

Лабораторная работа 5. Решение задачи Дирихле  
для уравнения Пуассона методом установления.  
Вариант 2

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2021  
Май

## Описание задачи

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Имеем задачу Дирихле для уравнения Пуассона:

$$\begin{cases} \frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} = -25\pi^2 \sin(3\pi x) \sin(4\pi y) \\ u|_{\Gamma} = 0 \end{cases} \quad (1)$$

$x$  и  $y$  заданы на отрезке от 0 до 1.

## Аналитическое решение

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Аналитическое решение можно подобрать, заметив, что можно искать решение в виде  $u = a \cdot \sin(3\pi x) \sin(4\pi y)$ . Тогда, подставив решение в таком виде в исходное уравнение, найдем, что  $a = 1$ . Итоговое аналитическое решение есть  $u = \sin(3\pi x) \sin(4\pi y)$ .

## Программная реализация и практические исследования

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В качестве разностной схемы рассмотрим:

$$\begin{cases} \frac{u_{l,m}^{n+1} - u_{l,m}^n}{\tau_n} = \frac{u_{l+1,m}^n - 2u_{l,m}^n + u_{l-1,m}^n}{h_x^2} + \frac{u_{l,m+1}^n - 2u_{l,m}^n + u_{l,m-1}^n}{h_y^2} + 25\pi^2 \sin(3\pi x) (\sin(4\pi y)) \\ l = 1, 2, \dots, L-1, m = 1, 2, \dots, M-1, n = 0, 1, \dots, N \\ u_{l,m}^0 = \psi_{l,m}, l = 0, 1, \dots, L, m = 0, 1, \dots, M \\ u_{0,m}^n = u_{L,m}^n = u_{l,0}^n = u_{l,M}^n = 0, l = 1, 2, \dots, L-1, m = 0, 1, \dots, M, n = 0, 1, \dots, N \end{cases} \quad (2)$$

Здесь  $\psi$  - произвольная функция, определенная в области интегрирования. Для удобства зададим ее равной нулю.

Определимся с оптимальными шагами по времени. Они задаются следующим выражением:

$$\tau_n = \frac{2}{2\pi^2 + 4(L^2 + M^2) + [4(L^2 + M^2) - 2\pi^2] \cos(\frac{\pi(2n-1)}{2N})} \quad (3)$$

Для придания устойчивости необходимо изменить порядок следования элементов в полученной последовательности  $\tau_n$ . Как это сделать - написано в предложенной к лабораторной работе литературы. Здесь же будет приведена итоговая последовательность с измененным порядком для каждой тройки  $L, M, N$ .

В нашем случае, ошибка  $\epsilon$  задана равной  $10^{-4}$ . Для придания устойчивости  $N$  оценивается как

$$N \geq \ln\left(\frac{2}{\epsilon}\right) / \ln \frac{\sqrt{4(L^2 + M^2)} + \sqrt{2\pi^2}}{\sqrt{4(L^2 + M^2)} - \sqrt{2\pi^2}} \quad (4)$$

Поэтому, задавая набор  $L, M$  мы сразу определяем и необходимое значение  $N$ , и

набор  $\tau_n$ .

Перейдем к расчетам. Их будем проводить на последовательно удваеваемых сетках  $L = M$ , начиная с  $L = 5$ , заканчивая  $L = 320$ .

