

# Лабораторная работа №5

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Используя явную и неявную конечно-разностные схемы, а также схему Кранка - Николсона, решить начально-краевую задачу для дифференциального уравнения параболического типа. Осуществить реализацию трех вариантов аппроксимации граничных условий, содержащих производные: двухточечная аппроксимация с первым порядком, трехточечная аппроксимация со вторым порядком, двухточечная аппроксимация со вторым порядком. В различные моменты времени вычислить погрешность численного решения путем сравнения результатов с приведенным в задании аналитическим решением  $U(x, t)$ . Исследовать зависимость погрешности от сеточных параметров  $\tau, h$ .

## Вариант 6

$$\frac{\partial u}{\partial t} = \frac{\partial^2 u}{\partial x^2} + \cos x(\cos t + \sin t)$$

$$u(0, t) = \sin t$$

$$u_x\left(\frac{\pi}{2}, t\right) = -\sin t$$

$$u(x, 0) = 0$$

$$U(x, t) = \sin t \cos x$$

```
In [22]: import numpy as np
import matplotlib.pyplot as plt
```

```
In [23]: eq = {
    'l': np.pi / 2,
    'psi': lambda x: 0,
    'f': lambda x, t: np.cos(x) * (np.cos(t) + np.sin(t)),
    'phi0': lambda t: np.sin(t),
    'phi1': lambda t: -np.sin(t),
    'solution': lambda x, t: np.sin(t) * np.cos(x),
}
```

```
In [24]: def tma(a, b, c, d):
    size = len(a)
    p, q = [], []
    p.append(-c[0] / b[0])
    q.append(d[0] / b[0])
    for i in range(1, size):
        p_tmp = -c[i] / (b[i] + a[i] * p[i - 1])
        q_tmp = (d[i] - a[i] * q[i - 1]) / (b[i] + a[i] * p[i - 1])
        p.append(p_tmp)
        q.append(q_tmp)
    x = [0 for _ in range(size)]
    x[size - 1] = q[size - 1]
    for i in range(size - 2, -1, -1):
        x[i] = p[i] * x[i + 1] + q[i]
    return x

def get_zeros(N, K):
    lst = [np.zeros(N) for _ in range(0, 4)]
    lst.append(np.zeros((K, N)))
    return lst
```

```
In [25]: class Eq:
    def __init__(self, args):
        self.l = args['l']
        self.f = args['f']
        self.psi = args['psi']
        self.phi0 = args['phi0']
        self.phi1 = args['phi1']
        self.bound_type = args['bound_type']
        self.solve = args['solution']

class Parabolic:
    def __init__(self, args, N, K, T):
        self.alpha = 0
        self.beta = 1
        self.gamma = 1
        self.delta = 0
        self.data = Eq(args)
        self.a = 1
        self.b = 0
        self.c = 0
        self.h = self.data.l / N
        self.tau = T / K
        self.sigma = self.a ** 2 * self.tau / (self.h ** 2)

    def solve_analytic(self, N, K, T):
        self.h = self.data.l / N
        self.tau = T / K
        u = np.zeros((K, N))
        for i in range(K):
            for j in range(N):
                u[i][j] = self.data.solve(j * self.h, i * self.tau)
        return u
```

```

def calc(self, a, b, c, d, u, k, N, l, K):
    t = np.arange(0, T, T / K)
    for j in range(1, N):
        a[j] = self.sigma
        b[j] = -(1 + 2 * self.sigma)
        c[j] = self.sigma
        d[j] = -u[k][j]

    if self.data.bound_type == 'a1p1':
        a[0] = 0
        b[0] = self.beta - (self.alpha / self.h)
        c[0] = self.alpha / self.h
        d[0] = self.data.phi0(t[k]) / (self.beta - self.alpha /
a[-1] = -self.gamma / self.h
        b[-1] = self.gamma / self.h + self.delta
        c[-1] = 0
        d[-1] = self.data.phi1(t[k]) / (self.gamma / self.h + s
    elif self.data.bound_type == 'a1p2':
        a[0] = 0
        b[0] = -(1 + 2 * self.sigma)
        c[0] = self.sigma
        d[0] = -(u[k - 1][0] + self.sigma * self.data.phi0(k *
                self.tau * self.data.f(0, k * self.tau)
        a[-1] = self.sigma
        b[-1] = -(1 + 2 * self.sigma)
        c[-1] = 0
        d[-1] = -(u[k - 1][-1] + self.sigma * self.data.phi1(k *
                self.tau * self.data.f((N - 1) * self.h, k * se
    elif self.data.bound_type == 'a1p3':
        a[0] = 0
        b[0] = -(1 + 2 * self.sigma)
        c[0] = self.sigma
        d[0] = -((1 - self.sigma) * u[k - 1][1] + self.sigma /
                * self.data.f(0, k * self.tau) - self.sigma * se
                k * self.tau)
        a[-1] = self.sigma
        b[-1] = -(1 + 2 * self.sigma)
        c[-1] = 0
        d[-1] = self.data.phi1(k * self.tau) + self.data.f((N -
                * self.h / (2 * self.tau) * u[k - 1][-1]

def solve_implicit(self, N, K, T):
    lst = get_zeros(N, K)
    a = lst[0]
    b = lst[1]
    c = lst[2]
    d = lst[3]
    u = lst[4]

    for i in range(1, N - 1):
        u[0][i] = self.data.psi(i * self.h)
    u[0][-1] = 0

    for k in range(1, K):
        self.calc(a, b, c, d, u, k, N, T, K)
        u[k] = tma(a, b, c, d)

    return u

