# **Арсланов Тимур Маратович М80-402Б-20**

# Лабораторная работа №8 по курсу Численные методы

Москва 2023

# 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 T,  $h_x$ ,  $h_y$ .

```
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)
```

```
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

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

$$(5.78)$$

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

### 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 - 1)
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 1 in range(1, Nx):
                    if (l == i - 1) | (l == i + 1):
                        rows.append(ax)
                    elif 1 == 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 - 1)[j]
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 - 1)
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 - 1)
ax) * u[k + 1][i][j] + 
                                   ax * u[k + 1][i+1][j] )) # npa6\kappa a 3
                for 1 in range(1, Ny):
                    if (l == j - 1) | (l == j + 1):
                        rows.append(ay)
                    elif 1 == 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 [7]: [t, u = clean_u_t(u, t)
```

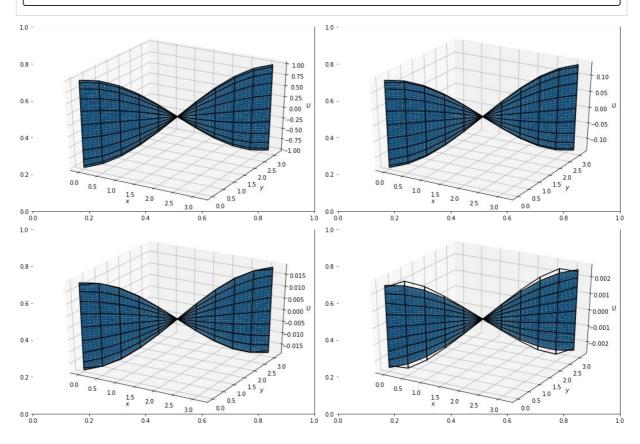
### Графики решения

```
import matplotlib.pyplot as plt
from mpl_toolkits.mplot3d import Axes3D
%matplotlib inline
```

```
dx = (rx * m1 - lx) / 1000
dy = (ry * m2 - ly) / 1000
dt = len(t) // 4
for k in range(4):
    ulist = 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, ulist, 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)
```

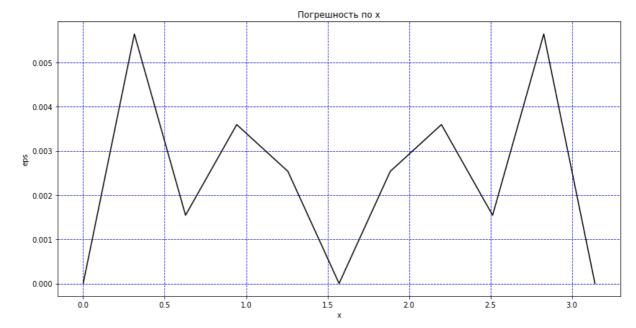
MSE = 0.010349404217681588

#### Графики погрешности

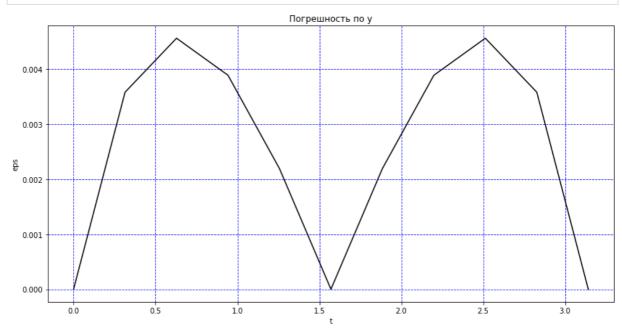
```
In [15]:
        \# ошибки по х
         def errors_x(x, y, t, u, U, m1, m2):
             errors= []
             for i, x_ in enumerate(x):
                  err = 0
                  for k, t_ in enumerate(t):
                      for j, y_ in enumerate(y):
                          err += (U(x_,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

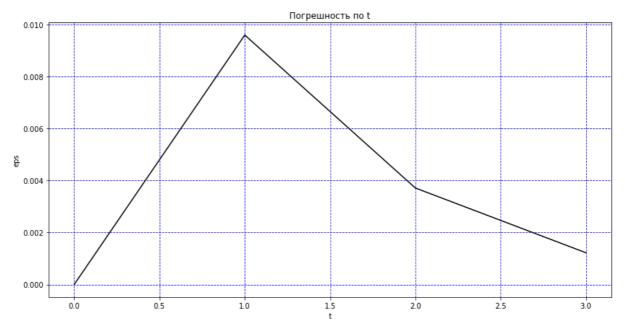
```
plt.figure(figsize=(14,7))
    # погрешность по х
    plt.plot(x, errors_x(x, y, t, u, U, m1, m2), color = 'black')
    # отрисовка координатной сетки
    plt.grid(color = 'blue', linestyle = '--')
    # легенда
    plt.xlabel('x')
    plt.ylabel('eps')
    plt.title(f'Погрешность по x')
    plt.show()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





