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Лабораторная работа №8 по курсу Численные методы

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Постановка задачи

Вариант 1

Уравнение:

$$\frac{\partial u}{\partial t} = a \frac{\partial^2 u}{\partial x^2} + a \frac{\partial^2 u}{\partial y^2}$$

$$u(0, y, t) = \cos(\mu_2 y) \exp(-(\mu_1^2 + \mu_2^2)at)$$

$$u(\pi, y, t) = (-1)^{\mu_1} \cos(\mu_2 y) \exp(-(\mu_1^2 + \mu_2^2)at)$$

$$u(x, 0, t) = \cos(\mu_1 x) \exp(-(\mu_1^2 + \mu_2^2)at)$$

$$u(x, \pi, t) = (-1)^{\mu_2} \cos(\mu_1 x) \exp(-(\mu_1^2 + \mu_2^2)at)$$

$$u(x, y, 0) = \cos(\mu_1 x) \cos(\mu_2 y)$$

Аналитическое решение:

$$U(x, y, t) = \cos(\mu_1 x) \cos(\mu_2 y) \exp(-(\mu_1^2 + \mu_2^2)at)$$

1) $\mu_1 = 1, \mu_2 = 1$

2) $\mu_1 = 2, \mu_2 = 1$

3) $\mu_1 = 1, \mu_2 = 2$

Используя схемы переменных направлений и дробных шагов, решить двумерную начально-краевую задачу для дифференциального уравнения параболического типа. В различные моменты времени вычислить погрешность численного решения путем сравнения результатов с приведенным в задании аналитическим решением $U(x, y, t)$. Исследовать зависимость погрешности от сеточных параметров τ, h_x, h_y .

In [1]: `import numpy as np`

In [2]: `a = 1`

```
def U(x, y, t, m1, m2):
    return np.cos(m1 * x) * np.cos(m2 * y) * np.exp(-(m1 ** 2 + m2 ** 2) * a * t)
```

```

* a * t)

def u0jk(m1, m2, y, t, j, k):
    return np.cos(m2 * y[j]) * np.exp(-(m1 ** 2 + m2 ** 2) * a * t[k])

def uNxjk(m1, m2, y, t, j, k):
    return (-1) ** m1 * np.cos(m2 * y[j]) * np.exp(-(m1 ** 2 + m2 ** 2)
* a * t[k])

def ui0k(m1, m2, x, t, i, k):
    return np.cos(m1 * x[i]) * np.exp(-(m1 ** 2 + m2 ** 2) * a * t[k])

def uiNyk(m1, m2, x, t, i, k):
    return (-1) ** m2 * np.cos(m1 * x[i]) * np.exp(-(m1 ** 2 + m2 ** 2)
* a * t[k])

def uij0(m1, m2, x, y, i, j):
    return np.cos(m1 * x[i]) * np.cos(m2 * y[j])

```

Метод переменных направлений

$$\frac{u_{ij}^{k+1/2} - u_{ij}^k}{\tau/2} = \frac{a}{h_1^2} (u_{i+1j}^{k+1/2} - 2u_{ij}^{k+1/2} + u_{i-1j}^{k+1/2}) + \frac{a}{h_2^2} (u_{ij+1}^k - 2u_{ij}^k + u_{ij-1}^k) + f_{ij}^{k+1/2},$$

(5.78)

$$\frac{u_{ij}^{k+1} - u_{ij}^{k+1/2}}{\tau/2} = \frac{a}{h_1^2} (u_{i+1j}^{k+1/2} - 2u_{ij}^{k+1/2} + u_{i-1j}^{k+1/2}) + \frac{a}{h_2^2} (u_{ij+1}^{k+1} - 2u_{ij}^{k+1} + u_{ij-1}^{k+1}) + f_{ij}^{k+1/2}.$$

Реализация

In [3]:

```

# метод прогонки
def tridig_matrix_alg(A, b):

    X = [0 for i in range(len(A[0]))]
    P = [0 for i in range(len(A[0]))]
    Q = [0 for i in range(len(A[0]))]
    P[0] = -A[0][1] / A[0][0]
    Q[0] = b[0] / A[0][0]

    for i in range(1, len(b)):
        if i != len(A[0]) - 1:
            P[i] = -A[i][i + 1] / (A[i][i] + P[i - 1] * A[i][i - 1])

```

```

        else:
            P[i] = 0
            Q[i] = (b[i] - Q[i - 1] * A[i][i - 1]) / (A[i][i] + P[i - 1] *
A[i][i - 1])
        for i in range(len(b) - 1, -1, -1):
            if i != len(A[0]) - 1:
                X[i] = X[i + 1] * P[i] + Q[i]
            else:
                X[i] = Q[i]
    return X

```

```

In [4]: def alternating_direction_method(T, Nx, Ny, K, m1, m2, lx=0, rx=np.pi,
ly=0, ry=np.pi):
    tau = T / K
    hx = (rx - lx) / Nx
    hy = (ry - ly) / Ny
    x = [lx + i * hx for i in range(Nx + 1)]
    y = [ly + j * hy for j in range(Ny + 1)]
    t = [k * tau / 2 for k in range(2 * K + 1)]
    u = []
    row_x = []
    for i in range(Nx + 1):
        row_y = []
        for j in range(Ny + 1):
            row_y.append(uij0(m1, m2, x, y, i, j))
        row_x.append(row_y)
    u.append(row_x)
    u = np.array(u)
    ax = a / hx ** 2
    ay = a / hy ** 2

    for k in range(0, 2 * K + 1 - 2, 2):
        u = np.append(u, [[[0] * (Ny + 1)] * (Nx + 1)], axis=0)
        for j in range(Ny + 1):
            u[k + 1][0][j] = u0jk(m1, m2, y, t, j, k + 1)
            u[k + 1][Nx][j] = uNxjk(m1, m2, y, t, j, k + 1)
        for i in range(Nx + 1):
            u[k + 1][i][0] = ui0k(m1, m2, x, t, i, k + 1)
            u[k + 1][i][Ny] = uiNyk(m1, m2, x, t, i, k + 1)

        for j in range(1, Ny):
            Ax = []
            bx = []