```

```
def solve_explicit(self, N, K, T):
    u = np.zeros((K, N))
    t = np.arange(0, T, T / K)
    x = np.arange(0, np.pi / 2, np.pi / 2 / N)
    for j in range(1, N - 1):
        u[0][j] = self.data.psi(j * self.h)

    for k in range(1, K):
        for j in range(1, N - 1):
            u[k][j] = (u[k - 1][j + 1] * (self.a ** 2.0 * self.tau / self.h ** 2)
                        - 2 * u[k - 1][j] * (self.a ** 2.0 * self.tau / self.h ** 2)
                        + u[k - 1][j - 1] * (self.a ** 2.0 * self.tau / self.h ** 2)
                        + u[k - 1][j]
                        + self.tau * self.data.f(x[j], t[k]))

        if self.data.bound_type == 'a1p1':
            u[k][0] = self.data.phi0(t[k])
            u[k][-1] = (self.data.phi1(t[k]) + self.gamma / self.tau)
        elif self.data.bound_type == 'a1p2':
            u[k][0] = self.data.phi0(t[k])
            u[k][-1] = (((2.0 * self.gamma * self.a / self.h /
                           (self.gamma * self.h / self.tau / (2.0 * self.data.l, t[k])
                           + self.data.phi1(t[k])
                           (2.0 * self.gamma * self.a / self.h /
                            self.gamma * self.h / self.tau
                            self.gamma * self.h * self.tau
                            2.0 * self.a + self.h *
                           )
                           )
                           )
                           )
        elif self.data.bound_type == 'a1p3':
            u[k][0] = self.data.phi0(t[k])
            u[k][-1] = (self.data.phi1(k * self.tau) + u[k][-2]
                         (1 / self.h + 2 * self.tau / self.h))

    return u

def solve_crank_nicolson(self, N, K, T):
    theta = 0.5
    lst = get_zeros(N, K)
    a = lst[0]
    b = lst[1]
    c = lst[2]
    d = lst[3]
    u = lst[4]
    for i in range(1, N - 1):
        u[0][i] = self.data.psi(i * self.h)

    for k in range(1, K):
        self.calc(a, b, c, d, u, k, N, T, K)

        tmp_imp = tma(a, b, c, d)

        tmp_exp = np.zeros(N)
        tmp_exp[0] = self.data.phi0(self.tau)
        for j in range(1, N - 1):
            tmp_exp[j] = self.sigma * u[k - 1][j + 1] + (1 - 2 * self.sigma) * u[k - 1][j] + self.tau * self.data.f(x[j], t[k])
        tmp_exp[-1] = self.data.phi1(self.tau)
```

```
        for j in range(N):  
            u[k][j] = theta * tmp_imp[j] + (1 - theta) * tmp_ex  
  
    return u
```

