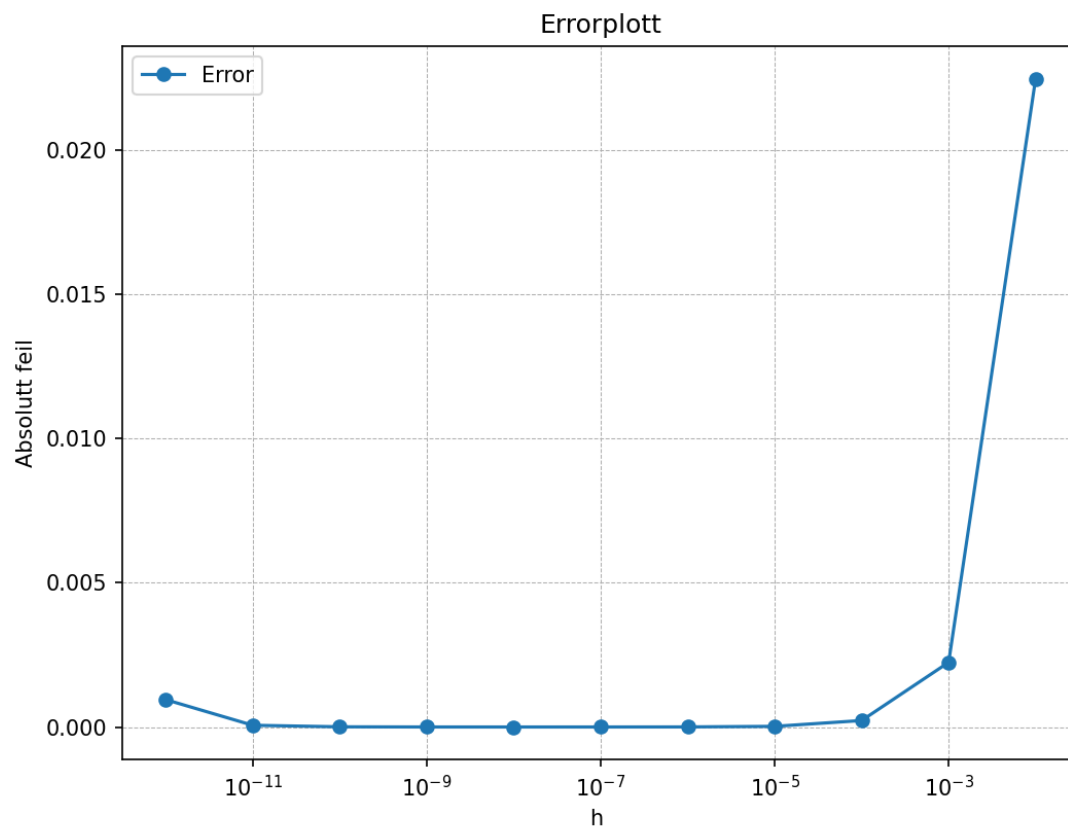


Standardprosjekt

- 1) For å løse denne oppgaven lagde jeg et pythonprogram som finner feilen for ulike h og plotter den.

```
1 import numpy as np
2 import matplotlib.pyplot as plt
3
4 h_values=[0.01,0.001,0.0001,0.00001,0.000001,0.0000001,0.00000001,0.000000001,0.0000000001,0.00000000001,0.000000000001]
5
6 def f(x):
7     return np.exp(x)
8
9 def f_prime_analytic(x):
10    return np.exp(x)
11
12 def f_prime_numerical(x,h):
13    return (f(x+h)-f(x))/h
14
15 calc_list=[]
16
17 for h in h_values:
18    calc_list.append(f_prime_numerical(1.5,h))
19
20
21 for i in range(len(calc_list)-1):
22    print("h: ",h_values[i]," Error: ",abs(calc_list[i]-f_prime_analytic(1.5)))
23
24 plt.figure(figsize=(8,6))
25 plt.plot(h_values,abs(calc_list-f_prime_analytic(1.5)),'-o',label='Error')
26 plt.xscale('log')
27
28 plt.xlabel("h ")
29 plt.ylabel("Absolutt feil")
30 plt.title("Errorplott")
31
32
33
34 plt.legend()
35 plt.grid(True, linestyle="--", linewidth=0.5)
36 plt.show()
37
38
```



Her ser man at for store h blir feilen veldig stor, for veldig små h vil feilen begynne å øke. Så svaret er da at for h mindre enn $1E-11$ vil det oppstå problemer.

2)

$$f(x+h) = f(x) + hf'(x) + \frac{h^2}{2}f''(x) + \frac{h^3}{6}f'''(x) + O(h^4)$$

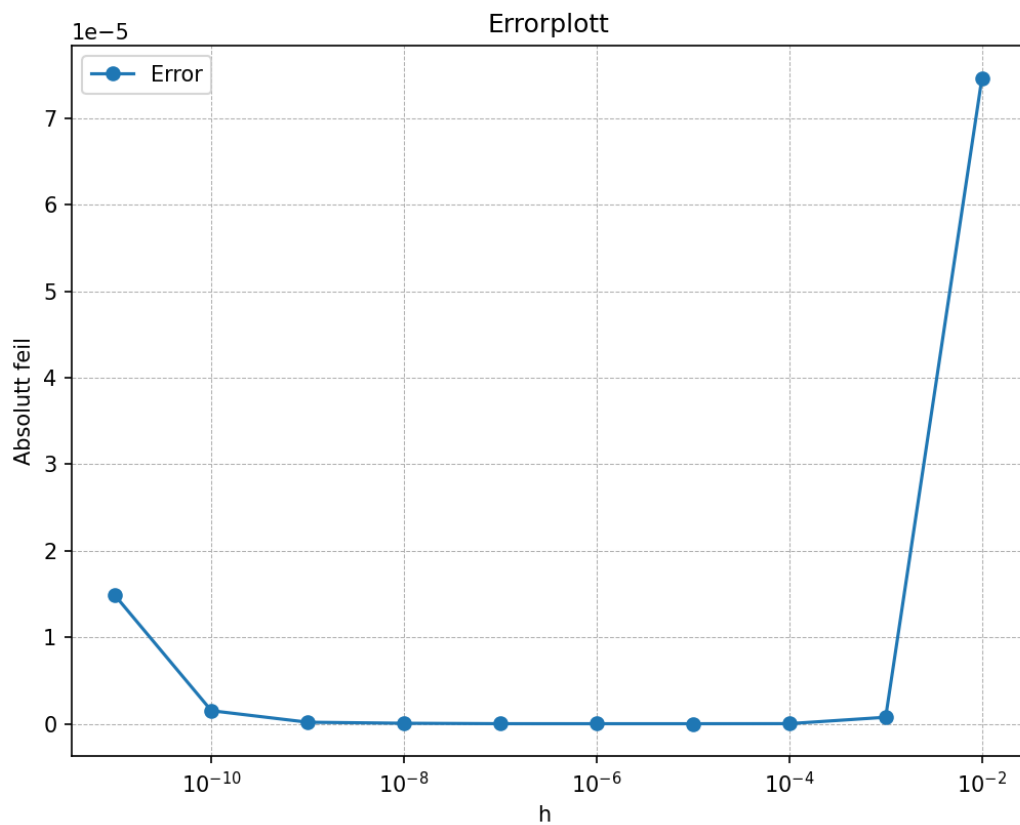
$$f(x-h) = f(x) - hf'(x) + \frac{h^2}{2}f''(x) - \frac{h^3}{6}f'''(x) + O(h^4)$$

$$f(x+h) - f(x-h) = 2hf'(x) + \frac{2h^3}{6}f'''(x)$$

$$\frac{f(x+h) - f(x-h)}{2h} = f'(x) + \frac{h^2}{6}f'''(x)$$

$$feil = \frac{h^2}{6}f'''(x)$$

Her har jeg beskrevet Taylorutviklingen for $f(x+h)$ og $f(x-h)$ og delt på $2h$ for å få en Taylorutvikling for gitt formel. Feilen blir da det siste leddet. Antakelsen er at for store h vil feilen så klart bli veldig stor, og bli lavere for mindre h , helt til h blir veldig liten og feilen blir større.



Dette er plottet for formelen over. Her ser man at antakelsen stemmer. Feilen er veldig stor for $h > 10^{-2}$ og blir større igjen for $h < 10^{-10}$.

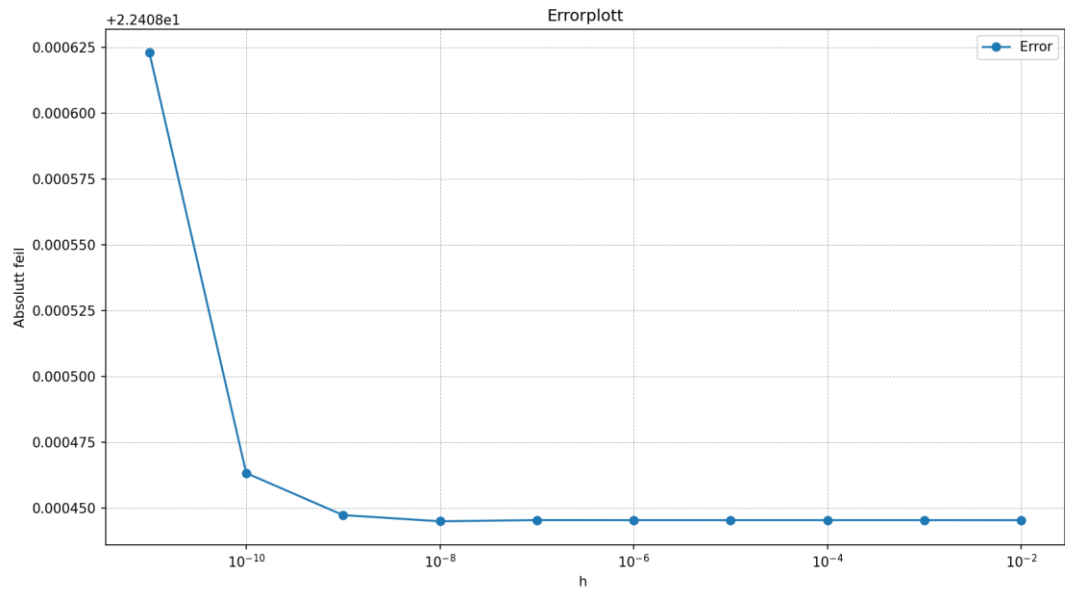
Koden er så si identisk som forrige.

```

1 import numpy as np
2 import matplotlib.pyplot as plt
3
4 h_values=[0.01,0.001,0.0001,0.00001,0.000001,0.0000001,0.00000001,0.000000001,0.0000000001,0.00000000001]
5
6 def f(x):
7     return np.exp(x)
8
9 def f_prime_analytic(x):
10    return np.exp(x)
11
12 def f_prime_numerical(x,h):
13    return (f(x+h)-f(x-h))/(2*h)
14
15
16
17 calc_list=[]
18
19 for h in h_values:
20     calc_list.append(f_prime_numerical(1.5,h))
21
22
23 for i in range(len(calc_list)):
24     print("h: ",h_values[i]," Error: ",abs(calc_list[i]-f_prime_analytic(1.5)))
25
26 plt.figure(figsize=(8,6))
27 plt.plot(h_values,abs(calc_list-f_prime_analytic(1.5)),'-o',label='Error')
28 plt.xscale('log')
29
30 plt.xlabel("h ")
31 plt.ylabel("Absolutt feil")
32 plt.title("Errorplott")
33
34
35
36 plt.legend()
37 plt.grid(True, linestyle="--", linewidth=0.5)
38 plt.show()
39

```

3) Samme prosedyre, så si samme kode her også.



Feilen vokser for ekstremt små h, her begynner feilen å vokse for $h < 10^{-8}$.

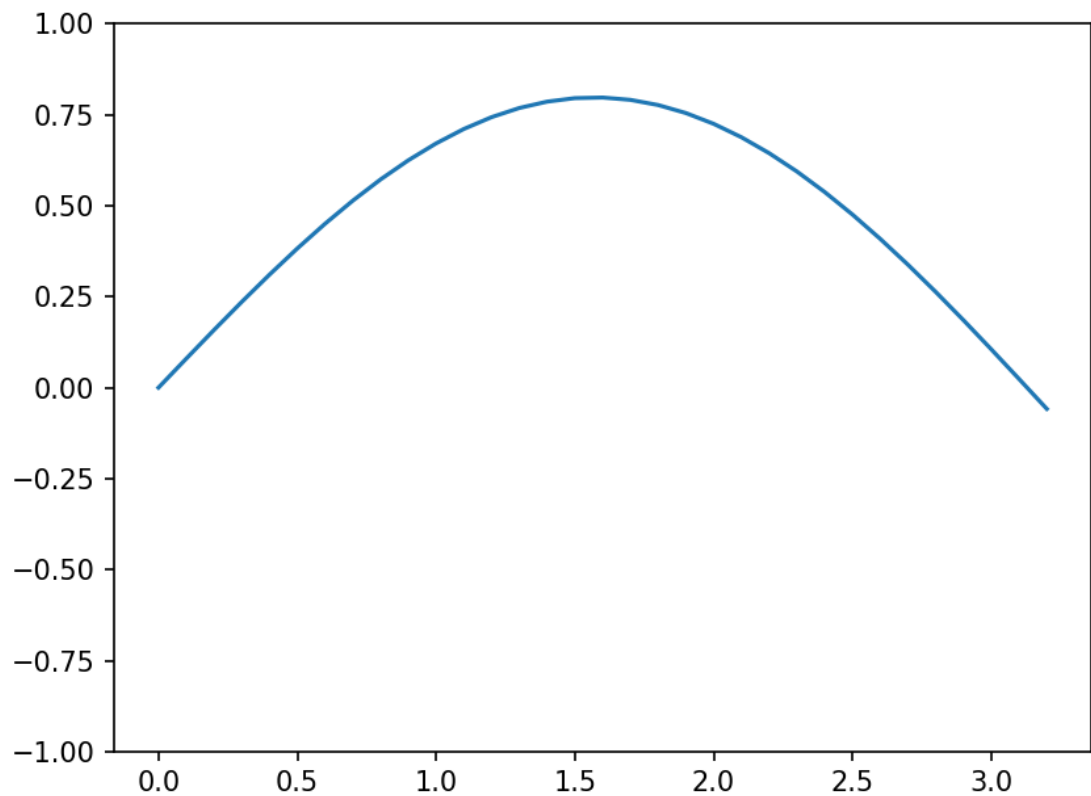
4)

```

1 import numpy as np
2 import matplotlib.pyplot as plt
3 import matplotlib.animation as animation
4
5 def solve_heat_equation(h, k, T, L):
6     x = np.arange(0, L+h, h)
7     t = np.arange(0, T+k, k)
8
9     u = np.sin(x)
10    U = [u.copy()]
11
12    lambd = k / h**2
13    if lambd > 0.5:
14        print(f"Ustabil! lambd={lambd:.2f}, velg annen h og/eller k")
15        return None
16
17    for i in t[1:]:
18        u_new = u.copy()
19        for j in range(1, len(x)-1):
20            u_new[j] = u[j] + lambd * (u[j+1] - 2*u[j] + u[j-1])
21        u = u_new
22        U.append(u.copy())
23
24    return x, t, U
25
26 def animate_solution(x, t, U):
27     fig, ax = plt.subplots()
28     line, = ax.plot(x, U[0])
29     ax.set_ylim(-1, 1)
30
31     def update(n):
32         line.set_ydata(U[n])
33         return line,
34
35     ani = animation.FuncAnimation(fig, update, frames=len(t), interval=50)
36     plt.show()
37
38 h_values = [1000, 0.1, 0.05, 0.01]
39 k_values = [0.01, 0.01, 0.001, 0.001]
40 T, L = 1, np.pi
41
42 for h in h_values:
43     for k in k_values:
44         print(f"Kjører for h={h}, k={k}")
45         result = solve_heat_equation(h, k, T, L)
46         if result:
47             x, t, U = result
48             animate_solution(x, t, U)

```

Her har jeg definert variabelen k/h^2 . Linje 39 og 40 er to lister for ulike verdier for h og k. Her itererer jeg gjennom begge listene og animerer resultatene.



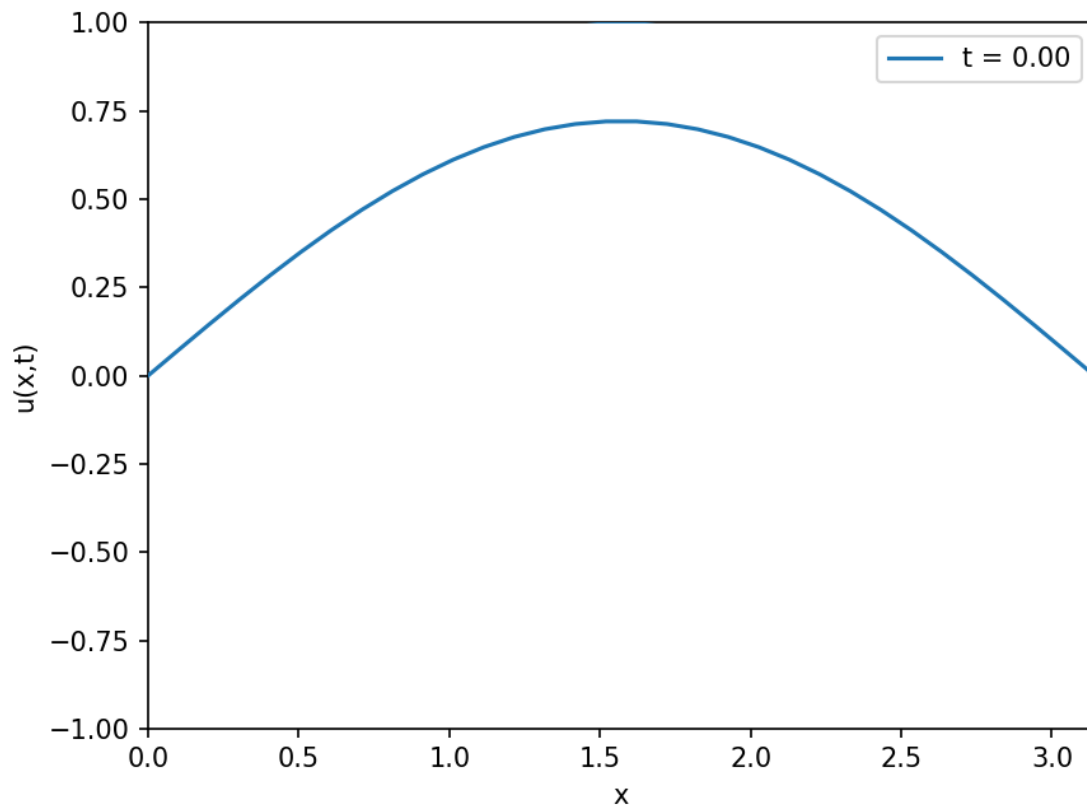
```
Kjører for h=0.1, k=0.01
Ustabil for lambda=1.00, velg annen h og/eller k
Kjører for h=0.05, k=0.01
Ustabil for lambda=4.00, velg annen h og/eller k
Kjører for h=0.05, k=0.01
Ustabil for lambda=4.00, velg annen h og/eller k
Kjører for h=0.05, k=0.001
Kjører for h=0.05, k=0.01
Ustabil for lambda=4.00, velg annen h og/eller k
Kjører for h=0.01, k=0.01
Ustabil for lambda=100.00, velg annen h og/eller k
Kjører for h=0.01, k=0.01
Ustabil for lambda=100.00, velg annen h og/eller k
Kjører for h=0.01, k=0.001
Ustabil for lambda=10.00, velg annen h og/eller k
Kjører for h=0.01, k=0.01
Ustabil for lambda=100.00, velg annen h og/eller k
```

```

1 import numpy as np
2 import matplotlib.pyplot as plt
3 import matplotlib.animation as animation
4
5 def solve_heat_eq(h, k, t, T):
6     Nx = int(L / h)
7     Nt = int(T / h)
8     x = np.linspace(0, L, Nx+1)
9     t = np.linspace(0, T, Nt+1)
10    x = x / L * pi
11
12    U = np.zeros((Nx+1, Nt+1))
13    U[:, 0] = np.sin(x)
14
15    A = np.zeros((Nx+1, Nx+1))
16    for i in range(Nx+1):
17        A[i, i] = 1 + 2 * h
18        if i > 0:
19            A[i, i-1] = -h
20        if i < Nx-1:
21            A[i, i+1] = -h
22
23    for j in range(Nt):
24        b = U[:, 0, j]
25        U[:, 1:Nx, j+1] = np.linalg.solve(A, b)
26
27    return x, t, U
28
29 def animate(x, t, U):
30     fig, ax = plt.subplots()
31     line, = ax.plot(x, U[:, 0, 0], label="t = 0")
32     ax.set_xlim([0, L], [0, 1])
33     ax.set_ylim([0, 1])
34     ax.set_xlabel("x")
35     ax.set_ylabel("u(x,t)")
36     ax.legend()
37
38     def update(frame):
39         line.set_ydata(U[:, frame])
40         ax.legend([f"t = {t[frame]:.2f}"])
41         return line,
42
43    ani = animation.FuncAnimation(fig, update, frames=len(t), interval=10, blit=True)
44    plt.show()
45
46 def main():
47     h_values=[0.1, 0.05, 0.001]
48     k_values=[0.01, 0.01, 0.00001]
49     for h in h_values:
50         for k in k_values:
51             print(f"Kjörur for h={h}, k={k}")
52             h=T/h, k=1
53             x,t,U=solve_heat_eq(h,k,L,T)
54             animate(x,t,U)
55
56 if __name__ == "__main__":
57     main()

```

5)



6)

```

def solve_heat_eq(h, k, L, T):
    Nx = int(L / h)
    Nt = int(1 / h)
    x = np.linspace(0, L, Nx+1)
    t = np.linspace(0, 1, Nt+1)
    r = h / h**2
    U = np.zeros((Nx+1, Nt+1))
    U[:, 0] = np.sin(x)

    A = np.zeros((Nx+1, Nx+1))
    B = np.zeros((Nx+1, Nx+1))
    for i in range(Nx-1):
        A[i, i] = 1 + 2 * r
        B[i, i+1] = -r
        B[i+1, i] = -r
        if i > 0:
            A[i, i-1] = -r
            B[i, i-1] = -r
        if i < Nx-1:
            A[i, i+1] = -r
            B[i, i+1] = -r
    for j in range(Nt):
        b = 0 * dot(U[:, Nx, j])
        U[:, Nx, j+1] = np.linalg.solve(A, b)
    return x, t, U

def animate(x, t, U):
    fig, ax = plt.subplots()
    line, = ax.plot(x, U[:, 0], label='t = 0')
    ax.set_xlim([0, L], [0, 1])
    ax.set_ylim([0, 1])
    ax.set_xlabel('x')
    ax.set_ylabel('u(x,t)')
    ax.legend()

def update(frame):
    line.set_ydata(U[:, frame])
    ax.legend(['t = ' + t[frame::2f]'])
    return line

ani = animation.FuncAnimation(fig, update, frames=len(t), interval=50, blit=True)
plt.show()

def main():
    h_values=[0.1, 0.001, 0.0001]
    k_values=[0.01, 0.001, 0.000001]
    for h in h_values:
        for k in k_values:
            print(f'Kjører for h={h}, k={k}')
            L, Np, Nt = 1, 1000, 1000
            x, t, U = solve_heat_eq(h, k, L, T)
            animate(x, t, U)

if __name__ == "__main__":
    main()

```

Sammenlikningen forteller at for store h og store k foretrekkes Crank Nickolson og implisitt metode. Dette er på grunn av at $\lambda \leq 0.5$ for stabilitet. Ellers vil feilen bli for stor. Eksplisitt metode er da bedre egnet for enkle beregninger gitt $\lambda \leq 0.5$.