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N = 16, L = 5, M = 5
theta(n) = [1.0, 31.0, 15.0, 17.0, 7.0, 25.0, 9.0, 23.0, 3.0, 29.0, 13.0, 19.0, 5.0, 27.0, 11.0, 21.0]
Numerical solution:
6x6 DataFrame
  Row  0.0    0.2    0.4    0.6    0.8    1.0
     Float64 Float64 Float64 Float64 Float64 Float64
1      0.0    0.0    0.0    0.0    0.0    0.0
2      0.0  0.884718 -1.4315  1.4315 -0.884718  0.0
3      0.0 -0.546786  0.884718 -0.884718  0.546786  0.0
4      0.0 -0.546786  0.884718 -0.884718  0.546786  0.0
5      0.0  0.884718 -1.4315  1.4315 -0.884718  0.0
6      0.0    0.0    0.0    0.0    0.0    0.0
Analytical solution:
6x6 DataFrame
  Row  0.0    0.2    0.4    0.6    0.8    1.0
     Float64 Float64 Float64 Float64 Float64 Float64
1      0.0    0.0    0.0    0.0    0.0    0.0
2      0.0  0.559017 -0.904508  0.904508 -0.559017  0.0
3      0.0 -0.345492  0.559017 -0.559017  0.345492  0.0
4      0.0 -0.345492  0.559017 -0.559017  0.345492  0.0
5      0.0  0.559017 -0.904508  0.904508 -0.559017  0.0
6      0.0    0.0    0.0    0.0    0.0    0.0
Error:
6x6 DataFrame
  Row  0.0    0.2    0.4    0.6    0.8    1.0
     Float64 Float64 Float64 Float64 Float64 Float64
1      0.0    0.0    0.0    0.0    0.0    0.0
2      0.0  0.325701  0.526996  0.526996  0.325701  0.0
3      0.0  0.201295  0.325701  0.325701  0.201295  0.0
4      0.0  0.201295  0.325701  0.325701  0.201295  0.0
5      0.0  0.325701  0.526996  0.526996  0.325701  0.0
6      0.0    0.0    0.0    0.0    0.0    0.0
Max error = 0.5269958645393001
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N = 32, L = 10, M = 10
theta(n) = [1.0, 63.0, 31.0, 33.0, 15.0, 49.0, 17.0, 47.0, 7.0, 57.0, 25.0, 39.0, 9.0, 55.0, 23.0, 41.0, 3.0, 61.0, 29.0, 35.0, 13.0, 51.0, 19.0, 45.0, 5.0, 59.0, 27.0, 37.0, 11.0, 53.0, 21.0, 43.0]
Numerical solution:
6x6 DataFrame
  Row  0.0    0.2    0.4    0.6    0.8    1.0
     Float64 Float64 Float64 Float64 Float64 Float64
1      0.0    0.0    0.0    0.0    0.0    0.0
2      0.0  0.624343 -1.01021  1.01021 -0.624343  0.0
3      0.0 -0.385865  0.624343 -0.624343  0.385865  0.0
4      0.0 -0.385865  0.624343 -0.624343  0.385865  0.0
5      0.0  0.624343 -1.01021  1.01021 -0.624343  0.0
6      0.0    0.0    0.0    0.0    0.0    0.0
Analytical solution:
6x6 DataFrame
  Row  0.0    0.2    0.4    0.6    0.8    1.0
     Float64 Float64 Float64 Float64 Float64 Float64
1      0.0    0.0    0.0    0.0    0.0    0.0
2      0.0  0.559017 -0.904508  0.904508 -0.559017  0.0
3      0.0 -0.345492  0.559017 -0.559017  0.345492  0.0
4      0.0 -0.345492  0.559017 -0.559017  0.345492  0.0
5      0.0  0.559017 -0.904508  0.904508 -0.559017  0.0
6      0.0    0.0    0.0    0.0    0.0    0.0
Error:
6x6 DataFrame
  Row  0.0    0.2    0.4    0.6    0.8    1.0
     Float64 Float64 Float64 Float64 Float64 Float64
1      0.0    0.0    0.0    0.0    0.0    0.0
2      0.0  0.0653258  0.105699  0.105699  0.0653258  0.0
3      0.0  0.0403736  0.0653258  0.0653258  0.0403736  0.0
4      0.0  0.0403736  0.0653258  0.0653258  0.0403736  0.0
5      0.0  0.0653258  0.105699  0.105699  0.0653258  0.0
6      0.0    0.0    0.0    0.0    0.0    0.0
Max error = 0.11113892056633368
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N = 63, L = 20, M = 20  
theta(n) = [1.0, 125.0, 61.0, 65.0, 29.0, 97.0, 33.0, 93.0, 13.0, 113.0, 49.0, 77.0, 17.0, 109.0, 45.0, 81.0, 5.0, 121.0, 57.0, 69.0, 25.0, 101.0, 37.0, 89.0, 9.0, 117.0, 53.0, 73.0, 21.0, 105.0, 41.0, 85.0, 3.0, 123.0, 59.0, 67.0, 27.0, 99.0, 35.0, 91.0, 11.0, 115.0, 51.0, 75.0, 19.0, 107.0, 43.0, 83.0, 7.0, 119.0, 55.0, 71.0, 23.0, 103.0, 39.0, 87.0, 15.0, 111.0, 47.0, 79.0, 31.0, 95.0, 63.0]  
Numerical solution:

6x6 DataFrame

Row	0.0	0.2	0.4	0.6	0.8	1.0
	Float64	Float64	Float64	Float64	Float64	Float64
1	0.0	0.0	0.0	0.0	0.0	0.0
2	0.0	0.574759	-0.929979	0.929979	-0.574759	0.0
3	0.0	-0.35522	0.574759	-0.574759	0.35522	0.0
4	0.0	-0.35522	0.574759	-0.574759	0.35522	0.0
5	0.0	0.574759	-0.929979	0.929979	-0.574759	0.0
6	0.0	0.0	0.0	0.0	0.0	0.0

Analytical solution:

6x6 DataFrame

Row	0.0	0.2	0.4	0.6	0.8	1.0
	Float64	Float64	Float64	Float64	Float64	Float64
1	0.0	0.0	0.0	0.0	0.0	0.0
2	0.0	0.559017	-0.904508	0.904508	-0.559017	0.0
3	0.0	-0.345492	0.559017	-0.559017	0.345492	0.0
4	0.0	-0.345492	0.559017	-0.559017	0.345492	0.0
5	0.0	0.559017	-0.904508	0.904508	-0.559017	0.0
6	0.0	0.0	0.0	0.0	0.0	0.0

Error:

6x6 DataFrame

Row	0.0	0.2	0.4	0.6	0.8	1.0
	Float64	Float64	Float64	Float64	Float64	Float64
1	0.0	0.0	0.0	0.0	0.0	0.0
2	0.0	0.0157416	0.0254704	0.0254704	0.0157416	0.0
3	0.0	0.00972883	0.0157416	0.0157416	0.00972883	0.0
4	0.0	0.00972883	0.0157416	0.0157416	0.00972883	0.0
5	0.0	0.0157416	0.0254704	0.0254704	0.0157416	0.0
6	0.0	0.0	0.0	0.0	0.0	0.0