In [26]:

```

def compare_error(dict_):
    error = [[abs(i - j) for i, j in zip(x, y)] for x, y in zip(dict_.values(), dict_.values())]
    return error

def show(dict_, time=0):
    fig = plt.figure()
    plt.title('Линии уровня')
    plt.plot(dict_['implicit'][time], color='r', label='implicit')
    plt.plot(dict_['explicit'][time], color='b', label='explicit')
    plt.plot(dict_['crank_nicolson'][time], color='y', label='crank_nicolson')
    plt.plot(dict_['analytic'][time], color='g', label='analytic')
    plt.legend(loc='best')
    plt.ylabel('U')
    plt.xlabel('number')
    plt.show()

    plt.title('Погрешность explicit')
    plt.plot(abs(dict_['explicit'][time] - dict_['analytic'][time]), color='g', label='error')
    plt.ylabel('Err')
    plt.xlabel('t')
    plt.show()

    plt.title('Погрешность implicit')
    plt.plot(abs(dict_['implicit'][time] - dict_['analytic'][time]), color='g', label='error')
    plt.ylabel('Err')
    plt.xlabel('t')
    plt.show()

    plt.title('Погрешность crank_nicolson')
    plt.plot(abs(dict_['crank_nicolson'][time] - dict_['analytic'][time]), color='g', label='error')
    plt.ylabel('Err')
    plt.xlabel('t')
    plt.show()

```

```

N, K, T = 12, 10000, 18

```

```

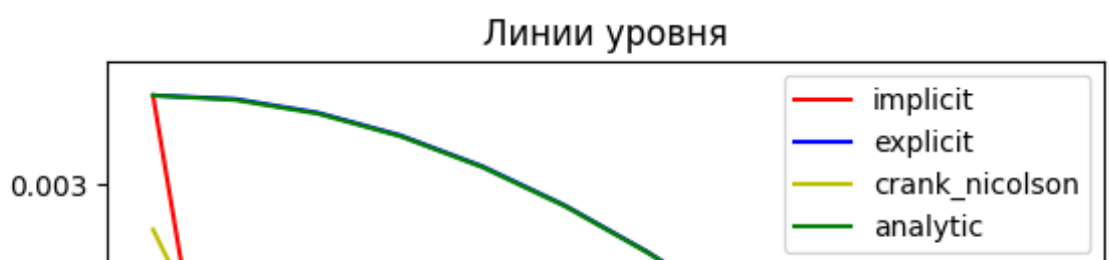
args = eq
args['bound_type'] = 'a1p1'

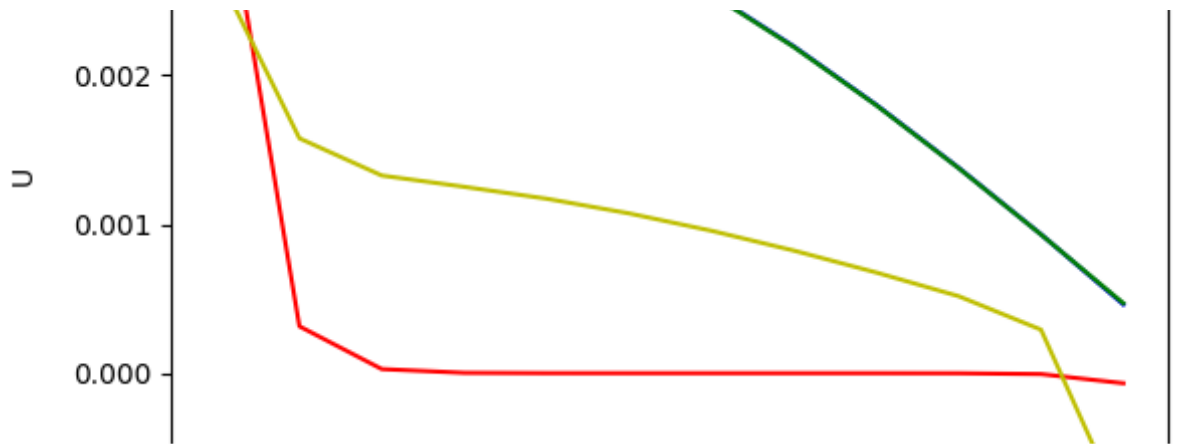
```

```

solver = Parabolic(args, N, K, T)
dict_ans = {
    'implicit': solver.solve_implicit(N, K, T),
    'explicit': solver.solve_explicit(N, K, T),
    'crank_nicolson': solver.solve_crank_nicolson(N, K, T),
    'analytic': solver.solve_analytic(N, K, T)
}
show(dict_ans, 2)

