Assignment 2: Θ -method for Heat equation

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1 Introduction

The θ method is a semi-implicit forward time centered space scheme to solve the heat equation. The heat equation in one dimension is given as follows:

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

2 Θ Scheme

The scheme to solve the equation is as follows:

$$\partial_t^+ U_j^k = (\theta) \partial_x^+ \partial_x^- U_j^{k+1} + (1 - \theta) \partial_x^+ \partial_x^- U_j^k$$

$$\left(\frac{U_j^{k+1} - U_j^k}{\Delta t}\right) = (\theta) \left(\frac{U_{j+1}^{k+1} - 2U_j^{k+1} + U_{j-1}^{k+1}}{\Delta x^2}\right)$$

$$+ (1 - \theta) \left(\frac{U_{j+1}^k - 2U_j^k + U_{j-1}^k}{\Delta x^2}\right)$$

Setting $\lambda_1 = \left(\frac{\Delta t}{\Delta x^2}\right)\theta$ and $\lambda_2 = \left(\frac{\Delta t}{\Delta x^2}\right)(\theta - 1)$, we can write it in a matrix-vector form,

$$\begin{bmatrix} 1+2\lambda_1 & -\lambda_1 & 0 & \dots & 0 \\ -\lambda_1 & 1+2\lambda_1 & -\lambda_1 & \ddots & 0 \\ 0 & -\lambda_1 & 1+2\lambda_1 & \ddots & 0 \\ \vdots & \ddots & \ddots & \ddots & \vdots \\ 0 & \dots & 0 & -\lambda_1 & 1+2\lambda_1 \end{bmatrix} \begin{bmatrix} U_1^{k+1} \\ U_2^{k+1} \\ \vdots \\ U_{J-1}^{k+1} \end{bmatrix} = \begin{bmatrix} 1+2\lambda_2 & -\lambda_2 & 0 & \dots & 0 \\ -\lambda_2 & 1+2\lambda_2 & -\lambda_2 & \ddots & 0 \\ 0 & -\lambda_2 & 1+2\lambda_2 & \ddots & 0 \\ \vdots & \ddots & \ddots & \ddots & \vdots \\ 0 & \dots & 0 & -\lambda_2 & 1+2\lambda_2 \end{bmatrix} \begin{bmatrix} U_1^k \\ U_2^k \\ \vdots \\ U_{J-1}^k \end{bmatrix}$$

Taking the inverse, we can write it as a linear equation

$$U^{k+1} = \mathbf{A}(\mu, \theta)U^k$$

The eigenvalues of the matrix **A** dictate the stability of the scheme.

3 Stability Analysis using eigenvalues

It can be clearly seen in Figure 1 that the method is stable for $\theta \ge 0.5$ and for $\theta < 0.5$ when $\mu < 0.5$. This is what we obtained from Von-Neumann Stability analysis of the method.

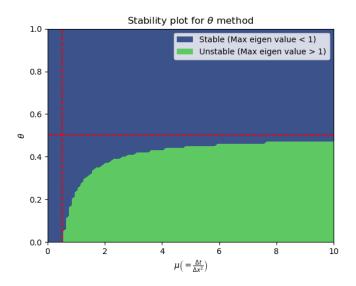


Figure 1: Stability plot fot θ method

4 Convergence study

The following set of plots show the solution for various values of μ and θ . The red line represents the exact solution and the blue line represents the numerical solution. The convergence plots are log log plotted against μ and Δt . The red line in this plot represents $\mu = 0.5$. The contour plot represents the solution in the x-t plane.