Max error = 0.026781161447603896

N = 127, L = 40, M = 40  
theta(n) = [1.0, 253.0, 125.0, 129.0, 61.0, 193.0, 65.0, 189.0, 29.0, 225.0, 97.0, 157.0, 33.0, 221.0, 93.0, 161.0, 13.0, 241.0, 113.0, 141.0, 49.0, 205.0, 77.0, 177.0, 17.0, 237.0, 109.0, 145.0, 45.0, 209.0, 81.0, 173.0, 5.0, 249.0, 121.0, 133.0, 57.0, 197.0, 69.0, 185.0, 25.0, 229.0, 101.0, 153.0, 37.0, 217.0, 89.0, 165.0, 9.0, 245.0, 117.0, 137.0, 53.0, 201.0, 73.0, 181.0, 21.0, 233.0, 105.0, 149.0, 41.0, 213.0, 85.0, 169.0, 3.0, 251.0, 123.0, 131.0, 59.0, 195.0, 67.0, 187.0, 27.0, 227.0, 99.0, 155.0, 35.0, 219.0, 91.0, 163.0, 11.0, 243.0, 115.0, 139.0, 51.0, 203.0, 75.0, 179.0, 19.0, 235.0, 107.0, 147.0, 43.0, 211.0, 83.0, 171.0, 7.0, 247.0, 119.0, 135.0, 55.0, 199.0, 71.0, 183.0, 23.0, 231.0, 103.0, 151.0, 39.0, 215.0, 87.0, 167.0, 15.0, 239.0, 111.0, 143.0, 47.0, 207.0, 79.0, 175.0, 31.0, 223.0, 95.0, 159.0, 63.0, 191.0, 127.0]  
Numerical solution:

6x6 DataFrame

Row	0.0	0.2	0.4	0.6	0.8	1.0
	Float64	Float64	Float64	Float64	Float64	Float64
1	0.0	0.0	0.0	0.0	0.0	0.0
2	0.0	0.562903	-0.910796	0.910796	-0.562903	0.0
3	0.0	-0.347893	0.562903	-0.562903	0.347893	0.0
4	0.0	-0.347893	0.562903	-0.562903	0.347893	0.0
5	0.0	0.562903	-0.910796	0.910796	-0.562903	0.0
6	0.0	0.0	0.0	0.0	0.0	0.0

Analytical solution:

6x6 DataFrame

Row	0.0	0.2	0.4	0.6	0.8	1.0
	Float64	Float64	Float64	Float64	Float64	Float64
1	0.0	0.0	0.0	0.0	0.0	0.0
2	0.0	0.559017	-0.904508	0.904508	-0.559017	0.0
3	0.0	-0.345492	0.559017	-0.559017	0.345492	0.0
4	0.0	-0.345492	0.559017	-0.559017	0.345492	0.0
5	0.0	0.559017	-0.904508	0.904508	-0.559017	0.0
6	0.0	0.0	0.0	0.0	0.0	0.0

Error:

6x6 DataFrame

Row	0.0	0.2	0.4	0.6	0.8	1.0
	Float64	Float64	Float64	Float64	Float64	Float64
1	0.0	0.0	0.0	0.0	0.0	0.0
2	0.0	0.00388559	0.00628701	0.00628701	0.00388559	0.0
3	0.0	0.00240142	0.00388559	0.00388559	0.00240142	0.0
4	0.0	0.00240142	0.00388559	0.00388559	0.00240142	0.0
5	0.0	0.00388559	0.00628701	0.00628701	0.00388559	0.0
6	0.0	0.0	0.0	0.0	0.0	0.0