```

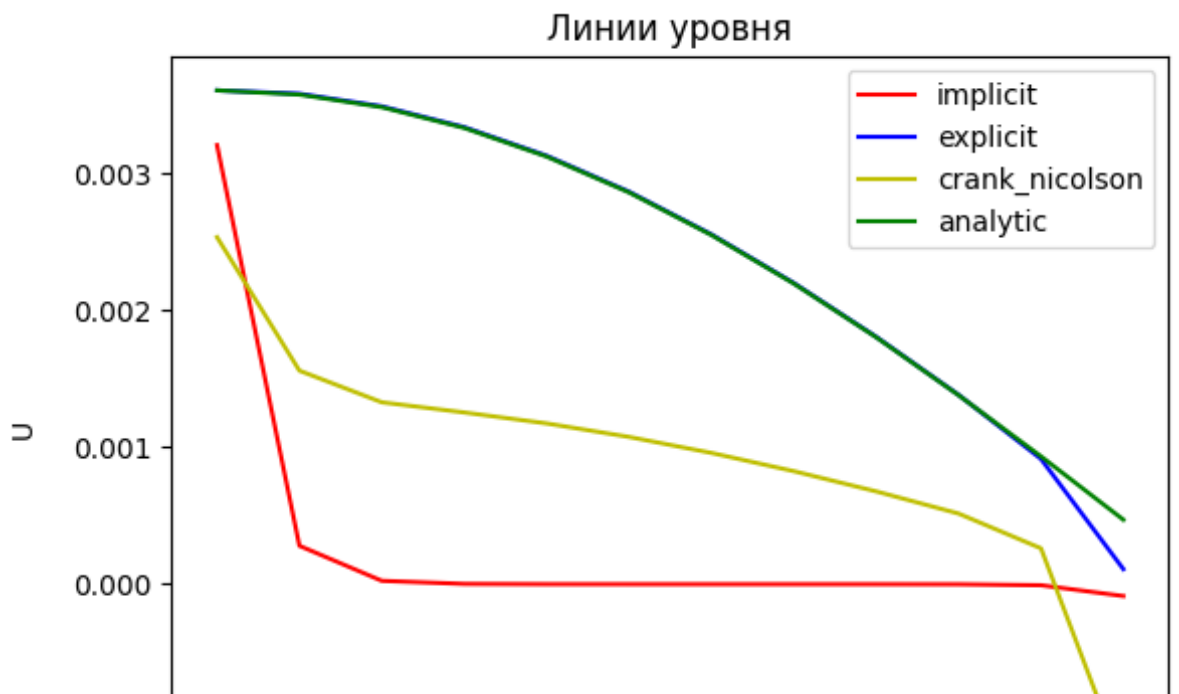




## Аппроксимация 3-х точечная второго порядка

In [27]: `N, K, T = 12, 10000, 18`

```
args = eq
args['bound_type'] = 'a1p2'
solver = Parabolic(args, N, K, T)
dict_ans = {
    'implicit': solver.solve_implicit(N, K, T),
    'explicit': solver.solve_explicit(N, K, T),
    'crank_nicolson': solver.solve_crank_nicolson(N, K, T),
    'analytic': solver.solve_analytic(N, K, T)
}
show(dict_ans, 2)
```