```

```

        for i in range(1, Nx):
            rows = []
            if i == 1:
                bx.append(- (ay * u[k][i][j - 1] + 2 * (1 / tau -
ay) * u[k][i][j] +\
                                ay * u[k][i][j + 1] + ax * u[k + 1][i -
1][j])) #
                rows = [ - 2 * (ax + 1 / tau) if (p == 1) else 0 for
p in range(1, Nx)] #
                rows[1] = ax
                Ax.append(rows)
                continue
            elif i == Nx - 1:
                bx.append(- (ay * u[k][i][j - 1] + 2 * (1 / tau -
ay) * u[k][i][j] +\
                                ay * u[k][i][j + 1] + ax * u[k + 1][i +
1][j])) #
                rows = [ - 2 * (ax + 1 / tau) if (p == Nx - 1) else
0 for p in range(1, Nx)]#
                rows[Nx - 3] = ax
                Ax.append(rows)
                continue
            else:
                bx.append(- (ay * u[k][i][j - 1] + 2 * (1/tau - ay)
* u[k][i][j] +\
                                ay * u[k][i][j+1])) #
                for l in range(1, Nx):
                    if (l == i - 1) | (l == i + 1):
                        rows.append(ax)
                    elif l == i:
                        rows.append(- 2 * (ax + 1 / tau))
                    else:
                        rows.append(0)
                Ax.append(rows)
            res = tridig_matrix_alg(Ax, bx)

        for i in range (1, Nx):
            u[k + 1][i][j] = res[i - 1]

u = np.append(u, [[[0] * (Ny + 1)] * (Nx + 1)], axis=0)
for j in range(Ny + 1):
    u[k + 2][0][j] = u0jk(m1, m2, y, t, j, k + 2)
    u[k + 2][Nx][j] = uNxjk(m1, m2, y, t, j, k + 2)

```

```

for i in range(Nx + 1):
    u[k + 2][i][0] = ui0k(m1, m2, x, t, i, k + 2)
    u[k + 2][i][Ny] = uiNyk(m1, m2, x, t, i, k + 2)

for i in range(1, Nx):
    Ay = []
    by = []
    for j in range(1, Ny):
        rows = []
        if j == 1:
            by.append( - (ax * u[k + 1][i - 1][j] + 2*(1/tau -
ax) * u[k + 1][i][j] + \
                        ax * u[k + 1][i+1][j] + ay * u[k+2][i][j-
1])) #
            rows = [ - 2 * (ay + 1 / tau) if (p == 1) else 0 for
p in range(1, Ny)]#
            rows[1] = ay
            Ay.append(rows)
            continue
        elif j == Ny - 1:
            by.append( - (ax * u[k + 1][i - 1][j] + 2*(1/tau -
ax) * u[k + 1][i][j] + \
                        ax * u[k + 1][i+1][j] + ay * u[k+2][i]
[j+1])) #
            rows = [ - 2 * (ay + 1 / tau) if (p == Ny -1) else 0
for p in range(1, Ny)] #
            rows[Ny - 3] = ay
            Ay.append(rows)
            continue
        else:
            by.append( - (ax * u[k + 1][i - 1][j] + 2*(1/tau -
ax) * u[k + 1][i][j] + \
                        ax * u[k + 1][i+1][j] )) # правка 3
    for l in range(1, Ny):
        if (l == j - 1) | (l == j + 1):
            rows.append(ay)
        elif l == j:
            rows.append(- 2 * (ay + 1 / tau))
        else:
            rows.append(0)
    Ay.append(rows)
res = tridig_matrix_alg(Ay, by)
for j in range (1, Ny):

```

```

        u[k + 2][i][j] = res[j - 1]
    return x, y, t, u

```

Тест

```

In [5]: def clean_u_t(u, t):
        new_u = np.array([u[k] for k in range(0, len(u), 2)])
        new_t = np.array([t[k] for k in range(0, len(t), 2)])
        return new_t, new_u

```

```

In [6]: m1 = 1
        m2 = 1
        x, y, t, u = alternating_direction_method(3, 10, 10, 3, m1, m2)

```

```

In [7]: t, u = clean_u_t(u, t)

```

Графики решения

```

In [8]: import matplotlib.pyplot as plt
        from mpl_toolkits.mplot3d import Axes3D
        %matplotlib inline

```

```

In [9]: def calc_U_2d(U, t, m1, m2, lx, rx, ly, ry, dx, dy):
        xs = np.arange(lx, rx * m1 + dx, dx)
        ys = np.arange(ly, ry * m2 + dy, dy)
        U_ = []

        for x_ in xs:
            row = []
            for y_ in ys:
                row.append(U(x_, y_, t, m1, m2))
            U_.append(row)
        return xs, ys, U_

