Лабораторная работа №4

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Теория

Задача

$$\frac{\partial u}{\partial t} = \frac{\partial^2 u}{\partial x^2} + t^2 \sin xt + x \cos xt, 0 < x < 1, 0 < t \le 0.5,$$

$$\begin{cases} u(x,0) = 0, 0 \le x \le 1, \\ u(0,t) = 0, 0 \le t \le \frac{1}{2}, \\ u(1,t) = \sin(t), 0 \le t \le \frac{1}{2}. \end{cases}$$

Точное решение

$$u(x,t) = \sin(xt) \tag{1}$$

Сетка

- au шаг сетки по времени, $N_2 = \frac{1}{2 au}$.
- h шаг сетки по пространству, $N_1 = \frac{1}{h}$.
- $\bullet \ \varphi_i^j = f(x_i, t_{j+\frac{1}{2}})$

Граничные условия

$$\begin{cases} y_i^0 = 0, i = \overline{0, N_1} \\ y_0^j = 0, j = \overline{0, N_2} \\ y_{N_1}^j = \sin(j\tau), j = \overline{0, N_2} \end{cases}$$

Разностная схема

- 1. Явная $\frac{y_i^{j+1}-y_i^j}{ au}=\Lambda y_i^j+arphi_i^j$
- 2. Чисто неявная $\frac{y_i^{j+1}-y_i^j}{\tau}=\Lambda y_i^{j+1}+\varphi_i^j$
- 3. Кранка-Николсона $\frac{y_i^{j+1}-y_i^j}{\tau}=\frac{1}{2}(\Lambda y_i^j+\Lambda y_i^{j+1})+\varphi_i^j$

Устойчивость и порядок аппроксимации

- Явная $O(\tau + h^2)$, не является абсолютно устойчивой.
- Чисто неявная $O(\tau + h^2)$, является абсолютно устойчивой.
- Кранка-Николсона $O(\tau^2 + h^2)$, является абсолютно устойчивой.

Листинг кода

```
import numpy as np
import matplotlib.pyplot as plt
from mpl_toolkits.mplot3d import Axes3D
a = 0.
b = 1.
c = 0.
d = 0.5
draw_plots = False
def u(x, t):
    return np.sin(t * x)
def u0(x):
   return 0
def u1(t):
    return 0.
def u2(t):
    return np.sin(t)
def f(x, t):
    return t**2 * np.sin(x*t) + x * np.cos(x * t)
def TDMA(a, b, c, f):
    beta = []
    1 = len(c) - 1
    alfa = [b[0] / c[0]]
    beta.append(f[0] / c[0])
    for i in range(1, 1):
        check = (c[i] - a[i - 1] * alfa[i - 1])
        if abs(check) < 1e-19: exit()</pre>
        alfa.append(b[i] / check)
        \texttt{beta.append((f[i] + a[i-1] * beta[i-1]) / check)}
```

```
\texttt{beta.append}((\texttt{f[1]} + \texttt{a[1-1]} * \texttt{beta[1-1]}) \ / \ (\texttt{c[1]} - \texttt{a[1-1]} * \texttt{alfa[1-1]}))
    y = [0] * (1 + 1)
    y[1] = beta[1]
    for i in range (1 - 1, -1, -1):
        y[i] = alfa[i] * y[i + 1] + beta[i]
    return np.array(y)
def plot(y):
    data = np.array(y)
    length = data.shape[0]
    width = data.shape[1]
    x, y = np.meshgrid(np.arange(width), np.arange(length))
    ax = plt.axes(projection='3d')
    ax.plot_surface(x, y, data)
    if draw_plots:
        plt.show()
def build_exact(h, tau):
    n = int((b - a) / h) + 1
    m = int((d - c) / tau) + 1
    ys = np.zeros((n, m))
    for i in range(n):
        for j in range(m):
            ys[i, j] = u(h*i, tau*j)
    return ys
def build_explicit(h, tau):
    xs = np.linspace(a, b, int((b - a) / h) + 1)
    ts = np.linspace(c, d, int((d - c) / tau) + 1)
    ys = np.zeros((len(xs), len(ts)))
    for i in range(len(xs)):
        ys[i, 0] = u0(xs[i])
    for j in range(len(ts) -1):
        ys[0, j + 1] = u1(ts[j + 1])
        for i in range(1, len(xs) -1):
            a_i = (ys[i + 1, j] - 2 * ys[i, j] + ys[i - 1, j]) / (h * h)
            b_i = f(xs[i], ts[j])
            ys[i, j + 1] = tau * (a_i + b_i) + ys[i, j]
        ys[len(xs) - 1, j + 1] = u2(ts[j + 1])
    plot(ys)
```

```
def build_implicit(h, tau):
    xs = np.linspace(a, b, int((b - a) / h) + 1)
   ts = np.linspace(c, d, int((d - c) / tau) + 1)
   ys = np.zeros((len(xs), len(ts)))
    for i in range(len(xs)):
        ys[i, 0] = u0(xs[i])
    for j in range(len(ts)-1):
        c_i = [1]
        c_i = c_i + [1 / tau + 2 / (h * h)] * (len(xs) - 2)
        c_i.append(1)
        a_i = []
        a_i = a_i + [-1 / (h * h)] * (len(xs) - 2)
        a_i.append(0)
        b_i = [0]
        b_i = b_i + [-1 / (h * h)] * (len(xs) - 2)
        f_i = [u1(ts[j + 1])]
        for i in range(1, len(xs) - 1):
            l = f(xs[i], ts[j + 1])
            f_i.append((ys[i, j] / tau) + 1)
        f_i.append(u2(ts[j + 1]))
        ys[:, j + 1] = TDMA(-np.array(a_i), -np.array(b_i), c_i, f_i)
    plot(ys)
    return ys
def build_nickolson(h, tau):
   xs = np.linspace(a, b, int((b - a) / h) + 1)
   ts = np.linspace(c, d, int((d - c) / tau) + 1)
   ys = np.zeros((len(xs), len(ts)))
    for i in range(len(xs)):
        ys[i, 0] = u0(xs[i])
    for j in range(len(ts)-1):
        c_i = [1]
        c_i = c_i + [1 / tau + 1 / (h * h)] * (len(xs) - 2)
        c_i.append(1)
```

