 4 & 1 & 0 & 0 & 1 & 0 & 0 & 0 & \dots \\ 1 & 0 & 0 & 1 & 4 & 1 & 0 & 0 & 1 & 0 & 0 & \dots \\ 0 & 1 & 0 & 0 & 1 & 4 & 1 & 0 & 0 & 1 & 0 & \dots \\ 0 & 0 & 1 & 0 & 0 & 1 & 4 & 1 & 0 & 0 & 1 & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ \dots & 0 & 0 & 1 & 0 & 0 & 1 & 4 & 1 & 0 & 0 & 1 \\ \dots & 0 & 0 & 0 & 1 & 0 & 0 & 1 & 4 & 1 & 0 & 0 \\ \dots & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 1 & 4 & 1 & 0 \\ \dots & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 1 & 4 & 1 \\ \dots & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 1 & 4 \end{bmatrix}$$

\mathbf{A} ma wymiary 128x128, natomiast \mathbf{e} jest wektorem o wszystkich składowych równych 1.

W metodzie Gaussa-Seidela wartości wyliczam jako:

$$x_i^{(k+1)} = \left(b_i - \sum_{j=1}^{i-1} a_{ij}x_j^{(k+1)} - \sum_{j=i+1}^N a_{ij}x_j^{(k)} \right) / a_{ii}$$

Natomiast metoda gradientów sprzężonych przebiega zgodnie z następującym algorytmem:

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 $\mathbf{r}_0 := \mathbf{b} - \mathbf{Ax}_0$ 
if  $\mathbf{r}_0$  is sufficiently small, then return  $\mathbf{x}_0$  as the result
 $\mathbf{p}_0 := \mathbf{r}_0$ 
 $k := 0$ 
repeat
     $\alpha_k := \frac{\mathbf{r}_k^\top \mathbf{r}_k}{\mathbf{p}_k^\top \mathbf{Ap}_k}$ 
     $\mathbf{x}_{k+1} := \mathbf{x}_k + \alpha_k \mathbf{p}_k$ 
     $\mathbf{r}_{k+1} := \mathbf{r}_k - \alpha_k \mathbf{Ap}_k$ 
    if  $\mathbf{r}_{k+1}$  is sufficiently small, then exit loop
     $\beta_k := \frac{\mathbf{r}_{k+1}^\top \mathbf{r}_{k+1}}{\mathbf{r}_k^\top \mathbf{r}_k}$ 
     $\mathbf{p}_{k+1} := \mathbf{r}_{k+1} + \beta_k \mathbf{p}_k$ 
     $k := k + 1$ 
end repeat
return  $\mathbf{x}_{k+1}$  as the result
    