```

```

In [10]: def calc_Us_2d(k, u):
        return u[k]

```

```

In [11]: def plot_solution(x, y, t, m1, m2, u, U, lx, rx, ly, ry):
        fig, axes = plt.subplots(2, 2)
        fig.set_figheight(10)
        fig.set_figwidth(15)

```

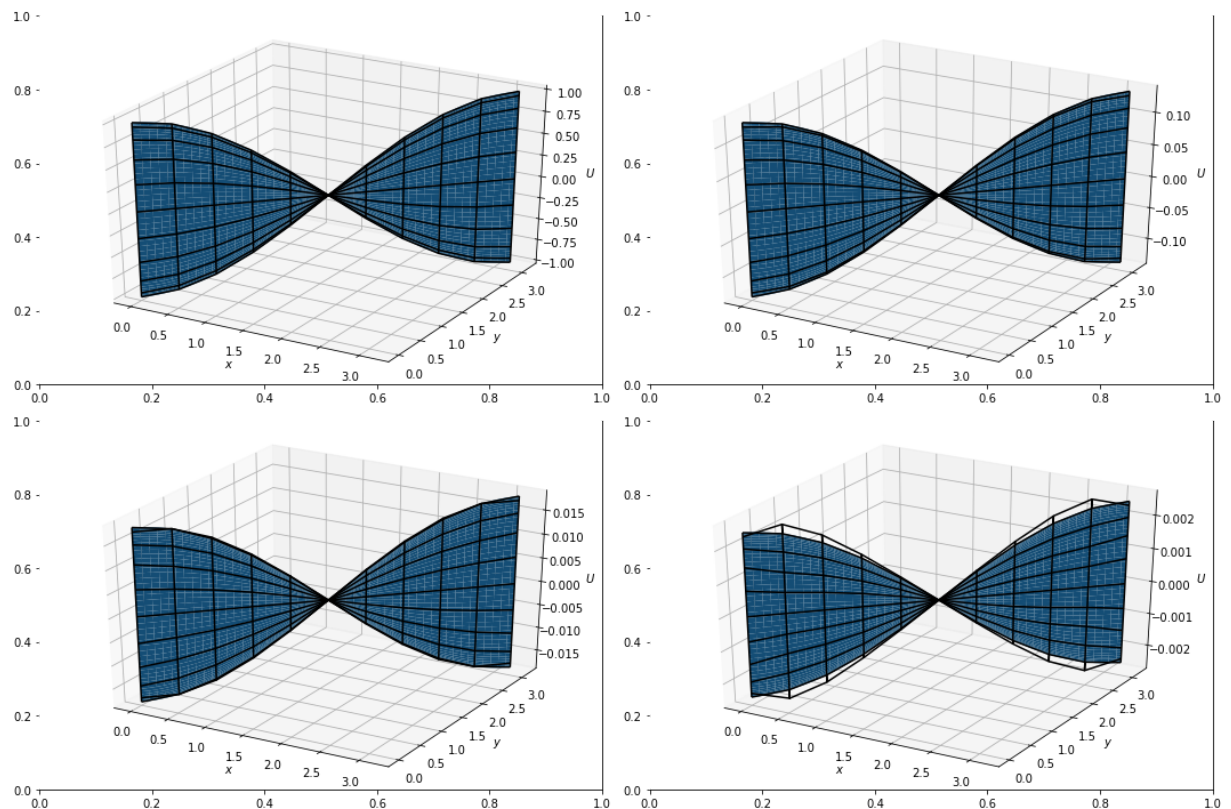
```

dx = (rx * m1 - lx) / 1000
dy = (ry * m2 - ly) / 1000
dt = len(t) // 4
for k in range(4):
    uelist = calc_Us_2d(k * dt, u)
    xarr, yarr, Ulist = calc_U_2d(U, k * dt, m1, m2, lx, rx, ly,
    ry, dx, dy)

    ax = fig.add_subplot(2, 2, k + 1, projection='3d')
    ax.plot_surface(np.array(xarr), np.array(yarr),
    np.array(Ulist))
    ax.plot_wireframe(x, y, uelist, color="black")
    ax.set(xlabel='$x$', ylabel='$y$', zlabel='$U$')
    fig.tight_layout()

```

In [12]: `plot_solution(x, y, t, m1, m2, u, U, 0, np.pi, 0, np.pi)`



Оценка погрешности

MSE

```

In [13]: def MSE(x, y, t, u, U, m1, m2):
    s = 0
    for k, t_ in enumerate(t):
        for i, x_ in enumerate(x):
            for j, y_ in enumerate(y):

```

```
s += (U(x_, y_, t_, m1, m2) - u[k][i][j]) ** 2  
return np.sqrt(s)
```

```
In [14]: print("MSE = {}".format(MSE(x, y, t, u, U, m1, m2)))
```