# Метод дробных шагов

$$\begin{split} \frac{u_{ij}^{k+1/2} - u_{ij}^k}{\tau} &= \frac{a}{h_1^2} \left( u_{i+1j}^{k+1/2} - 2u_{ij}^{k+1/2} + u_{i-1j}^{k+1/2} \right) + \frac{f_{ij}^k}{2} \ , \\ \frac{u_{ij}^{k+1} - u_{ij}^{k+1/2}}{\tau} &= \frac{a}{h_2^2} \left( u_{ij+1}^{k+1} - 2u_{ij}^{k+1} + u_{ij-1}^{k+1} \right) + \frac{f_{ij}^{k+1}}{2} \ . \end{split}$$

# 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)
```

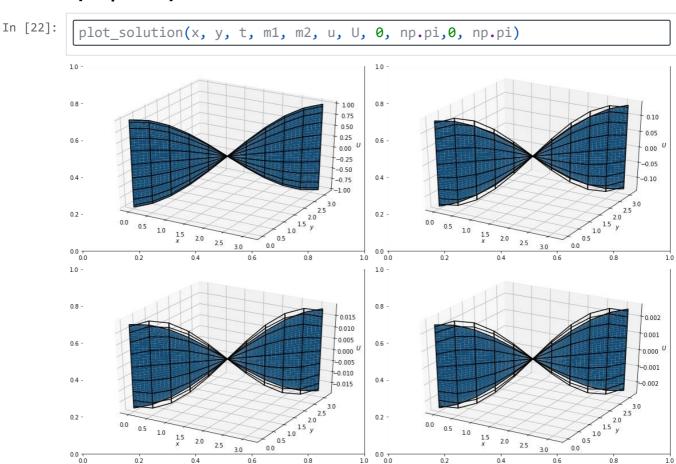
```
u = np.array(u)
    ax = a / hx ** 2
    ay = a / hy ** 2
    for k in range(0, 2 * K + 1 - 2, 2):
        u = np.append(u, [[[0] * (Ny + 1)] * (Nx + 1)], axis=0)
        for j in range(Ny + 1):
            u[k + 1][0][j] = u0jk(m1, m2, y, t, j, k + 1)
            u[k + 1][Nx][j] = uNxjk(m1, m2, 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 1 in range(1, Nx):
                    if (l == i - 1) | (l == i + 1):
                        rows.append(ax)
                    elif 1 == 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 1 in range(1, Ny):
                    if (1 == j - 1) | (1 == j + 1):
                        rows.append(ay)
                    elif 1 == j:
                        rows.append(-(2 * ay + 1 / tau))
                    else:
```

#### Тест

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

### Графики решения



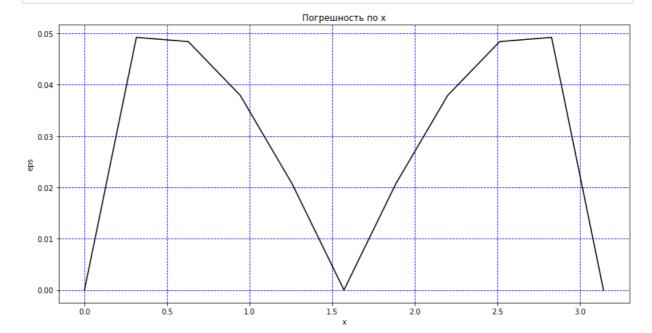
### Оценка погрешности

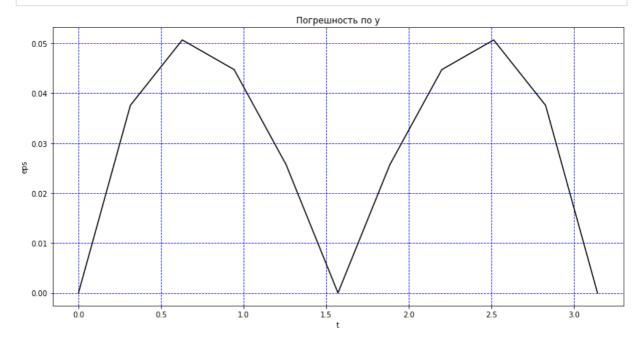
#### **MSE**

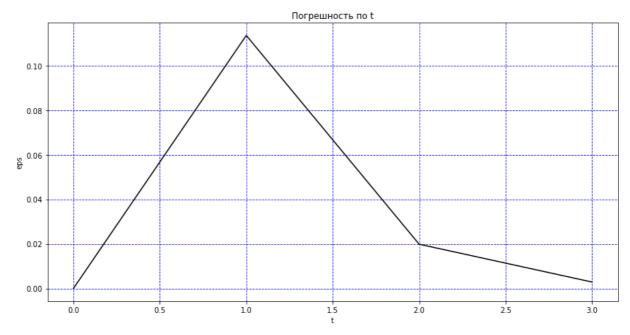
MSE = 0.11522428415985596

### Графики погрешности









# Вывод

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

Метод переменных направлений показал лучший резлуьтат, 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 увелечении размерности пространства, а метод дробных шагов абсолютно устойчив.