```

W celu wykonania tego zadania napisałam dwa programy w języku C++.

W lewej kolumnie tabeli przedstawiam wyniki dla metody Gaussa-Seidela, natomiast w prawej kolumnie dla metody gradientów sprzężonych.

Metoda Gaussa-Seidela	Metoda gradientów sprzężonych
x1 = 0.198291	x1 = 0.19986
x2 = 0.123737	x2 = 0.121552
x3 = 0.160337	x3 = 0.162844
x4 = 0.132176	x4 = 0.129411
x5 = 0.171622	x5 = 0.174738
x6 = 0.0828655	x6 = 0.0790087
x7 = 0.146687	x7 = 0.15109
x8 = 0.102453	x8 = 0.0976596
x9 = 0.139647	x9 = 0.144773
x10 = 0.09812	x10 = 0.0926282
x11 = 0.152323	x11 = 0.158277
x12 = 0.103786	x12 = 0.0974205
x13 = 0.143951	x13 = 0.150655
x14 = 0.108208	x14 = 0.101208
x15 = 0.144405	x15 = 0.151699
x16 = 0.105436	x16 = 0.0978345
x17 = 0.142412	x17 = 0.150295
x18 = 0.109216	x18 = 0.101092
x19 = 0.139707	x19 = 0.14804
x20 = 0.110059	x20 = 0.101533
x21 = 0.139681	x21 = 0.148389
x22 = 0.111337	x22 = 0.102471
x23 = 0.137644	x23 = 0.146641
x24 = 0.113105	x24 = 0.104005
x25 = 0.136558	x25 = 0.145741
x26 = 0.113855	x26 = 0.104607
x27 = 0.135501	x27 = 0.144791
x28 = 0.115177	x28 = 0.105868
x29 = 0.134252	x29 = 0.143557
x30 = 0.116171	x30 = 0.106891
x31 = 0.133414	x31 = 0.142651
x32 = 0.117062	x32 = 0.107889
x33 = 0.132452	x33 = 0.141539
x34 = 0.117989	x34 = 0.109007
x35 = 0.131631	x35 = 0.140489
x36 = 0.118729	x36 = 0.110012
x37 = 0.130904	x37 = 0.139462
x38 = 0.119459	x38 = 0.111078
x39 = 0.130202	x39 = 0.138387
x40 = 0.120107	x40 = 0.112133
x41 = 0.129602	x41 = 0.137351
x42 = 0.120681	x42 = 0.113174
x43 = 0.129046	x43 = 0.136296
x44 = 0.121212	x44 = 0.114233
x45 = 0.128548	x45 = 0.135243
x46 = 0.121677	x46 = 0.115279

x47 = 0.128108
x48 = 0.122097
x49 = 0.127709
x50 = 0.122472
x51 = 0.127358
x52 = 0.122803
x53 = 0.127045
x54 = 0.123099
x55 = 0.126766
x56 = 0.123361
x57 = 0.126518
x58 = 0.123595
x59 = 0.126297
x60 = 0.123805
x61 = 0.126098
x62 = 0.123994
x63 = 0.125917
x64 = 0.124168
x65 = 0.125749
x66 = 0.124332
x67 = 0.125589
x68 = 0.12449
x69 = 0.125432
x70 = 0.124647
x71 = 0.125272
x72 = 0.124811
x73 = 0.125103
x74 = 0.124987
x75 = 0.124919
x76 = 0.125182
x77 = 0.12471
x78 = 0.125405
x79 = 0.124473
x80 = 0.12566
x81 = 0.124194
x82 = 0.125963
x83 = 0.123868
x84 = 0.126312
x85 = 0.123491
x86 = 0.126727
x87 = 0.123036
x88 = 0.127214
x89 = 0.122523
x90 = 0.127762
x91 = 0.121917
x92 = 0.128433
x93 = 0.121206
x94 = 0.129154
x95 = 0.12046
x96 = 0.129995
x97 = 0.119497
x98 = 0.131007
x99 = 0.118543

x47 = 0.134196
x48 = 0.116331
x49 = 0.133142
x50 = 0.117383
x51 = 0.132094
x52 = 0.118431
x53 = 0.131042
x54 = 0.119484
x55 = 0.129991
x56 = 0.120534
x57 = 0.12894
x58 = 0.121585
x59 = 0.127889
x60 = 0.122636
x61 = 0.126839
x62 = 0.123687
x63 = 0.125787
x64 = 0.124738
x65 = 0.124738
x66 = 0.125787
x67 = 0.123687
x68 = 0.126839
x69 = 0.122636
x70 = 0.127889
x71 = 0.121585
x72 = 0.12894
x73 = 0.120534
x74 = 0.129991
x75 = 0.119484
x76 = 0.131042
x77 = 0.118431
x78 = 0.132094
x79 = 0.117383
x80 = 0.133142
x81 = 0.116331
x82 = 0.134196
x83 = 0.115279
x84 = 0.135243
x85 = 0.114233
x86 = 0.136296
x87 = 0.113174
x88 = 0.137351
x89 = 0.112133
x90 = 0.138387
x91 = 0.111078
x92 = 0.139462
x93 = 0.110012
x94 = 0.140489
x95 = 0.109007
x96 = 0.141539
x97 = 0.107889
x98 = 0.142651
x99 = 0.106891

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x100 = 0.131923
x101 = 0.117456
x102 = 0.13328
x103 = 0.116011
x104 = 0.134472
x105 = 0.115115
x106 = 0.135717
x107 = 0.113176
x108 = 0.137936
x109 = 0.111707
x110 = 0.138157
x111 = 0.110665
x112 = 0.14107
x113 = 0.106669
x114 = 0.14328
x115 = 0.109232
x116 = 0.143028
x117 = 0.104604
x118 = 0.151612
x119 = 0.0987309
x120 = 0.139118
x121 = 0.102904
x122 = 0.146319
x123 = 0.0831468
x124 = 0.171417
x125 = 0.132339
x126 = 0.160212
x127 = 0.123818
x128 = 0.198259