MSE = 0.010349404217681588

Графики погрешности

```
In [15]: # ошибки по x  
def errors_x(x, y, t, u, U, m1, m2):  
    errors = []  
    for i, x_ in enumerate(x):  
        err = 0  
        for k, t_ in enumerate(t):  
            for j, y_ in enumerate(y):  
                err += (U(x_, y_, t_, m1, m2) - u[k][i][j]) ** 2  
            errors.append(err ** 0.5)  
    return errors  
  
# ошибки по y  
def errors_y(x, y, t, u, U, m1, m2):  
    errors = []  
    for j, y_ in enumerate(y):  
        err = 0  
        for k, t_ in enumerate(t):  
            for i, x_ in enumerate(x):  
                err += (U(x_, y_, t_, m1, m2) - u[k][i][j]) ** 2  
            errors.append(err ** 0.5)  
    return errors  
  
# ошибки по t  
def errors_t(x, y, t, u, U, m1, m2):  
    errors = []  
    for k, t_ in enumerate(t):  
        err = 0  
        for i, x_ in enumerate(x):  
            for j, y_ in enumerate(y):  
                err += (U(x_, y_, t_, m1, m2) - u[k][i][j]) ** 2  
            errors.append(err ** 0.5)  
    return errors  
  
# функция отрисовки графиков ошибки по x  
def plot_errors_x(x, y, t, u, U, m1, m2):
```

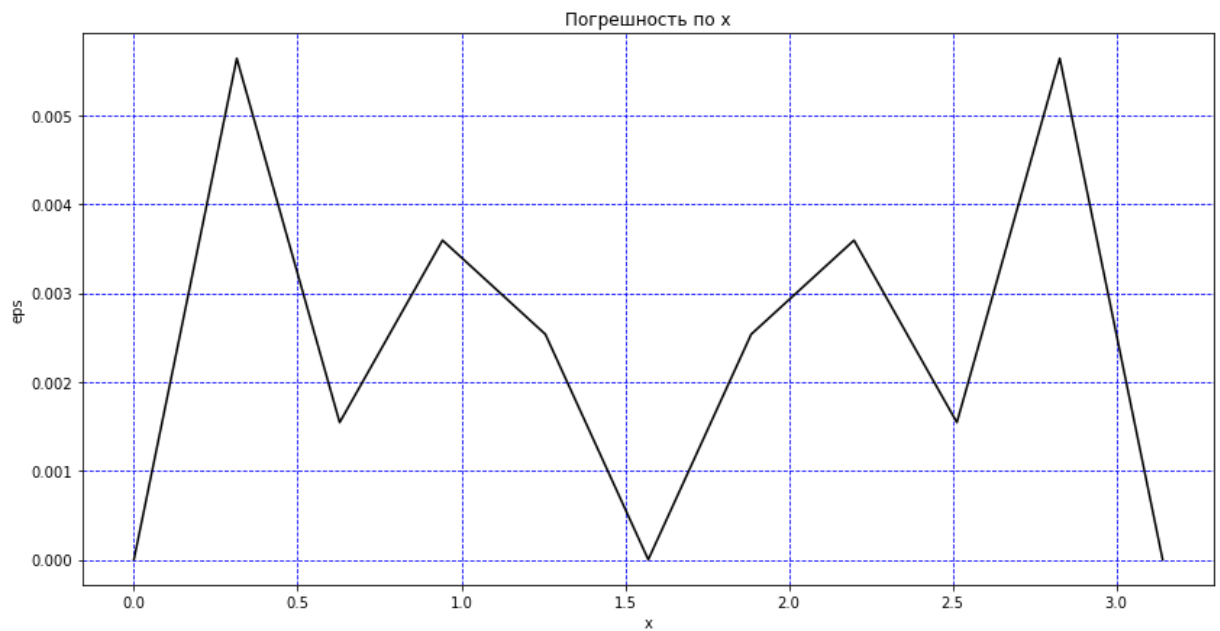


```
plt.figure(figsize=(14,7))
# погрешность по x
plt.plot(x, errors_x(x, y, t, u, U, m1, m2), color = 'black')
# отрисовка координатной сетки
plt.grid(color = 'blue', linestyle = '--')
# легенда
plt.xlabel('x')
plt.ylabel('eps')
plt.title(f'Погрешность по x')
plt.show()

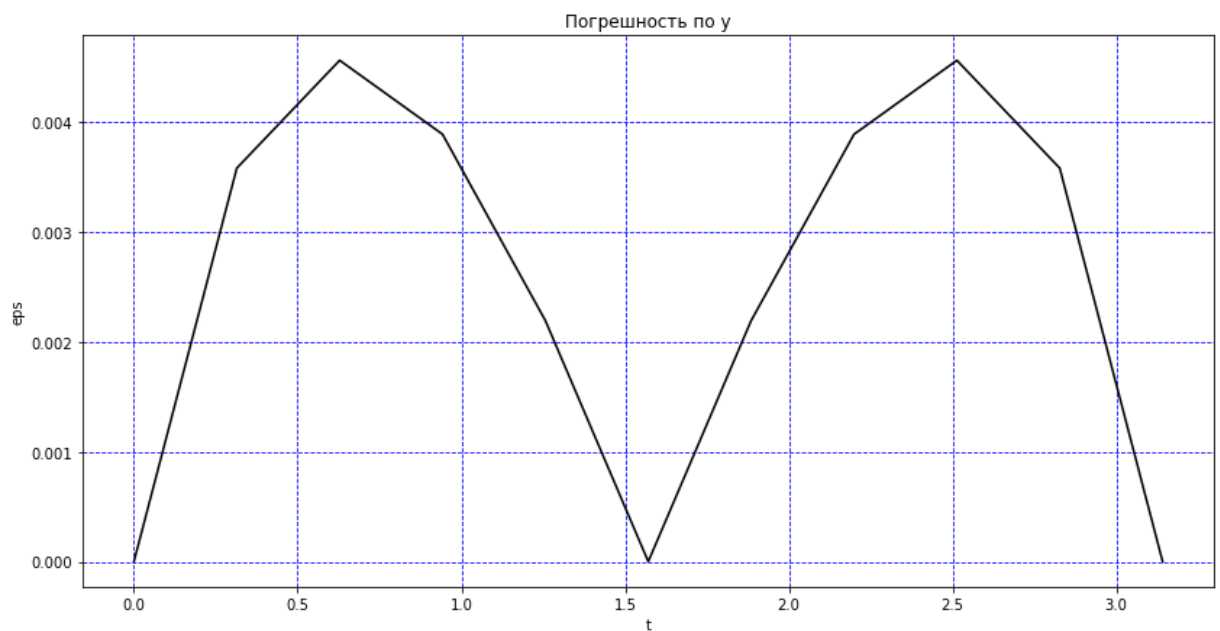
# функция отрисовки графиков ошибки по t
def plot_errors_t(x, y, t, u, U, m1, m2):
    plt.figure(figsize=(14,7))
    # погрешность по t
    plt.plot(t, errors_t(x, y, t, u, U, m1, m2), color = 'black')
    # отрисовка координатной сетки
    plt.grid(color = 'blue', linestyle = '--')
    # легенда
    plt.xlabel('t')
    plt.ylabel('eps')
    plt.title(f'Погрешность по t')
    plt.show()

# функция отрисовки графиков ошибки по t
def plot_errors_y(x, y, t, u, U, m1, m2):
    plt.figure(figsize=(14,7))
    # погрешность по t
    plt.plot(y, errors_y(x, y, t, u, U, m1, m2), color = 'black')
    # отрисовка координатной сетки
    plt.grid(color = 'blue', linestyle = '--')
    # легенда
    plt.xlabel('t')
    plt.ylabel('eps')
    plt.title(f'Погрешность по y')
    plt.show()
```