Max error = 0.006950747671868429

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N = 253, L = 80, M = 80
theta(n) = [1.0, 505.0, 251.0, 255.0, 125.0, 381.0, 127.0, 379.0, 61.0, 445.0, 191.0, 315.0, 65.0, 441.0, 187.0, 319.0, 29.0,
477.0, 223.0, 283.0, 97.0, 409.0, 155.0, 351.0, 33.0, 473.0, 219.0, 287.0, 93.0, 413.0, 159.0, 347.0, 13.0, 493.0, 239.0, 267.
0, 113.0, 393.0, 139.0, 367.0, 49.0, 457.0, 203.0, 303.0, 77.0, 429.0, 175.0, 331.0, 17.0, 489.0, 235.0, 271.0, 109.0, 397.0, 1
43.0, 363.0, 45.0, 461.0, 207.0, 299.0, 81.0, 425.0, 171.0, 335.0, 5.0, 501.0, 247.0, 259.0, 121.0, 385.0, 131.0, 375.0, 57.0,
449.0, 195.0, 311.0, 69.0, 437.0, 183.0, 323.0, 25.0, 481.0, 227.0, 279.0, 101.0, 405.0, 151.0, 355.0, 37.0, 469.0, 215.0, 291.
0, 89.0, 417.0, 163.0, 343.0, 9.0, 497.0, 243.0, 263.0, 117.0, 389.0, 135.0, 371.0, 53.0, 453.0, 199.0, 307.0, 73.0, 433.0, 17
9.0, 327.0, 21.0, 485.0, 231.0, 275.0, 105.0, 401.0, 147.0, 359.0, 41.0, 465.0, 211.0, 295.0, 85.0, 421.0, 167.0, 339.0, 3.0, 5
03.0, 249.0, 257.0, 123.0, 383.0, 129.0, 377.0, 59.0, 447.0, 193.0, 313.0, 67.0, 439.0, 185.0, 321.0, 27.0, 479.0, 225.0, 281.
0, 99.0, 407.0, 153.0, 353.0, 35.0, 471.0, 217.0, 289.0, 91.0, 415.0, 161.0, 345.0, 11.0, 495.0, 241.0, 265.0, 115.0, 391.0, 13
7.0, 369.0, 51.0, 455.0, 201.0, 305.0, 75.0, 431.0, 177.0, 329.0, 19.0, 487.0, 233.0, 273.0, 107.0, 399.0, 145.0, 361.0, 43.0,
463.0, 209.0, 297.0, 83.0, 423.0, 169.0, 337.0, 7.0, 499.0, 245.0, 261.0, 119.0, 387.0, 133.0, 373.0, 55.0, 451.0, 197.0, 309.
0, 71.0, 435.0, 181.0, 325.0, 23.0, 483.0, 229.0, 277.0, 103.0, 403.0, 149.0, 357.0, 39.0, 467.0, 213.0, 293.0, 87.0, 419.0, 16
5.0, 341.0, 15.0, 491.0, 237.0, 269.0, 111.0, 395.0, 141.0, 365.0, 47.0, 459.0, 205.0, 301.0, 79.0, 427.0, 173.0, 333.0, 31.0,
475.0, 221.0, 285.0, 95.0, 411.0, 157.0, 349.0, 63.0, 443.0, 189.0, 317.0, 253.0]
Numerical solution:

```

**6x6 DataFrame**

Row	0.0	0.2	0.4	0.6	0.8	1.0
	Float64	Float64	Float64	Float64	Float64	Float64
1	0.0	0.0	0.0	0.0	0.0	0.0
2	0.0	0.559983	-0.906072	0.906072	-0.559983	0.0
3	0.0	-0.346089	0.559983	-0.559983	0.346089	0.0
4	0.0	-0.346089	0.559983	-0.559983	0.346089	0.0
5	0.0	0.559983	-0.906072	0.906072	-0.559983	0.0
6	0.0	0.0	0.0	0.0	0.0	0.0

Analytical solution:

**6x6 DataFrame**

Row	0.0	0.2	0.4	0.6	0.8	1.0
	Float64	Float64	Float64	Float64	Float64	Float64
1	0.0	0.0	0.0	0.0	0.0	0.0
2	0.0	0.559017	-0.904508	0.904508	-0.559017	0.0
3	0.0	-0.345492	0.559017	-0.559017	0.345492	0.0
4	0.0	-0.345492	0.559017	-0.559017	0.345492	0.0
5	0.0	0.559017	-0.904508	0.904508	-0.559017	0.0
6	0.0	0.0	0.0	0.0	0.0	0.0

Error:

**6x6 DataFrame**

Row	0.0	0.2	0.4	0.6	0.8	1.0
	Float64	Float64	Float64	Float64	Float64	Float64
1	0.0	0.0	0.0	0.0	0.0	0.0
2	0.0	0.00096622	0.00156338	0.00156338	0.00096622	0.0
3	0.0	0.000597157	0.00096622	0.00096622	0.000597157	0.0
4	0.0	0.000597157	0.00096622	0.00096622	0.000597157	0.0
5	0.0	0.00096622	0.00156338	0.00156338	0.00096622	0.0
6	0.0	0.0	0.0	0.0	0.0	0.0

Max error = 0.0017284271030424403

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N = 505, L = 160, M = 160
theta(n) = [1.0, 1009.0, 503.0, 507.0, 251.0, 759.0, 253.0, 757.0, 125.0, 885.0, 379.0, 631.0, 127.0, 883.0, 377.0, 633.0, 61.0, 949.0, 443.0, 567.0, 191.0, 819.0, 313.0, 697.0, 65.0, 945.0, 439.0, 571.0, 187.0, 823.0, 317.0, 693.0, 29.0, 981.0, 475.0, 535.0, 223.0, 787.0, 281.0, 729.0, 97.0, 913.0, 407.0, 603.0, 155.0, 855.0, 349.0, 661.0, 33.0, 977.0, 471.0, 539.0, 219.0, 791.0, 285.0, 725.0, 93.0, 917.0, 411.0, 599.0, 159.0, 851.0, 345.0, 665.0, 13.0, 997.0, 491.0, 519.0, 239.0, 771.0, 265.0, 745.0, 113.0, 897.0, 391.0, 619.0, 139.0, 871.0, 365.0, 645.0, 49.0, 961.0, 455.0, 555.0, 203.0, 807.0, 301.0, 709.0, 77.0, 933.0, 427.0, 583.0, 175.0, 835.0, 329.0, 681.0, 17.0, 993.0, 487.0, 523.0, 235.0, 775.0, 269.0, 741.0, 109.0, 901.0, 395.0, 615.0, 143.0, 867.0, 361.0, 649.0, 45.0, 965.0, 459.0, 551.0, 207.0, 803.0, 297.0, 713.0, 81.0, 929.0, 423.0, 587.0, 