```

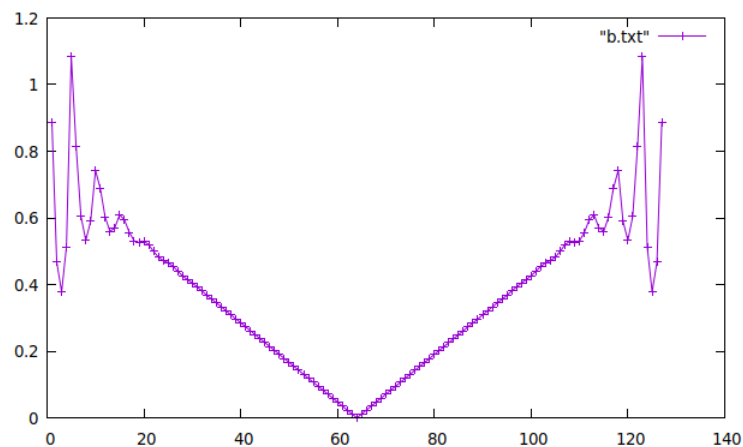
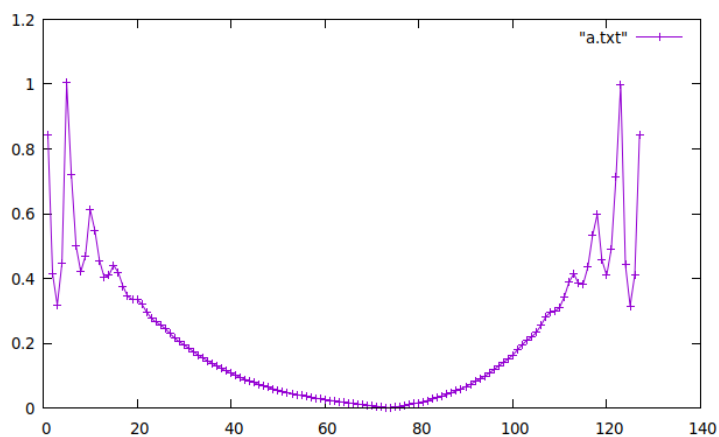
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x100 = 0.143557
x101 = 0.105868
x102 = 0.144791
x103 = 0.104607
x104 = 0.145741
x105 = 0.104005
x106 = 0.146641
x107 = 0.102471
x108 = 0.148389
x109 = 0.101533
x110 = 0.14804
x111 = 0.101092
x112 = 0.150295
x113 = 0.0978345
x114 = 0.151699
x115 = 0.101208
x116 = 0.150655
x117 = 0.0974205
x118 = 0.158277
x119 = 0.0926282
x120 = 0.144773
x121 = 0.0976596
x122 = 0.15109
x123 = 0.0790087
x124 = 0.174738
x125 = 0.129411
x126 = 0.162844
x127 = 0.121552
x128 = 0.19986

```

Wyliczone normy zapisałam w plikach a.txt i b.txt

Oto wykresy na ich podstawie (odpowiednio dla metody Gaussa-Seidela i gradientów sprzężonych), narysowane w gnuplocie:



Złożoność obliczeniowa rozkładu Cholesky'ego dla tej macierzy wynosi $O(n^3)$

Złożoność metody Gaussa-Seidela wynosi $O(n)$, natomiast gradientów sprzężonych $O(k \cdot n^2)$ gdzie k jest szerokością pasma.

Widać zatem, że te metody są dużo wydajniejsze od metody Cholesky'ego.