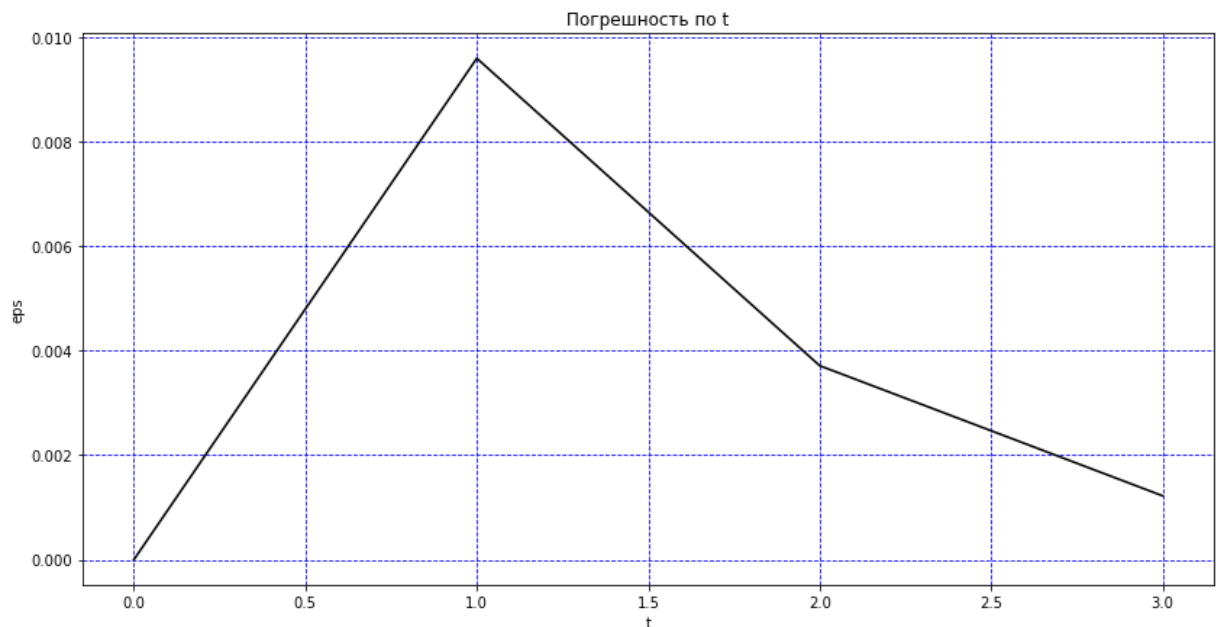
In [16]: `plot_errors_x(x, y, t, u, U, m1, m2)`



In [17]: `plot_errors_y(x, y, t, u, U, m1, m2)`



In [18]: `plot_errors_t(x, y, t, u, U, m1, m2)`



Метод дробных шагов

$$\frac{u_{ij}^{k+1/2} - u_{ij}^k}{\tau} = \frac{a}{h_1^2} (u_{i+1j}^{k+1/2} - 2u_{ij}^{k+1/2} + u_{i-1j}^{k+1/2}) + \frac{f_{ij}^k}{2},$$

$$\frac{u_{ij}^{k+1} - u_{ij}^{k+1/2}}{\tau} = \frac{a}{h_2^2} (u_{ij+1}^{k+1} - 2u_{ij}^{k+1} + u_{ij-1}^{k+1}) + \frac{f_{ij}^{k+1}}{2}.$$