171.0, 839.0, 333.0, 677.0, 5.0, 1005.0, 499.0, 511.0, 247.0, 763.0, 257.0, 753.0, 121.0, 889.0, 383.0, 627.0, 131.0, 879.0, 373.0, 637.0, 57.0, 953.0, 447.0, 563.0, 195.0, 815.0, 309.0, 701.0, 69.0, 941.0, 435.0, 575.0, 183.0, 827.0, 321.0, 689.0, 25.0, 985.0, 479.0, 531.0, 227.0, 783.0, 277.0, 733.0, 101.0, 909.0, 403.0, 607.0, 151.0, 859.0, 353.0, 657.0, 37.0, 973.0, 467.0, 543.0, 215.0, 795.0, 289.0, 721.0, 89.0, 921.0, 415.0, 595.0, 163.0, 847.0, 341.0, 669.0, 9.0, 1001.0, 495.0, 515.0, 243.0, 767.0, 261.0, 749.0, 117.0, 893.0, 387.0, 623.0, 135.0, 875.0, 369.0, 641.0, 53.0, 957.0, 451.0, 559.0, 199.0, 811.0, 305.0, 705.0, 73.0, 937.0, 431.0, 579.0, 179.0, 831.0, 325.0, 685.0, 21.0, 989.0, 483.0, 527.0, 231.0, 779.0, 273.0, 737.0, 105.0, 905.0, 399.0, 611.0, 147.0, 863.0, 357.0, 653.0, 41.0, 969.0, 463.0, 547.0, 211.0, 799.0, 293.0, 717.0, 85.0, 925.0, 419.0, 591.0, 167.0, 843.0, 337.0, 673.0, 3.0, 1007.0, 501.0, 509.0, 249.0, 761.0, 255.0, 755.0, 123.0, 887.0, 381.0, 629.0, 129.0, 881.0, 375.0, 635.0, 59.0, 951.0, 445.0, 565.0, 193.0, 817.0, 311.0, 699.0, 67.0, 943.0, 437.0, 573.0, 185.0, 825.0, 319.0, 691.0, 27.0, 983.0, 477.0, 533.0, 225.0, 785.0, 279.0, 731.0, 99.0, 911.0, 405.0, 605.0, 153.0, 857.0, 351.0, 659.0, 35.0, 975.0, 469.0, 541.0, 217.0, 793.0, 287.0, 723.0, 91.0, 919.0, 413.0, 597.0, 161.0, 849.0, 343.0, 667.0, 11.0, 999.0, 493.0, 517.0, 241.0, 769.0, 263.0, 747.0, 115.0, 895.0, 389.0, 621.0, 137.0, 873.0, 367.0, 643.0, 51.0, 959.0, 453.0, 557.0, 201.0, 809.0, 303.0, 707.0, 75.0, 935.0, 429.0, 581.0, 177.0, 833.0, 327.0, 683.0, 19.0, 991.0, 485.0, 525.0, 233.0, 777.0, 271.0, 739.0, 107.0, 903.0, 397.0, 613.0, 145.0, 865.0, 359.0, 651.0, 43.0, 967.0, 461.0, 549.0, 209.0, 801.0, 295.0, 715.0, 83.0, 927.0, 421.0, 589.0, 169.0, 841.0, 335.0, 675.0, 7.0, 1003.0, 497.0, 513.0, 245.0, 765.0, 259.0, 751.0, 119.0, 891.0, 385.0, 625.0, 133.0, 877.0, 371.0, 639.0, 55.0, 955.0, 449.0, 561.0, 197.0, 813.0, 307.0, 703.0, 71.0, 939.0, 433.0, 577.0, 181.0, 829.0, 323.0, 687.0, 23.0, 987.0, 481.0, 529.0, 229.0, 781.0, 275.0, 735.0, 103.0, 907.0, 401.0, 609.0, 149.0, 861.0, 355.0, 655.0, 39.0, 971.0, 465.0, 545.0, 213.0, 797.0, 291.0, 719.0, 87.0, 923.0, 417.0, 593.0, 165.0, 845.0, 339.0, 671.0, 15.0, 995.0, 489.0, 521.0, 237.0, 773.0, 267.0, 743.0, 111.0, 899.0, 393.0, 617.0, 141.0, 869.0, 363.0, 647.0, 47.0, 963.0, 457.0, 553.0, 205.0, 805.0, 299.0, 711.0, 79.0, 931.0, 425.0, 585.0, 173.0, 837.0, 331.0, 679.0, 31.0, 979.0, 473.0, 537.0, 221.0, 789.0, 283.0, 727.0, 95.0, 915.0, 409.0, 601.0, 157.0, 853.0, 347.0, 663.0, 63.0, 947.0, 441.0, 569.0, 189.0, 821.0, 315.0, 695.0, 505.0]
Numerical solution:
6x6 DataFrame
  Row  0.0      0.2      0.4      0.6      0.8      1.0
    Float64 Float64 Float64 Float64 Float64 Float64
1      0.0      0.0      0.0      0.0      0.0      0.0
2      0.0  0.559256 -0.904896  0.904896 -0.559256  0.0
3      0.0 -0.345639  0.559256 -0.559256  0.345639  0.0
4      0.0 -0.345639  0.559256 -0.559256  0.345639  0.0
5      0.0  0.559256 -0.904896  0.904896 -0.559256  0.0
6      0.0      0.0      0.0      0.0      0.0      0.0
Analytical solution:
6x6 DataFrame
  Row  0.0      0.2      0.4      0.6      0.8      1.0
    Float64 Float64 Float64 Float64 Float64 Float64
1      0.0      0.0      0.0      0.0      0.0      0.0
2      0.0  0.559017 -0.904508  0.904508 -0.559017  0.0
3      0.0 -0.345492  0.559017 -0.559017  0.345492  0.0
4      0.0 -0.345492  0.559017 -0.559017  0.345492  0.0
5      0.0  0.559017 -0.904508  0.904508 -0.559017  0.0
6      0.0      0.0      0.0      0.0      0.0      0.0
Error:
6x6 DataFrame
  Row  0.0      0.2      0.4      0.6      0.8      1.0
    Float64 Float64 Float64 Float64 Float64 Float64
1      0.0  0.0  0.000239216  0.00038706  0.00038706  0.000239216  0.0
2      0.0  0.000147844  0.000239216  0.000239216  0.000147844  0.0
3      0.0  0.000147844  0.000239216  0.000239216  0.000147844  0.0