4.1 Explicit $(\theta = 0)$

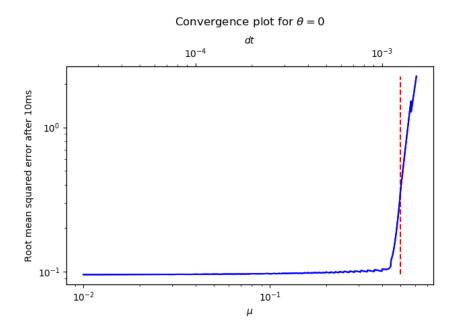


Figure 2: Convergence plot for $\theta = 0$

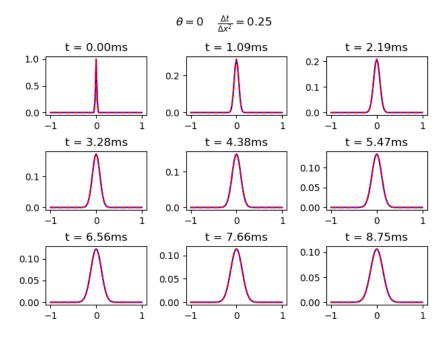


Figure 3: Solution for $\mu=0.2$ and $\theta=0$

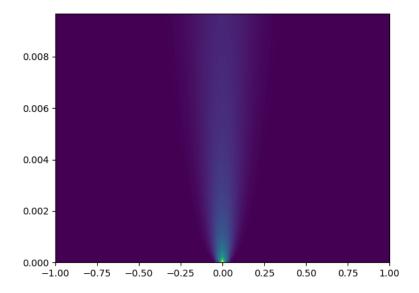


Figure 4: Solution for $\mu = 0.2$ and $\theta = 0$

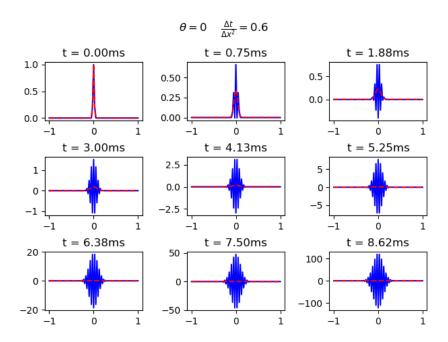


Figure 5: Solution for $\mu=0.6$ and $\theta=0$ (Blowup. Unstable)