Реализация

```
In [19]: def fractional_step_method(T, Nx, Ny, K, m1, m2, lx=0, rx=np.pi, ly=0,
ry=np.pi):
    rx = rx * m1
    ry = ry * m2
    tau = T / K
    hx = (rx - lx) / Nx
    hy = (ry - ly) / Ny
    x = [lx + i * hx for i in range(Nx + 1)]
    y = [ly + j * hy for j in range(Ny + 1)]
    t = [k * tau / 2 for k in range(2 * K + 1)]
    u = []

    row_x = []
    for i in range(Nx + 1):
        row_y = []
        for j in range(Ny + 1):
            row_y.append(uij0(m1, m2, x, y, i, j))
        row_x.append(row_y)
    u.append(row_x)
```

```

u = np.array(u)
ax = a / hx ** 2
ay = a / hy ** 2

for k in range(0, 2 * K + 1 - 2, 2):
    u = np.append(u, [[[0] * (Ny + 1)] * (Nx + 1)], axis=0)

    for j in range(Ny + 1):
        u[k + 1][0][j] = u0jk(m1, m2, y, t, j, k + 1)
        u[k + 1][Nx][j] = uNxjk(m1, m2, y, t, j, k + 1)
    for i in range(Nx + 1):
        u[k + 1][i][0] = ui0k(m1, m2, x, t, i, k + 1)
        u[k + 1][i][Ny] = uiNyk(m1, m2, x, t, i, k + 1)
    for j in range(1, Ny):
        Ax = []
        bx = []
        for i in range(1, Nx):
            rows = []
            if i == 1:
                bx.append( - (u[k][i][j] / tau + ax * u[k + 1][i - 1]
[j])) #
                rows = [ - (2 * ax + 1 / tau) if (p == 1) else 0 for
p in range(1, Ny)]#
                rows[1] = ax
                Ax.append(rows)
                continue
            elif i == Nx - 1:
                bx.append( -(u[k][i][j] / tau + ax * u[k + 1][i + 1]
[j])) #
                rows = [ - (2 * ax + 1 / tau) if (p == Nx - 1) else 0
for p in range(1, Ny)] #
                rows[Nx - 3] = ax
                Ax.append(rows)
                continue
            else:
                bx.append( - u[k][i][j] / tau)
        for l in range(1, Nx):
            if (l == i - 1) | (l == i + 1):
                rows.append(ax)
            elif l == i:
                rows.append(- (2 * ax + 1 / tau))
            else:

```

```

        rows.append(0)
    Ax.append(rows)
    res = tridig_matrix_alg(Ax, bx)
    for i in range(1, Nx):
        u[k + 1][i][j] = res[i - 1]

u = np.append(u, [[[0] * (Ny + 1)] * (Nx + 1)], axis=0)
for j in range(Ny + 1):
    u[k + 2][0][j] = u0jk(m1, m2, y, t, j, k + 2)
    u[k + 2][Nx][j] = uNxjk(m1, m2, y, t, j, k + 2)
    for i in range(Nx + 1):
        u[k + 2][i][0] = ui0k(m1, m2, x, t, i, k + 2)
        u[k + 2][i][Ny] = uiNyk(m1, m2, x, t, i, k + 2)
    for i in range(1, Nx):
        Ay = []
        by = []
        for j in range(1, Ny):
            rows = []
            if j == 1:
                by.append( - (u[k + 1][i][j] / tau + ay * u[k+2][i]
[j-1])) #
                rows = [ - (2 * ay + 1 / tau) if (p == 1) else 0 for
p in range(1, Ny)] #
                rows[1] = ay
                Ay.append(rows)
                continue
            elif j == Ny - 1:
                by.append( - (u[k + 1][i][j] / tau + ay * u[k+2][i]
[j+1]))#
                rows = [ - (2 * ay + 1 / tau) if (p == Ny - 1)else 0
for p in range(1, Ny)]#
                rows[Ny - 3] = ay
                Ay.append(rows)
                continue
            else:
                by.append( - u[k + 1][i][j] / tau)
        for l in range(1, Ny):
            if (l == j - 1) | (l == j + 1):
                rows.append(ay)
            elif l == j:
                rows.append(- (2 * ay + 1 / tau))
            else:

```

```

        rows.append(0)
        Ay.append(rows)
        res = tridig_matrix_alg(Ay, by)
        for j in range(1, Ny):
            u[k + 2][i][j] = res[j - 1]
        return x, y, t, u

```

Тест

```

In [20]: m1 = 1
         m2 = 1
         x, y, t, u = fractional_step_method(3, 10, 10, 3, m1, m2)