4      0.0  0.000239216  0.00038706  0.00038706  0.000239216  0.0
5      0.0  0.000239216  0.00038706  0.00038706  0.000239216  0.0
6      0.0  0.0  0.00042792310977457504  0.0  0.0  0.0
Max error = 0.00042792310977457504
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N = 1009, L = 320, M = 320
thetta(n) = [1.0, 2017.0, 1007.0, 1011.0, 503.0, 1515.0, 505.0, 1513.0, 251.0, 1767.0, 757.0, 1261.0, 253.0, 1765.0, 755.0, 126
3.0, 125.0, 1893.0, 883.0, 1135.0, 379.0, 1639.0, 629.0, 1389.0, 127.0, 1891.0, 881.0, 1137.0, 377.0, 1641.0, 631.0, 1387.0, 6
1.0, 1957.0, 947.0, 1071.0, 443.0, 1575.0, 565.0, 1453.0, 191.0, 1827.0, 817.0, 1201.0, 313.0, 1705.0, 695.0, 1323.0, 65.0, 195
3.0, 943.0, 1075.0, 439.0, 1579.0, 569.0, 1449.0, 187.0, 1831.0, 821.0, 1197.0, 317.0, 1701.0, 691.0, 1327.0, 29.0, 1989.0, 97
9.0, 1039.0, 475.0, 1543.0, 533.0, 1485.0, 223.0, 1795.0, 785.0, 1233.0, 281.0, 1737.0, 727.0, 1291.0, 97.0, 1921.0, 911.0, 110
7.0, 407.0, 1611.0, 601.0, 1417.0, 155.0, 1863.0, 853.0, 1165.0, 349.0, 1669.0, 659.0, 1359.0, 33.0, 1985.0, 975.0, 1043.0, 47
1.0, 1547.0, 537.0, 1481.0, 219.0, 1799.0, 789.0, 1229.0, 285.0, 1733.0, 723.0, 1295.0, 93.0, 1925.0, 915.0, 1103.0, 411.0, 160
7.0, 597.0, 1421.0, 159.0, 1859.0, 849.0, 1169.0, 345.0, 1673.0, 663.0, 1355.0, 13.0, 2005.0, 995.0, 1023.0, 491.0, 1527.0, 51
7.0, 1501.0, 239.0, 1779.0, 769.0, 1249.0, 265.0, 1753.0, 743.0, 1275.0, 113.0, 1905.0, 895.0, 1123.0, 391.0, 1627.0, 617.0, 14
01.0, 139.0, 1879.0, 869.0, 1149.0, 365.0, 1653.0, 643.0, 1375.0, 49.0, 1969.0, 959.0, 1059.0, 455.0, 1563.0, 553.0, 1465.0, 20
3.0, 1815.0, 805.0, 1213.0, 301.0, 1717.0, 707.0, 1311.0, 77.0, 1941.0, 931.0, 1087.0, 427.0, 1591.0, 581.0, 1437.0, 175.0, 184
3.0, 833.0, 1185.0, 329.0, 1689.0, 679.0, 1339.0, 17.0, 2001.0, 991.0, 1027.0, 487.0, 1531.0, 521.0, 1497.0, 235.0, 1783.0, 77
3.0, 1245.0, 269.0, 1749.0, 739.0, 1279.0, 109.0, 1909.0, 899.0, 1119.0, 395.0, 1623.0, 613.0, 1405.0, 143.0, 1875.0, 865.0, 11
53.0, 361.0, 1657.0, 647.0, 1371.0, 45.0, 1973.0, 963.0, 1055.0, 459.0, 1559.0, 549.0, 1469.0, 207.0, 1811.0, 801.0, 1217.0, 29
7.0, 1721.0, 711.0, 1307.0, 81.0, 1937.0, 927.0, 1091.0, 423.0, 1595.0, 585.0, 1433.0, 171.0, 1847.0, 837.0, 1181.0, 333.0, 168
5.0, 675.0, 1343.0, 5.0, 2013.0, 1003.0, 1015.0, 499.0, 1519.0, 509.0, 1509.0, 247.0, 1771.0, 761.0, 1257.0, 257.0, 1761.0, 75
1.0, 1267.0, 121.0, 1897.0, 887.0, 1131.0, 383.0, 1635.0, 625.0, 1393.0, 131.0, 1887.0, 877.0, 1141.0, 373.0, 1645.0, 635.0, 13
83.0, 57.0, 1961.0, 951.0, 1067.0, 447.0, 1571.0, 561.0, 1457.0, 195.0, 1823.0, 813.0, 1205.0, 309.0, 1709.0, 699.0, 1319.0, 6
9.0, 1949.0, 939.0, 1079.0, 435.0, 1583.0, 573.0, 1445.0, 183.0, 1835.0, 825.0, 1193.0, 321.0, 1697.0, 687.0, 1331.0, 25.0, 199
3.0, 983.0, 1035.0, 479.0, 1539.0, 529.0, 1489.0, 227.0, 1791.0, 781.0, 1237.0, 277.0, 1741.0, 731.0, 1287.0, 101.0, 1917.0, 90
7.0, 1111.0, 403.0, 1615.0, 605.0, 1413.0, 151.0, 1867.0, 857.0, 1161.0, 353.0, 1665.0, 655.0, 1363.0, 37.0, 1981.0, 971.0, 104
7.0, 467.0, 1551.0, 541.0, 1477.0, 215.0, 1803.0, 793.0, 1225.0, 289.0, 1729.0, 719.0, 1299.0, 89.0, 1929.0, 919.0, 1099.0, 41
5.0, 1603.0, 593.0, 1425.0, 163.0, 1855.0, 845.0, 1173.0, 341.0, 1677.0, 667.0, 1351.0, 9.0, 2009.0, 999.0, 1019.0, 495.0, 152
3.0, 513.0, 1505.0, 243.0, 1775.0, 765.0, 1253.0, 261.0, 1757.0, 747.0, 1271.0, 117.0, 1901.0, 891.0, 1127.0, 387.0, 1631.0, 62
1.0, 1397.0, 135.0, 1883.0, 873.0, 1145.0, 369.0, 1649.0, 639.0, 1379.0, 53.0, 1965.0, 955.0, 1063.0, 451.0, 1567.0, 557.0, 146
1.0, 199.0, 1819.0, 809.0, 1209.0, 305.0, 1713.0, 703.0, 1315.0, 73.0, 1945.0, 935.0, 1083.0, 431.0, 1587.0, 577.0, 1441.0, 17
9.0, 1839.0, 829.0, 1189.0, 325.0, 1693.0, 683.0, 1335.0, 21.0, 1997.0, 987.0, 1031.0, 483.0, 1535.0, 525.0, 1493.0, 231.0, 178
7.0, 777.0, 1241.0, 273.0, 1745.0, 735.0, 1283.0, 105.0, 1913.0, 903.0, 1115.0, 399.0, 1619.0, 609.0, 1409.0, 147.0, 1871.0, 86
1.0, 1157.0, 357.0, 1661.0, 651.0, 1367.0, 41.0, 1977.0, 967.0, 1051.0, 463.0, 1555.0, 545.0, 1473.0, 211.0, 1807.0, 797.0, 122
