## APL321: Introduction to Computational Fluid Dynamics

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Consider the non-dimensional 1-D convection-diffusion equation for  $\phi(x)$  in class:

$$Pe\frac{d\phi}{dx} = \frac{d^2\phi}{dx^2} \tag{1}$$

with boundary conditions  $\phi(0) = 0$  and  $\phi(1) = 1$ . Here Pe is the Peclet number; choose Pe = 50 for all the parts in this assignment. The exact solution to  $\phi(x)$  is given by  $\phi_{exact}(x) = (exp(xPe) - 1)/(exp(Pe) - 1)$ . Discretize Eqn. (1) using a Finite Difference CDS for convective term and diffusion term at N points  $x_1, x_2, ...x_N$  on a uniform grid with cell width  $h = \frac{1}{N-1}$ .

### Solution:

We solve equation (1) using the central differencing scheme (CDS) for the diffusive term (RHS) as well as the convective term (LHS). According to CDS, the first and second order approximation of the derivative at node i are given by:

$$\frac{d\phi}{dx}|_{i} \approx \frac{\phi_{i+1} - \phi_{i-1}}{2h} \tag{2}$$

$$\frac{d^2\phi}{dx^2}|_i \approx \frac{\phi_{i+1} + \phi_{i-1} - 2\phi_i}{h^2}$$
 (3)

Substituting these values in the original convection-diffusion equation (1) yields:

$$\left(1 - \frac{hPe}{2L}\right)\phi_{i+1} + \left(1 + \frac{hPe}{2L}\right)\phi_{i-1} - 2\phi_i = 0$$
(4)

Accordingly, the constants for the scheme are  $A_E = 1 - \frac{hPe}{2L}$ ,  $A_W = 1 + \frac{hPe}{2L}$ , and  $A_P = -2$ . Thus, we now have to solve a system of N equations with N unknowns given by:

$$\begin{bmatrix} 1 & 0 & 0 & 0 & \cdots & 0 \\ A_W & A_P & A_E & 0 & \cdots & 0 \\ 0 & A_W & A_P & A_E & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & \ddots & \vdots \\ \vdots & \vdots & 0 & A_W & A_P & A_E \\ 0 & 0 & \cdots & \cdots & 0 & 1 \end{bmatrix} \begin{bmatrix} \phi_1 \\ \phi_2 \\ \vdots \\ \vdots \\ \phi_N \end{bmatrix} = \begin{bmatrix} \phi_0 \\ 0 \\ \vdots \\ \vdots \\ \phi_L \end{bmatrix}$$

where  $\phi_0 = 0$  and  $\phi_L = 1$  are the boundary conditions at x = 0 and x = L.

We now use the inbuilt system of equations solver linsolve to solve the above system to get the values of  $\phi$  at all the nodes and store this solution for questions ahead.

1. The discretized equation has a form  $[L]\vec{\phi}_{exact} + \vec{\epsilon}_{\tau} = 0$  where L is an  $N \times N$  matrix,  $\vec{\phi}_{exact} = [\phi_{exact}(x_1)\phi_{exact}(x_2)...\phi_{exact}(x_N)]^T$ , and  $\vec{\epsilon}_{\tau} = [\epsilon_{\tau}(x_1)\epsilon_{\tau}(x_2)...\epsilon_{\tau}(x_N)]^T$  is the vector representing truncation error at  $x_1, x_2, ...x_N$ . The matrix [L] has tridiagonal form, with entries  $A_W, A_P$  and  $A_E$  at each  $x_i$ . Derive and state the values for  $A_W, A_P$  and  $A_E$  in terms of grid width h and Peclet number for both interior and boundary nodes. Use Taylor series expansion to derive and state the expression  $\epsilon_{\tau}(x_i)$  for only upto second order in h in terms of derivatives of  $\phi_{exact}$  at  $x_i$ .