```

```

In [21]: t, u = clean_u_t(u, t)

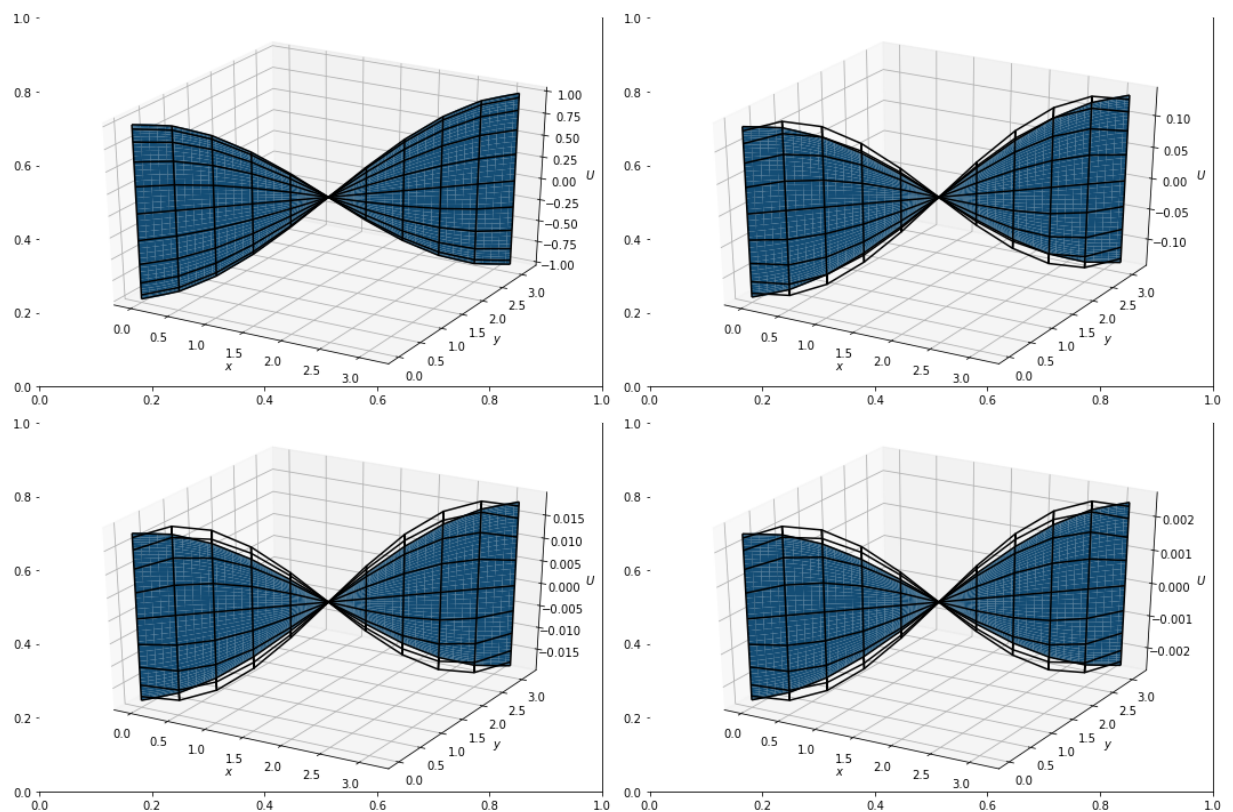
```

Графики решения

```

In [22]: plot_solution(x, y, t, m1, m2, u, U, 0, np.pi, 0, np.pi)

```



Оценка погрешности

MSE

```

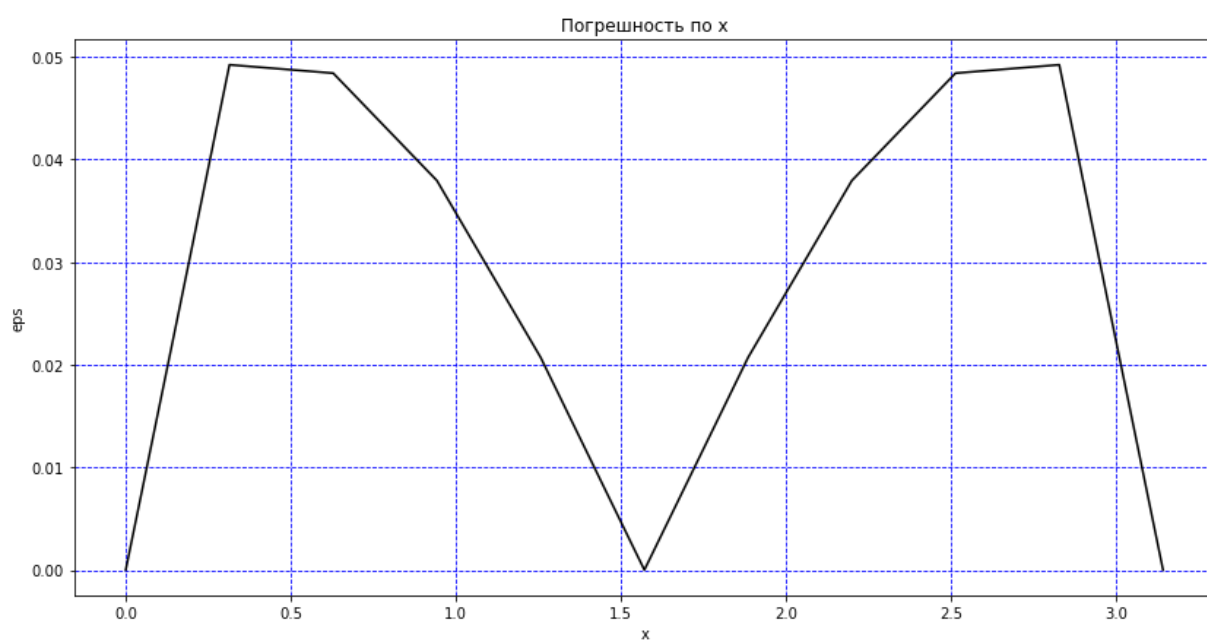
In [23]: print("MSE = {}".format(MSE(x, y, t, u, U, m1, m2)))