return ys

```
a_i = []
         a_i = a_i + [-1 / (2 * h * h)] * (len(xs) - 2)
         a_i.append(0)
         b_i = [0]
         b_i = b_i + [-1 / (2 * h * h)] * (len(xs) - 2)
         f_i = [u1(ts[j + 1])]
         for i in range(1, len(xs) - 1):
             1 = f(xs[i], ts[j] + 0.5 * tau)
             v = (1 / (2 * h * h)) * (ys[i + 1, j] - 2 * ys[i, j] + ys[i - 1, j])
             f_i.append((ys[i, j] / tau) + l + v)
         f_i.append(u2(ts[j + 1]))
         ys[:, j + 1] = TDMA(-np.array(a_i), -np.array(b_i), c_i, f_i)
    plot(ys)
    return ys
print(f'Explicit: {np.max(np.abs(build_exact(0.1,0.1) -build_explicit(0.1,0.1)))}')
 \textbf{print}(\texttt{f'Explicit: \{np.max(np.abs(build\_exact(0.1,0.005) - build\_explicit(0.1,0.005)))\}')} \\
print(f"Implicit: {np.max(np.abs(build_exact(0.1,0.1) - build_implicit(0.1,0.1)))}")
 \label{eq:print}  \textbf{print}(f"\texttt{Nickolson: } \{\texttt{np.max}(\texttt{np.abs}(\texttt{build\_exact}(\texttt{0.1,0.1}) - \texttt{build\_nickolson}(\texttt{0.1, 0.1}))\}")
```

Графики

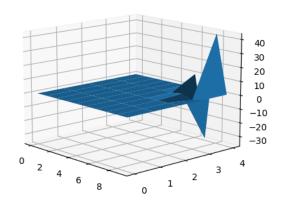


Рис. 1: Явная неустойчивая

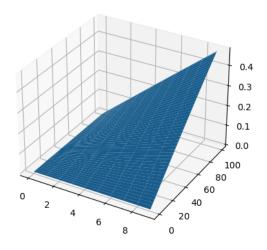


Рис. 2: Явная устойчивая

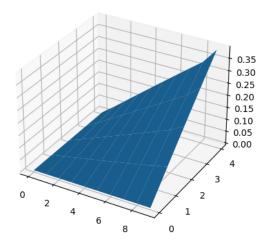


Рис. 3: Чисто неявная

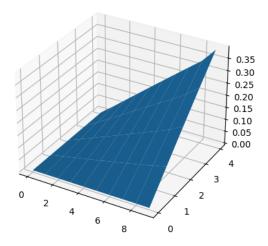


Рис. 4: Кранка-Николсона

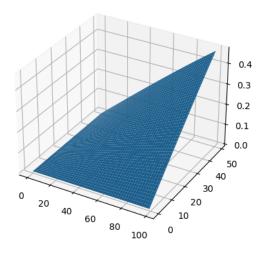


Рис. 5: Точное решение

Результаты вычислительного эксперимента

Схема	$\max \left y_i^j - u(x_i, t_j) \right $
Явная (нейстойчивая)	18.08729087566136
Явная (устойчивая)	2.770470845148143e-05
Чисто неявная	0.0004969953632136814
Кранка-Николсона	0.00017371480558825425