1.0, 293.0, 1725.0, 715.0, 1303.0, 85.0, 1933.0, 923.0, 1095.0, 419.0, 1599.0, 589.0, 1429.0, 167.0, 1851.0, 841.0, 1177.0, 33
7.0, 1681.0, 671.0, 1347.0, 3.0, 2015.0, 1005.0, 1013.0, 501.0, 1517.0, 507.0, 1511.0, 249.0, 1769.0, 759.0, 1259.0, 255.0, 176
3.0, 753.0, 1265.0, 123.0, 1895.0, 885.0, 1133.0, 381.0, 1637.0, 627.0, 1391.0, 129.0, 1889.0, 879.0, 1139.0, 375.0, 1643.0, 63
3.0, 1385.0, 59.0, 1959.0, 949.0, 1069.0, 445.0, 1573.0, 563.0, 1455.0, 193.0, 1825.0, 815.0, 1203.0, 311.0, 1707.0, 697.0, 132
1.0, 67.0, 1951.0, 941.0, 1077.0, 437.0, 1581.0, 571.0, 1447.0, 185.0, 1833.0, 823.0, 1195.0, 319.0, 1699.0, 689.0, 1329.0, 27.
0, 1991.0, 981.0, 1037.0, 477.0, 1541.0, 531.0, 1487.0, 225.0, 1793.0, 783.0, 1235.0, 279.0, 1739.0, 729.0, 1289.0, 99.0, 1919.
0, 909.0, 1109.0, 405.0, 1613.0, 603.0, 1415.0, 153.0, 1865.0, 855.0, 1163.0, 351.0, 1667.0, 657.0, 1361.0, 35.0, 1983.0, 973.
0, 1045.0, 469.0, 1549.0, 539.0, 1479.0, 217.0, 1801.0, 791.0, 1227.0, 287.0, 1731.0, 721.0, 1297.0, 91.0, 1927.0, 917.0, 1101.
0, 413.0, 1605.0, 595.0, 1423.0, 161.0, 1857.0, 847.0, 1171.0, 343.0, 1675.0, 665.0, 1353.0, 11.0, 2007.0, 997.0, 1021.0, 493.
0, 1525.0, 515.0, 1503.0, 241.0, 1777.0, 767.0, 1251.0, 263.0, 1755.0, 745.0, 1273.0, 115.0, 1903.0, 893.0, 1125.0, 389.0, 162
9.0, 619.0, 1399.0, 137.0, 1881.0, 871.0, 1147.0, 367.0, 1651.0, 641.0, 1377.0, 51.0, 1967.0, 957.0, 1061.0, 453.0, 1565.0, 55
5.0, 1463.0, 201.0, 1817.0, 807.0, 1211.0, 303.0, 1715.0, 705.0, 1313.0, 75.0, 1943.0, 933.0, 1085.0, 429.0, 1589.0, 579.0, 143
9.0, 177.0, 1841.0, 831.0, 1187.0, 327.0, 1691.0, 681.0, 1337.0, 19.0, 1999.0, 989.0, 1029.0, 485.0, 1533.0, 523.0, 1495.0, 23
3.0, 1785.0, 775.0, 1243.0, 271.0, 1747.0, 737.0, 1281.0, 107.0, 1911.0, 901.0, 1117.0, 397.0, 1621.0, 611.0, 1407.0, 145.0, 18
73.0, 863.0, 1155.0, 359.0, 1659.0, 649.0, 1369.0, 43.0, 1975.0, 965.0, 1053.0, 461.0, 1557.0, 547.0, 1471.0, 209.0, 1809.0, 79
9.0, 1219.0, 295.0, 1723.0, 713.0, 1305.0, 83.0, 1935.0, 925.0, 1093.0, 421.0, 1597.0, 587.0, 1431.0, 169.0, 1849.0, 839.0, 117
9.0, 335.0, 1683.0, 673.0, 1345.0, 7.0, 2011.0, 1001.0, 1017.0, 497.0, 1521.0, 511.0, 1507.0, 245.0, 1773.0, 763.0, 1255.0, 25
9.0, 1759.0, 749.0, 1269.0, 119.0, 1899.0, 889.0, 1129.0, 385.0, 1633.0, 623.0, 1395.0, 133.0, 1885.0, 875.0, 1143.0, 371.0, 16
47.0, 637.0, 1381.0, 55.0, 1963.0, 953.0, 1065.0, 449.0, 1569.0, 559.0, 1459.0, 197.0, 1821.0, 811.0, 1207.0, 307.0, 1711.0, 70
1.0, 1317.0, 71.0, 1947.0, 937.0, 1081.0, 433.0, 1585.0, 575.0, 1443.0, 181.0, 1837.0, 827.0, 1191.0, 323.0, 1695.0, 685.0, 133
3.0, 23.0, 1995.0, 985.0, 1033.0, 481.0, 1537.0, 527.0, 1491.0, 229.0, 1789.0, 779.0, 1239.0, 275.0, 1743.0, 733.0, 1285.0, 10
3.0, 1915.0, 905.0, 1113.0, 401.0, 1617.0, 607.0, 1411.0, 149.0, 1869.0, 859.0, 1159.0, 355.0, 1663.0, 653.0, 1365.0, 39.0, 197
9.0, 969.0, 1049.0, 465.0, 1553.0, 543.0, 1475.0, 213.0, 1805.0, 795.0, 1223.0, 291.0, 1727.0, 717.0, 1301.0, 87.0, 1931.0, 92
1.0, 1097.0, 417.0, 1601.0, 591.0, 1427.0, 165.0, 1853.0, 843.0, 1175.0, 339.0, 1679.0, 669.0, 1349.0, 15.0, 2003.0, 993.0, 102
5.0, 489.0, 1529.0, 519.0, 1499.0, 237.0, 1781.0, 771.0, 1247.0, 267.0, 1751.0, 741.0, 1277.0, 111.0, 1907.0, 897.0, 1121.0, 39
3.0, 1625.0, 615.0, 1403.0, 141.0, 1877.0, 867.0, 1151.0, 363.0, 1655.0, 645.0, 1373.0, 47.0, 1971.0, 961.0, 1057.0, 457.0, 156
1.0, 551.0, 1467.0, 205.0, 1813.0, 803.0, 1215.0, 299.0, 1719.0, 709.0, 1309.0, 79.0, 1939.0, 929.0, 1089.0, 425.0, 1593.0, 58
3.0, 1435.0, 173.0, 1845.0, 835.0, 1183.0, 331.0, 1687.0, 677.0, 1341.0, 31.0, 1987.0, 977.0, 1041.0, 473.0, 1545.0, 535.0, 148
3.0, 221.0, 1797.0, 787.0, 1231.0, 283.0, 1735.0, 725.0, 1293.0, 95.0, 1923.0, 913.0, 1105.0, 409.0, 1609.0, 599.0, 1419.0, 15
7.0, 1861.0, 851.0, 1167.0, 347.0, 1671.0, 661.0, 1357.0, 63.0, 1955.0, 945.0, 1073.0, 441.0, 1577.0, 567.0, 1451.0, 189.0, 182
9.0, 819.0, 1199.0, 315.0, 1703.0, 693.0, 1325.0, 1009.0]