```

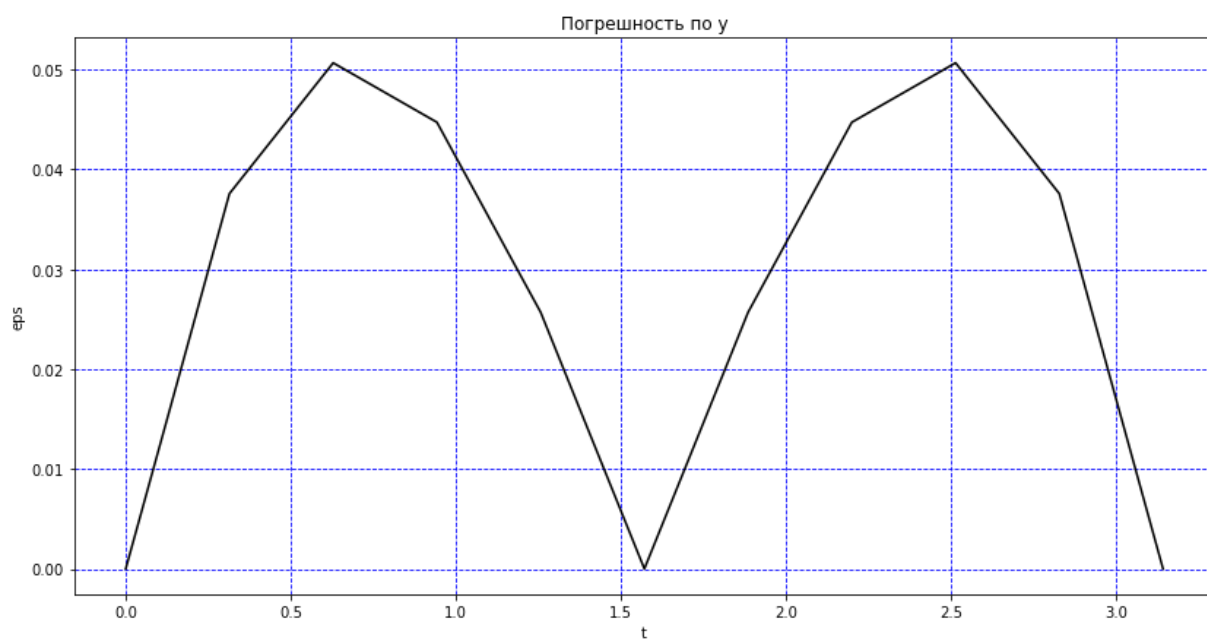
MSE = 0.11522428415985596

Графики погрешности

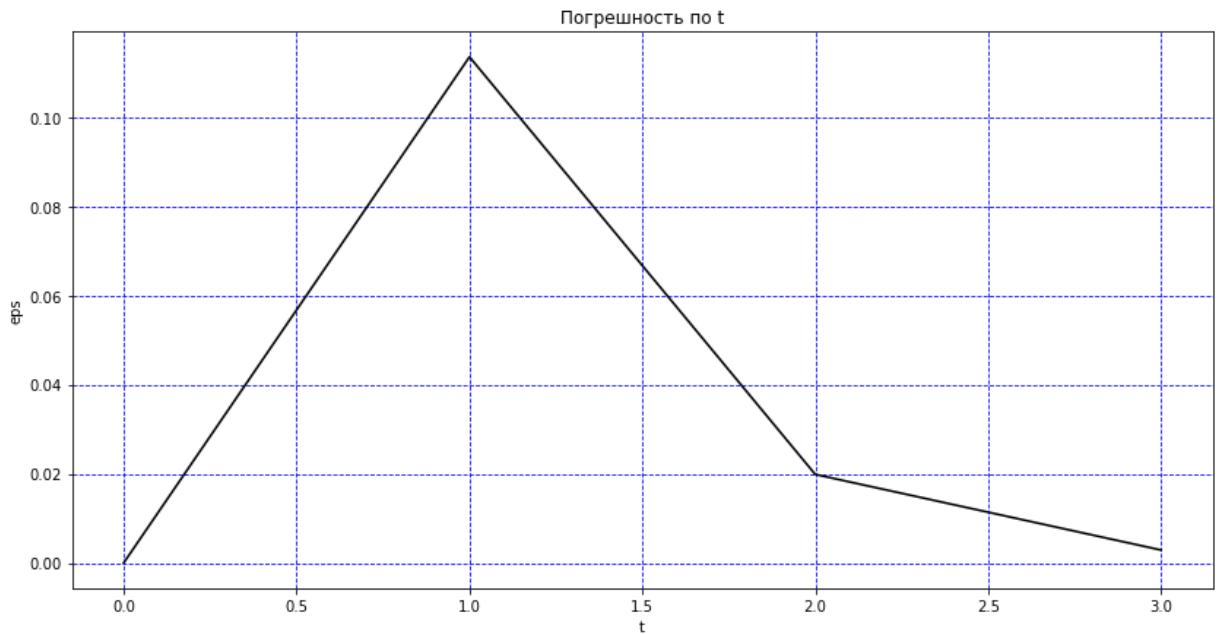
In [24]: `plot_errors_x(x, y, t, u, U, m1, m2)`



In [25]: `plot_errors_y(x, y, t, u, U, m1, m2)`



In [26]: `plot_errors_t(x, y, t, u, U, m1, m2)`



Вывод

В результате выполнения лабораторной работы были освоены две схемы для решения двумерной начально-краевой задачи для дифференциального уравнения параболического типа : метод переменных направлений и метод дробных шагов.

Метод переменных направлений показал лучший результат, чем метод дробных шагов, это можно заметить как по графикам решения, так и по графикам погрешности и величине средне-квадратичной ошибки. Стоит отметить, что метод переменных направлений условно устойчив при увеличении размерности пространства, а метод дробных шагов абсолютно устойчив.