### Solution 1:

We begin by using the Taylor Series expansion to find out the values of  $\phi$  at the points  $x_{i-1}$  and  $x_{i+1}$  for i = 2, 3, ...N - 1. The series expansion is given by:

$$\phi(x_{i-1}) = \phi(x_i) - h\phi'(x_i) + \frac{h^2}{2!}\phi''(x_i) - \frac{h^3}{3!}\phi'''(x_i) + \frac{h^4}{4!}\phi''''(x_i) + O(h^5)$$
 (5)

$$\phi(x_{i+1}) = \phi(x_i) + h\phi'(x_i) + \frac{h^2}{2!}\phi''(x_i) + \frac{h^3}{3!}\phi'''(x_i) + \frac{h^4}{4!}\phi''''(x_i) + O(h^5)$$
 (6)

Adding equations (2) and (3) and solving for  $\phi''(x_i)$  gives:

$$\phi''(x_i) = \frac{1}{h^2} [\phi(x_{i+1}) - 2\phi(x_i) + \phi(x_{i-1})] - \frac{h^2}{12} \phi''''(x_i) + O(h^4)$$
 (7)

Subtracting equations (2) and (3) and solving for  $\phi'(x_i)$  gives:

$$\phi'(x_i) = \frac{1}{2h} [\phi(x_{i+1}) - \phi(x_{i-1})] - \frac{h^2}{6} \phi'''(x_i) + O(h^4)$$
(8)

Substituting these expressions into the original convection-diffusion equation (1) and ignoring the higher order terms  $(O(h^4))$  yields:

$$Pe\left[\frac{1}{2h}\{\phi(x_{i+1}) - \phi(x_{i-1})\} - \frac{h^2}{6}\phi'''(x_i)\right] = \frac{1}{h^2}[\phi(x_{i+1}) - 2\phi(x_i) + \phi(x_{i-1})] - \frac{h^2}{12}\phi''''(x_i)$$

$$\implies \left[\frac{1}{h^2} - \frac{Pe}{2h}\right]\phi(x_{i+1}) - \frac{2}{h^2}\phi(x_i) + \left[\frac{1}{h^2} + \frac{Pe}{2h}\right]\phi(x_{i-1}) + \left[\frac{h^2Pe}{6}\phi'''(x_i) - \frac{h^2}{12}\phi''''(x_i)\right] = 0$$
(10)

The above equation takes the form

$$[L]\vec{\phi}_{exact} + \vec{\epsilon}_{\tau} = 0 \tag{11}$$

where the matrix [L] is given by:

$$[L] = \begin{bmatrix} 1 & 0 & 0 & 0 & \cdots & 0 \\ A_W & A_P & A_E & 0 & \cdots & 0 \\ 0 & A_W & A_P & A_E & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & \cdots & \vdots \\ \vdots & \vdots & 0 & A_W & A_P & A_E \\ 0 & 0 & \cdots & \cdots & 0 & 1 \end{bmatrix}$$

$$A_E = \frac{1}{h^2} - \frac{Pe}{2h}, A_P = -\frac{2}{h^2}, A_W = \frac{1}{h^2} + \frac{Pe}{2h}$$

Comparing equation (10) with (11) gives the form of  $\epsilon_{\tau}$  as:

$$\vec{\epsilon}_{\tau} = \begin{bmatrix} \epsilon_{\tau}(x_1) \\ \epsilon_{\tau}(x_2) \\ \vdots \\ \epsilon_{\tau}(x_N) \end{bmatrix} = \begin{bmatrix} 0 \\ \frac{h^2 Pe}{6} \phi'''(x_2) - \frac{h^2}{12} \phi''''(x_2) \\ \vdots \\ 0 \end{bmatrix}$$
(12)

where  $\phi'''$  and  $\phi''''$  are third and fourth order derivatives of the exact solution  $\phi_{exact}$ . Therefore:

$$\phi'''(x_i) = \frac{Pe^3 \exp(x_i Pe)}{\exp(Pe) - 1}$$
(13)

$$\phi''''(x_i) = \frac{Pe^4 \exp(x_i Pe)}{\exp(Pe) - 1}$$
(14)

Note that  $\epsilon_{\tau}(x_1)$  and  $\epsilon_{\tau}(x_N)$  are 0 since the  $\phi$  values at the boundary conditions are known with full accuracy.

2. Calculate and plot the approximate truncation error  $\tilde{\epsilon}_{\tau}(x_i)$  at for points using only upto second order in terms of derivatives of  $x_1, x_2, ...x_N$  at N = 41 points using only terms upto second order in  $h^2$  in terms of derivatives of  $\phi_{exact}$  at  $x_i$ . Comment on the trends observed in this plot.