```



```

Numerical solution:
6x6 DataFrame
  Row  0.0    0.2    0.4    0.6    0.8    1.0
     Float64 Float64 Float64 Float64 Float64 Float64
1      0.0    0.0    0.0    0.0    0.0    0.0
2      0.0  0.559075 -0.904602  0.904602 -0.559075  0.0
3      0.0 -0.345527  0.559075 -0.559075  0.345527  0.0
4      0.0 -0.345527  0.559075 -0.559075  0.345527  0.0
5      0.0  0.559075 -0.904602  0.904602 -0.559075  0.0
6      0.0    0.0    0.0    0.0    0.0    0.0

Analytical solution:
6x6 DataFrame
  Row  0.0    0.2    0.4    0.6    0.8    1.0
     Float64 Float64 Float64 Float64 Float64 Float64
1      0.0    0.0    0.0    0.0    0.0    0.0
2      0.0  0.559017 -0.904508  0.904508 -0.559017  0.0
3      0.0 -0.345492  0.559017 -0.559017  0.345492  0.0
4      0.0 -0.345492  0.559017 -0.559017  0.345492  0.0
5      0.0  0.559017 -0.904508  0.904508 -0.559017  0.0
6      0.0    0.0    0.0    0.0    0.0    0.0

Error:
6x6 DataFrame
  Row  0.0    0.2    0.4    0.6    0.8    1.0
     Float64 Float64 Float64 Float64 Float64 Float64
1      0.0  0.0    0.0    0.0    0.0    0.0
2      0.0  5.7692e-5  9.33476e-5  9.33476e-5  5.7692e-5  0.0
3      0.0  3.56556e-5  5.7692e-5  5.7692e-5  3.56556e-5  0.0
4      0.0  3.56556e-5  5.7692e-5  5.7692e-5  3.56556e-5  0.0
5      0.0  5.7692e-5  9.33476e-5  9.33476e-5  5.7692e-5  0.0
6      0.0  0.0    0.0    0.0    0.0    0.0
Max error = 0.00010320254066420453
-----

```

## Результаты и обсуждения

В данной работе представлена схема, по которой проводились численные расчеты, найдено аналитическое решение и ее след. Выявлен набор  $\tau_n$  для каждой тройки  $N, L, M$ . Найдено численное решение для каждой тройки с учетом выбранного чебышевского набора, причем наблюдается устойчивость и сходимость решения на последовательно удваиваемых сетках. На последовательно удваиваемых сетках также наблюдается уменьшение ошибки примерно в 4 раза при уменьшении сетки вдвое по каждой оси.