## Аппроксимация 2-х точечная 2 порядка

```
In [28]: N, K, T = 12, 10000, 18
```

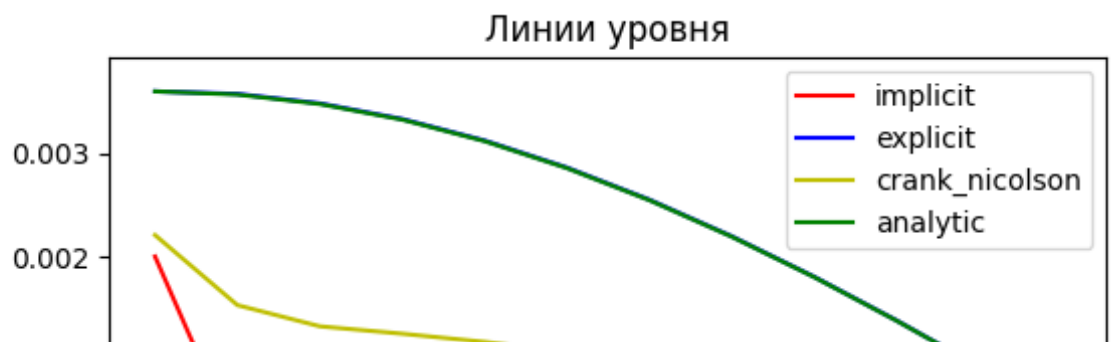
```
args = eq
args['bound_type'] = 'a1p3'
solver = Parabolic(args, N, K, T)
dict_ans = {
    'implicit': solver.solve_implicit(N, K, T),
    'explicit': solver.solve_explicit(N, K, T),
    'crank_nicolson': solver.solve_crank_nicolson(N, K, T),
    'analytic': solver.solve_analytic(N, K, T)
}
show(dict_ans, 2)
```

```
/var/folders/83/2lkmbrc10y3bnb04s4tnnht80000gn/T/ipykernel_23060/301663272.py:73: RuntimeWarning: overflow encountered in double_scalars
```

```
    d[-1] = self.data.phi1(k * self.tau) + self.data.f((N - 1) * self.h, k * self.tau) \
```

```
/var/folders/83/2lkmbrc10y3bnb04s4tnnht80000gn/T/ipykernel_23060/4192919542.py:8: RuntimeWarning: invalid value encountered in double_scalars
```

```
    q_tmp = (d[i] - a[i] * q[i - 1]) / (b[i] + a[i] * p[i - 1])
```