### Solution 2:

We use equations (12), (13) and (14) to calculate the truncation error  $\tilde{\epsilon}_{\tau}(x_i)$  at all  $x_i$ . From the plots, we observe that the truncation error increases sharply as x tends to 1. This is because of the exponential solution  $\phi_{exact}$  and its derivatives which means that its values increase exponentially with x. This makes sense because as we approach L, the significance of the higher order terms in the Taylor Series increases and hence accounts for the greater loss in accuracy of the error.

The plots of  $\tilde{\epsilon}_{\tau}(x_i)$  vs h and the MATLAB code implemented are given below:

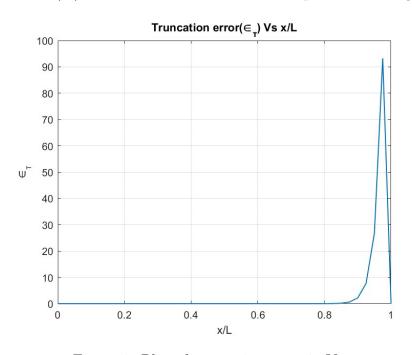


Figure 1: Plot of truncation error  $\tilde{\epsilon}_{\tau}$  Vs  $x_i$ 

```
Question 2
2
                            % Peclet No. = 50
   Pe = 50;
   N = 41;
                            % Grid Points = 41
4
   h = 1/(N-1);
                            % Grid Width = 1/(N-1)
5
6
   err = zeros(N,1);
                            % vector to store truncation
      error
   x = linspace(0,1,N);  % x_i values
   % third order derivative values of phi_exact
   phi3 = (Pe^3*exp(x*Pe))/(exp(Pe)-1);
   % fourth order derivative values of phi_exact
11
   phi4 = (Pe^4*exp(x*Pe))/(exp(Pe)-1);
12
13
   for i = 2:N-1
       % substituting the value of err(i)
14
15
       err(i) = (Pe*h*h/6)*phi3(i) - (h*h/12)*phi4(i);
16
   end
17
18
   % Plot of truncation error vs x_i
19
   figure;
20
   plot(x, err, 'LineWidth',1);
   title('Truncation error(epsilon_tau) Vs x/L')
22
   xlabel('x/L');
23
   ylabel('epsilon_tau');
24
   grid on;
```

MATLAB code for the variation of truncation error  $\tilde{\epsilon}_{\tau}$  with  $x_i$ 

3. The discrete solution  $\vec{\phi}_{exact}$  satisfies  $[L]\vec{\phi}_h=0$ , where  $\vec{\phi}_h=[\phi_h(x_1)\phi_h(x_2)...\phi_h(x_N)]^T$ . The discretization error is defined as  $\vec{\epsilon}_h=\vec{\phi}_{exact}-\vec{\phi}_h$ . Estimate the approximate value of  $\vec{\epsilon}_h$ , denoted as  $\vec{\epsilon}_h$ , from the identity  $\vec{\epsilon}_h=-[L]^{-1}\vec{\epsilon}_\tau$ . Here  $\vec{\epsilon}_\tau$  has been calculated from  $O(h^2)$  terms in part 2, for N=41 points. Compare  $\vec{\epsilon}_h$  and  $\vec{\epsilon}_h$ , on the same graph, in which  $\vec{\epsilon}_h$  and  $\vec{\epsilon}_h$ , have been plotted with respect to h. Comment on the difference between  $\vec{\epsilon}_h$ -vs-h and  $\vec{\epsilon}_h$ -vs-h curves.

### Solution 3:

We use the relations given in the question to compute the exact and absolute discretization errors. We can clearly observe from the graph that the exact error is slightly greater than the approximated error especially when x tends to 1. This is due to the fact that we ignored the higher order terms  $(O(h^4))$  while computing the truncation error in question 2. While this approximation holds good for lower values of x, it tends to deviate significantly as x increases.

The plots of  $\tilde{\epsilon}_h$  and  $\epsilon_h$  vs h and the MATLAB code implemented are given below:

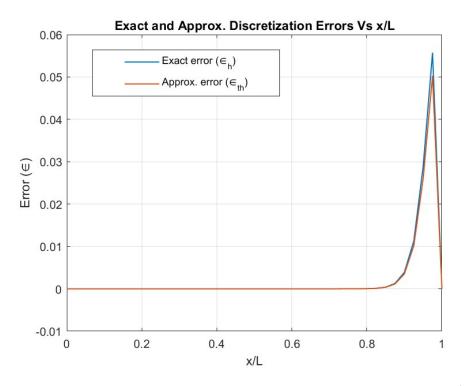


Figure 2: Plot of Exact and Approx. Discretization Errors Vs x/L

```
1
   % Question 3
2
  err_h = phi_exact - phi_h;  % discretization error
3
4
  L = zeros(N,N);
5
6
  L(1,1) = 1; L(N,N) = 1;
  Aw = 1/h^2 + Pe/(2*h);
  Ae = 1/h^2 - Pe/(2*h);
9
  Ap = -2/h^2;
10
11
  for j = 2:N-1
12
      L(j,j) = Ap;
13
      L(j,j-1) = Aw;
14
      L(j,j+1) = Ae;
15
  end
16
17
  18
  % Plotting exact and approx. discretization errors vs h
19
20
  figure;
  plot(x,err_h,'LineWidth',1);
21
22
  hold on
23
  plot(x, err_th, 'LineWidth',1);
  title('Exact and Approx. Discretization Errors Vs x/L')
25
  xlabel('x/L');
  ylabel('Error (epsilon)');
27 | legend('Exact error (epsilon_h)', 'Approx. error (epsilon_
```

```
{th})');

28  xlim([0.00 1.00])

29  ylim([-0.0100 0.0600])

30  legend("Position", [0.18496,0.7564,0.3824,0.12664]);

31  grid on;
```

MATLAB code for the computation of  $\vec{\tilde{\epsilon}}_h$  and  $\vec{\epsilon_h}$ 

4. Plot  $||\vec{\epsilon}_h||$  and  $||\epsilon_h||$  with respect to h on the same log-log graph, using N=41,81,161,321 points. Show a reference line to indicate the order of accuracy. Here ||.|| is the standard deviation norm. Comment on the differences between  $||\vec{\epsilon}_h||$ -vs-h and  $||\epsilon_h||$ -vs-h.

#### Solution 4:

We first find the vectors  $\vec{\epsilon}_h$  and  $\vec{\epsilon}_h$  using the same methods as in the previous questions. Next, we calculate their standard deviation norms using the inbuilt norm function. Finally, we plot these values against h on a log-log scale using the inbuilt loglog function.

From the graph, it is evident that both the curves follow a second order of accuracy since they are almost parallel to the quadratic curve  $y = ch^2$ . This means that both the exact and approximate discretization errors increase quadratically with grid size h. This obeys our theory since we approximated the errors till the  $O(h^2)$  terms. Furthermore, both the norms follow almost the same curve indicating that a second order approximation of the error is highly accurate with the exact error beginning to deviate slightly as h increases on account of the increasing significance of the higher order terms with increasing h.

The plots of  $||\vec{\epsilon}_h||$  and  $||\epsilon_h||$  vs h and the MATLAB code implemented are given below:

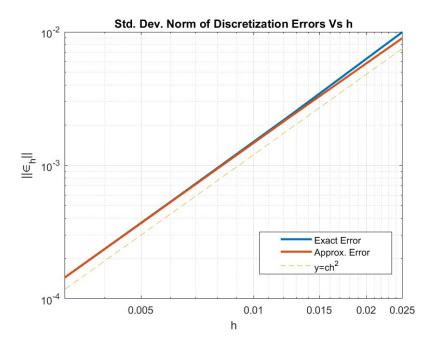


Figure 3: Plot of  $||\vec{\tilde{\epsilon}_h}||$  and  $||\vec{\epsilon_h}||$  vs h

```
% Question 4
1
2
   % Solving for phi_exact
4 \mid N = [41; 81; 161; 321];
   phi_exact = cell(size(N,1),1);
6
7
   for i=1:size(N,1)
8
       n = N(i);
       x = transpose(linspace(0,1,n));
9
       phi = (exp(x*Pe)-1)/(exp(Pe)-1);
10
11
       phi_exact{i,1} = phi;
12
   end
13
14 | % Solving for phi_h
15 | phi_h = cell(size(N,1),1);
16
17 | for i=1:size(N,1)
18
       n = N(i);
                                 % Grid Points = 41
19
       h = 1/(n-1);
                                 % Grid Width = 1/(N-1)
20
21
       % Setting the parameters of CDS
22
       Ae = 1 - (Pe*h)/2;
                              % coeff. of phi(i+1)
23
       Ap = -2;
                                 % coeff. of phi(i)
24
       Aw = 1 + (Pe*h)/2;
                                % coeff. of phi(i-1)
25
26
       % Defining the required sparse matrix A of size nxn
27
       A = zeros(n,n);
28
       A(1,1) = 1; A(n,n) = 1;
29
       for j = 2:n-1
30
           A(j,j) = Ap;
           A(j,j-1) = Aw;
31
32
           A(j,j+1) = Ae;
33
       end
34
35
       % Defining the load vector b of size nx1
36
       b = zeros(n,1);
37
       b(1) = phi_0; b(n) = phi_L;
38
39
       % Using the inbuilt fnc linsolve to solve system of
          equations
40
       soln = linsolve(A, b);
41
       phi_h\{i,1\} = soln;
42
   end
43
44 | % Solving for err_h
45
46 | err_h = cell(size(N,1),1);
```

```
47
48 | for i = 1:size(N,1)
49
       err_h{i,1} = phi_exact{i,1} - phi_h{i,1};
50 end
51
52
   \% Solving for the std. dev. norm of err_h
   err_h_norm = zeros(size(N,1),1);
54
55 for i = 1:size(N,1)
56
       val = norm(err_h{i,1})/sqrt(N(i));
57
       err_h_norm(i) = val;
58 end
59
60 | % Solving for truncation_error
61
62 trunc_error = cell(size(N,1),1);
63
64 | for i = 1:size(N,1)
                                 % Grid Points = 41
65
       n = N(i);
       h = 1/(n-1);
                                 % Grid Width = 1/(N-1)
66
67
       x = transpose(linspace(0,1,n));
68
       % third order derivative values of phi_exact
69
       phi3 = (Pe^3*exp(x*Pe))/(exp(Pe)-1);
70
       % fourth order derivative values of phi_exact
71
       phi4 = (Pe^4*exp(x*Pe))/(exp(Pe)-1);
72
       vec = zeros(n,1);
73
       for j = 2:n-1
74
           % substituting the value of err(i)
75
           vec(j) = (Pe*h*h/6)*phi3(j) - (h*h/12)*phi4(j);
76
       end
77
       trunc_error{i,1} = vec;
78 end
79
80
   % Solving for err_th
81
   err_th = cell(size(N,1),1);
82
83
   h_{vals} = zeros(size(N,1),1);
84
85
   for i = 1:size(N,1)
86
       n = N(i);
                                 % Grid Points = 41
       h = 1/(n-1);
                                 % Grid Width = 1/(N-1)
87
88
       h_{vals}(i) = h;
89
       L = zeros(n,n);
90
       L(1,1) = 1; L(n,n) = 1;
91
       Aw = 1/h^2 + Pe/(2*h);
92
       Ae = 1/h^2 - Pe/(2*h);
93
       Ap = -2/h^2;
94
```

```
95
        for j = 2:n-1
96
            L(j,j) = Ap;
97
            L(j,j-1) = Aw;
            L(j,j+1) = Ae;
98
99
        end
100
        % approx. discretization error
101
        err_th{i,1} = -inv(L)*trunc_error{i,1};
102
    end
103
104
   % Solving for the std. dev. norm of err_th
105
    err_th_norm = zeros(size(N,1),1);
106
107
   for i = 1:size(N,1)
108
        val = norm(err_th{i,1})/sqrt(N(i));
109
        err_th_norm(i) = val;
110 end
111
   \% Plotting err_h_norm and err_th_norm vs x_i
112
   c = 12;
113
114
115 figure;
116 loglog(h_vals, err_h_norm, 'Linewidth', 2);
117
   hold on
118 | loglog(h_vals, err_th_norm, 'Linewidth', 2);
119 hold on
120
   plot(h_vals,c*h_vals.*h_vals, '--', "LineWidth",0.5);
121
   xlabel('h');
122 | ylabel('||epsilon_h||');
123
   title('Std. Dev. Norm of Discretization Errors Vs h');
124 | legend('Exact Error', 'Approx. Error', 'y=ch^2');
125
   grid on;
   legend("Position", [0.57785,0.16243,0.30196,0.17396])
126
```

MATLAB code for the computation of std. dev. norm of  $||\vec{\tilde{\epsilon}}_h||$  and  $||\vec{\epsilon_h}||$