4.2 Implicit $(\theta = 1)$

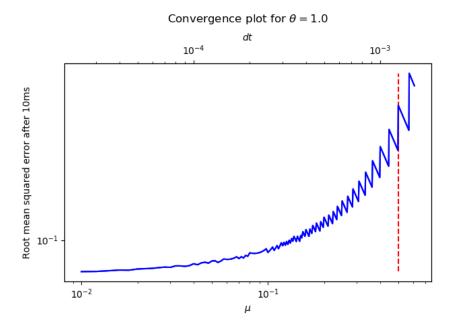


Figure 6: Convergence plot for $\theta = 1$

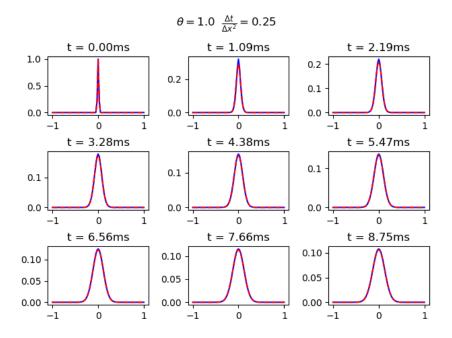


Figure 7: Solution for $\mu=0.2$ and $\theta=1$

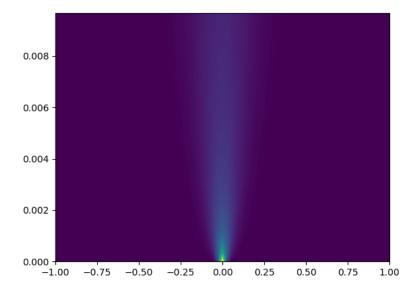


Figure 8: Solution for $\mu = 0.2$ and $\theta = 1$

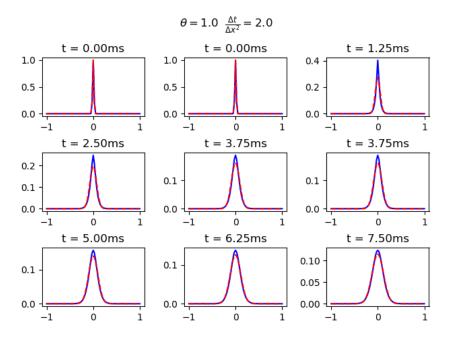


Figure 9: Solution for $\mu=2.0$ and $\theta=1$

4.3 CN $(\theta = 0.5)$

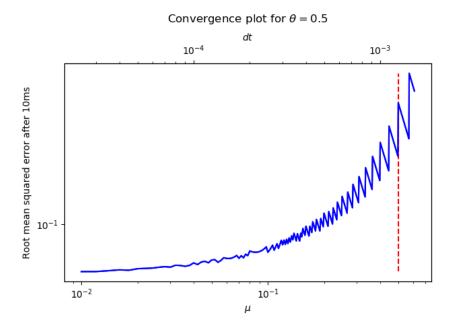


Figure 10: Convergence plot for $\theta = 0.5$

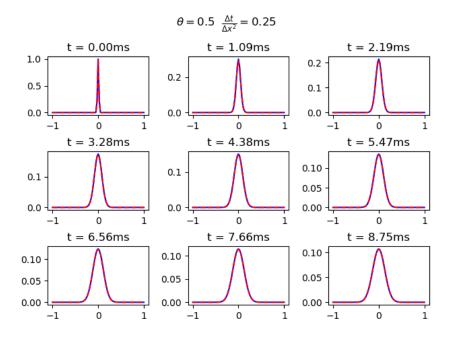


Figure 11: Solution for $\mu=0.2$ and $\theta=0.5$

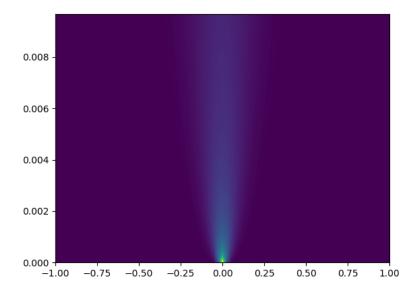


Figure 12: Solution for $\mu = 0.2$ and $\theta = 0.5$

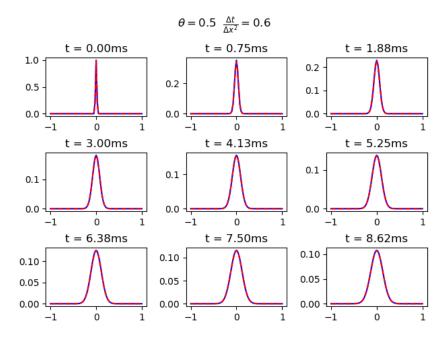


Figure 13: Solution for $\mu = 0.6$ and $\theta = 0.5$

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Figure 14: Convergence plot for $\theta = 0.6$

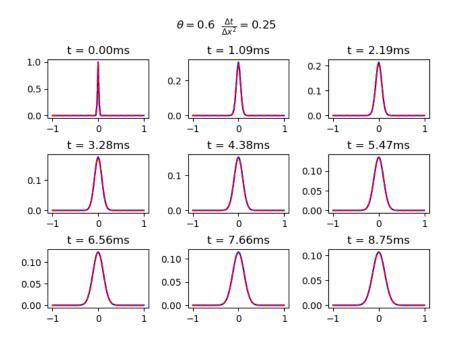


Figure 15: Solution for $\mu = 0.2$ and $\theta = 0.6$

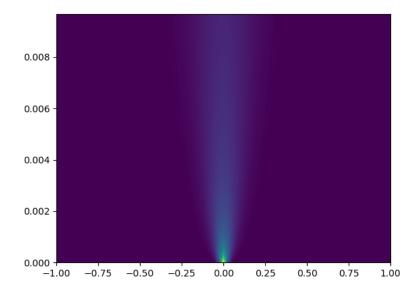


Figure 16: Solution for $\mu = 0.2$ and $\theta = 0.6$

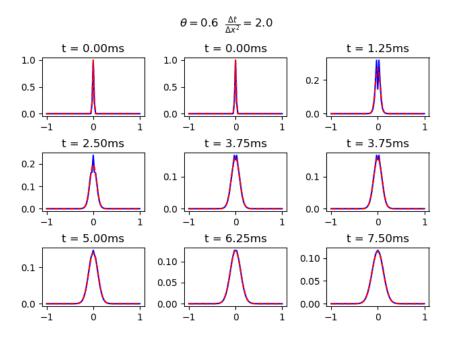


Figure 17: Solution for $\mu=2.0$ and $\theta=0.6$

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Figure 18: Convergence plot for $\theta = 0.4$

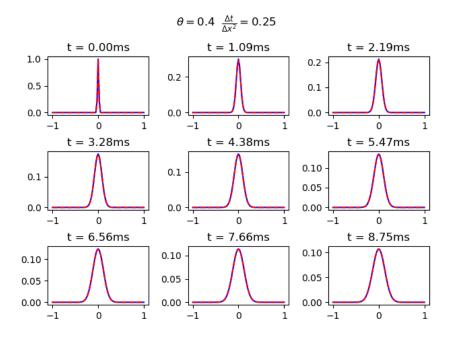


Figure 19: Solution for $\mu = 0.2$ and $\theta = 0.4$

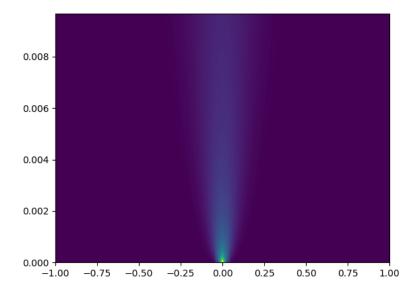


Figure 20: Solution for $\mu = 0.2$ and $\theta = 0.4$

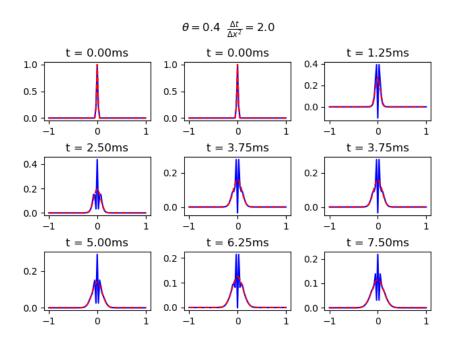


Figure 21: Solution for $\mu=2.0$ and $\theta=0.4$ (Unstable)

5 Code

```
1 import numpy as np
  import matplotlib.pyplot as plt
  #Following is the code for theta method to solve the heat equation
  #Initialization function
   def init_f(x, t=0):
        return np.sqrt (1e-4/(t+1e-4))*np.exp(-x**2/(4*(t+1e-4)))
#Function to construct matrix for 1 side (see report)
   def construct_matrix(mu, theta, size):
        l = mu*theta
        mat = np.zeros([size, size])
13
        \operatorname{mat}[\operatorname{np.arange}(\operatorname{size}),\operatorname{np.arange}(\operatorname{size})] = 1 + 2 * 1
14
        mat[np.arange(size-1)+1,np.arange(size-1)] = -1*1
        \operatorname{mat}\left[\operatorname{np.arange}\left(\operatorname{size}-1\right),\operatorname{np.arange}\left(\operatorname{size}-1\right)+1\right] = -1*1
16
17
        return mat
18
19 #Function to create main 'A' matrix'
   def construct_mat(mu, theta, size):
20
21
        m1 = construct_matrix (mu, theta, size)
        m2 = construct_matrix(mu, theta - 1, size)
22
23
        return np.linalg.inv(m1).dot(m2)
24
  #Main theta method code
   def theta_method(theta, mu, num_points, tsteps, init_fn = init_f):
26
        dx = 2.0/(num_points - 1)
27
28
        dt = mu*(dx**2)
        x = np. linspace(-1,1,num-points)
29
        t = np.arange(tsteps)*dt
30
        X,T = np. meshgrid(x,t)
31
        u_0 = init_f(x)
32
33
        U = np. zeros ([num_points, tsteps])
34
35
        U[:,0] = u_0
36
        mat = construct_mat(mu, theta, num_points-2)
37
38
        for i in range (tsteps -1):
39
             U[1:-1,i+1] = mat.dot(U[1:-1,i])
40
41
        return U.T, X, T
42
43
  #Plot convergence plots for a given theta
44
45
   def convergence_study(theta):
46
        num_points = 41
        mu = np. linspace (0.01, 0.61, 301)
47
        dx = 2.0/(num_points - 1)
48
        dt = mu*(dx**2)
49
        errors = np. zeros_like (mu)
50
        num\_steps = (0.01/(mu*dx**2)).astype(int)
51
52
        for i,m in enumerate(mu):
53
                            = \begin{array}{lll} theta\_method\,(\,theta\,,\,\,m,\,\,num\_points\,,num\_steps\,[\,i\,]\,) \\ = U[\,-1\,,:]\,\,-\,\,i\,ni\,t\_f\,(X[\,0\,\,,:]\,\,,num\_steps\,[\,i\,]*dt\,[\,i\,]\,) \end{array}
             U, X, T
54
             err
56
             errors [i]
                            = np. sqrt (np. sum (err **2))
57
        fig, ax = plt.subplots()
58
        ax.\,set\_title\,(\,r\,'Convergence\ plot\ for\ \$\backslash\,theta\,=\,\{\}\$\,'\,.\,format\,(\,theta\,)+\,'\backslash\,n\backslash\,n\,'\,)
59
        ax.loglog(mu, errors, 'b')
60
        ax.plot([0.5,0.5],[np.min(errors),np.max(errors)],'r-')
61
        ax.set_xlabel(r'\$\ms')
62
        ax.set_ylabel('Root mean squared error after 10ms')
63
        new_t = [xt*(dx**2) \text{ for } xt \text{ in } ax.get_xticks()]
64
        xmin, xmax = ax.get_xlim()
```

```
ax2 = ax.twinv()
66
       ax2.set_xlim([xmin*(dx**2), xmax*(dx**2)])
67
       ax2.loglog(dt, errors, 'b')
68
69
       ax2.set_xlabel('$dt$')
       plt.tight_layout()
70
       plt.savefig('theta{:.0f}.png'.format(theta*10))
71
72
73 #Eigenval stability analysis for various mu and theta values
   def stability_analysis():
74
       mu = np. linspace(0,10,100)
75
76
       theta = np.linspace(0,1,100)
       mu, theta = np. meshgrid (mu, theta)
77
       eigval = np.zeros_like(mu)
78
       for m, t, e in zip (np.nditer (mu), np.nditer (theta), np.nditer (eigval, op-flags=['readwrite
79
        '])):
                e[...] = np.max(abs(np.linalg.eigvals(construct_mat(m, t, 5))))
80
81
       eigval = (eigval > 1) + 0
82
       fig , ax = plt.subplots()
83
       cs = ax.contourf(mu, theta, eigval, 1)
84
85
       proxy = [plt.Rectangle((0,0),1,1,fc = pc.get_facecolor()[0])
                    for pc in cs.collections]
86
87
       plt.legend(proxy, ["Stable (Max eigen value < 1)", "Unstable (Max eigen value > 1)"])
88
89
       ax. plot ([10,0],[0.5,0.5],'r—')
90
       ax.plot([0.5,0.5],[1,0],'r-')
91
       ax.set_xlabel(r'\$\mu \lceil eff(= \frac{\beta rac}{Delta t}{\Delta x^2} right) 
92
       ax.set_ylabel(r'$\theta$')
93
       ax.set_title(r"Stability plot for $\theta$ method")
94
       plt.savefig('eig.png')
95
       plt.close()
96
   #Code to generate plots for 1 instance of the theta method
98
   def plot_instance(theta, mu):
99
100
       num_points = 81
       dx = 2.0/(num\_points-1)
       dt = mu*(dx**2)
       numsteps = int(0.01/(mu*dx**2))
104
       U, X, T
                    = theta_method(theta, mu, num_points, numsteps)
       fig , axs = plt.subplots(3,3)
       fig.suptitle(r'\$\theta = {}\$'.format(theta)+'\t'+r'\$\frac{\Delta t}{\Delta t} = '+''{}
106
        '. format (mu))
107
        for i,ax in enumerate(axs.flatten()):
108
           ax.plot(X[0,:],U[i*numsteps/9],'b-')
           ax.plot(X[0,:], init_f(X[0,:],T[i*numsteps/9,0]), 'r--')
           ax.set_title("t = {:.2f}ms".format(T[i*numsteps/9,0]*1000))
       plt.tight_layout()
113
       fig.subplots_adjust(top=0.85)
114
       plt.savefig('mu{:.0f}theta{:.0f}.png'.format(mu*10.0,theta*10.0))
116
       plt.close()
       plt.figure()
118
       plt.contourf(X,T,U,200)
       plt.savefig('c_mu\{:.0f\}theta\{:.0f\}.png'.format(mu*10.0,theta*10.0))
119
       plt.close()
120
121
122
      __name__='__main__':
        stability_analysis()
123
       for theta in [0,0.4,0.5,0.6,1.0]:
124
            convergence_study (theta)
125
        for mu in [0.25,0.6,2.0]:
126
            for theta in [0, 0.4, 0.5, 0.6, 1.0]:
127
                plot_instance(theta,mu)
128
```

6 Conclusion

We have successfully checked the convergence and stability of the θ method for the heat equation for various values of μ and θ . We have shown that the sceme is stable for large time steps when $\theta \geq 0.5$ and is conditionally stable otherwise.