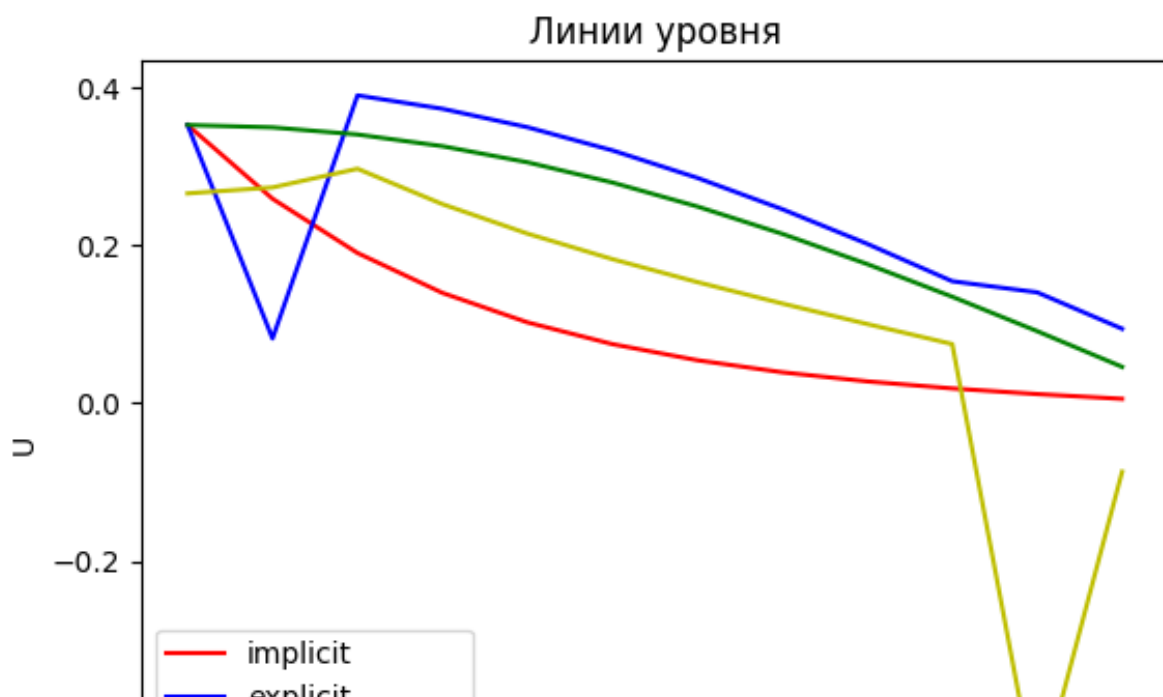


**Исследование зависимости погрешности от величина tau и h**



```
In [29]: N, K, T = 12, 100, 18
```

```
args = eq
args['bound_type'] = 'a1p1'
solver = Parabolic(args, N, K, T)
dict_ans = {
    'implicit': solver.solve_implicit(N, K, T),
    'explicit': solver.solve_explicit(N, K, T),
    'crank_nicolson': solver.solve_crank_nicolson(N, K, T),
    'analytic': solver.solve_analytic(N, K, T)
}
show(dict_ans, 2)
```



In [30]: `N, K, T = 120, 100, 18`

```
args = eq
args['bound_type'] = 'a1p2'
solver = Parabolic(args, N, K, T)
dict_ans = {
    'implicit': solver.solve_implicit(N, K, T),
    'explicit': solver.solve_explicit(N, K, T),
    'crank_nicolson': solver.solve_crank_nicolson(N, K, T),
    'analytic': solver.solve_analytic(N, K, T)
}
show(dict_ans, 2)
```

```
/var/folders/83/2lkmbrc10y3bnb04s4tnnht80000gn/T/ipykernel_23060/3
01663272.py:103: RuntimeWarning: overflow encountered in double_sc
alars
    u[k][j] = (u[k - 1][j + 1] * (self.a ** 2.0 * self.tau / self.h
** 2.0)
/var/folders/83/2lkmbrc10y3bnb04s4tnnht80000gn/T/ipykernel_23060/3
01663272.py:104: RuntimeWarning: overflow encountered in double_sc
alars
    - 2 * u[k - 1][j] * (self.a ** 2.0 * self.tau / self.h ** 2.0)
/var/folders/83/2lkmbrc10y3bnb04s4tnnht80000gn/T/ipykernel_23060/3
01663272.py:105: RuntimeWarning: overflow encountered in double_sc
alars
    + u[k - 1][j - 1] * (self.a ** 2.0 * self.tau / self.h ** 2.0)
/var/folders/83/2lkmbrc10y3bnb04s4tnnht80000gn/T/ipykernel_23060/3
01663272.py:114: RuntimeWarning: overflow encountered in double_sc
alars
    u[k][-1] = (((2.0 * self.gamma * self.a / self.h / (2.0 * self.a
+ self.h * self.b)) * u[k][-2] +
/var/folders/83/2lkmbrc10y3bnb04s4tnnht80000gn/T/ipykernel_23060/3
01663272.py:103: RuntimeWarning: overflow encountered in double_sc
```

## Выводы:

Как видно из графиков погрешности, в основном очередь на неё влияет величина параметра  $\tau$ , а вот количество шагов  $h$  оказывает отрицательное влияние, увеличивая погрешность в разы.

In [30]: