

Chapter 31-2: Solving PDEs Using Differential Quadrature.

Differential quadrature is the approximation of derivatives by using weighted sums of function values. Differential quadrature is of practical interest because it allows one to compute derivatives from noisy data. The name is in analogy with quadrature, meaning numerical integration, where weighted sums are used in methods such as Simpson's method or the Trapezoidal rule. There are various methods for determining the weight coefficients, for example, the Savitzky–Golay filter. Differential quadrature is used to solve partial differential equations. There are further methods for computing derivatives from noisy data.

1.1. Solve the one-dimensional Burgers' equation,

$$\frac{\partial u}{\partial t} - \nu \frac{\partial^2 u}{\partial x^2} + u \frac{\partial u}{\partial x} = 0$$

while observing a boundary condition of

$$u(x, 0) = f(x) \quad \text{in } \Omega$$

and initial condition of

$$u(x, t) = 0 \quad \text{on } \partial\Omega \times (0, T]$$

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In [1]: 1 '''
2       Numerical Solution of Burger's Equation based on Differential Quadrature method.
3       Reference Paper - https://onlinelibrary.wiley.com/doi/10.1002/num.22178
4       Solution taken from the Github repository of mn619.
5       '''
6
```

```
Out[1]: "\n      Numerical Solution of Burger's Equation based on Differential Quadrature method.\n      Refere
nce Paper - https://onlinelibrary.wiley.com/doi/10.1002/num.22178\n" (https://onlinelibrary.wiley.c
om/doi/10.1002/num.22178\n")
```

```
In [2]: 1 import numpy as np
2 import math
3 from mpl_toolkits.mplot3d import Axes3D
4 import matplotlib.pyplot as plt
5
```

```
In [3]: 1 N, M = 30, 30 #Mesh Size
2 iteration = 5 #Number of times to iterate for finding numerical solution
3 mu = 0.1
4 pi = np.pi
5
```

```
In [4]: 1 '''
2       Setting up everything as global variables
3       '''
4
5 x = [(1-np.cos(pi*(i - 1)/(M - 1)))*0.5 for i in range(M + 1)]
6 t = [(1-np.cos(pi*(i - 1)/(N - 1)))*0.5 for i in range(N + 1)]
7
8 A_init = np.zeros((N + 1, N + 1))
9 B_init = [np.zeros((M + 1, M + 1)) for i in range(3)]
10 A = np.zeros((N - 1, N - 1))
11 B = [np.zeros((M - 2, M - 2)) for i in range(3)]
12 alpha = [np.zeros((N - 1, M - 2)) for i in range(iteration + 1)]
13 beta = np.zeros((M - 2, M - 2))
14 F = [np.zeros((N - 1, M - 2)) for i in range(iteration + 1)]
15 U_init = [np.zeros((N + 1, M + 1)) for i in range(iteration + 1)]
16 U = [np.zeros((N - 1, M - 2)) for i in range(iteration + 1)]
17 beta_k = np.zeros(((N - 1)*(M - 2), (N - 1)*(M - 2)))
18
```

```
In [5]: 1 '''
2       All the functions required to initialize the variables
3       '''
4
5 def f(x):
6     return np.sin(np.pi*x)
7
8 def cal_coef(n, mu):
9     ans = 0
10    for i in range(0, 1000):
11        x = (2*i + 1)/2000
12        ans += math.exp(-(1-np.cos(np.pi*x))/(2*np.pi*mu))*np.cos(n*np.pi*x)*1/1000
```

```

13     if(n == 0):
14         return ans
15     else:
16         return 2*ans
17
18 def cal_A(n, m):
19     ans = 1
20     if(n != m):
21         for l in range(1, M + 1):
22             if(l != n and l != m):
23                 ans *= (x[n] - x[l])/(x[m] - x[l])
24             ans *= 1/(x[m] - x[n])
25     else:
26         ans = 0
27         for l in range(1, M + 1):
28             if(l != n):
29                 ans += 1/(x[n] - x[l])
30     return ans
31
32 def cal_B(n, m):
33     ans = 1
34     if(n != m):
35         for l in range(1, N + 1):
36             if(l != n and l != m):
37                 ans *= (t[n] - t[l])/(t[m] - t[l])
38             ans *= 1/(t[m] - t[n])
39     else:
40         ans = 0
41         for l in range(1, N + 1):
42             if(l != m):
43                 ans += 1/(t[m] - t[l])
44     return ans
45
46 def cal_alpha(n, m, k):
47     ans = 0
48     for j in range(1, M + 1):
49         ans += B_init[1][m + 2, j]*U_init[k][n + 2, j]
50     return ans
51
52 def cal_F(n, m, k):
53     return U_init[k][n + 2, m + 2]*alpha[k][n, m] - A_init[n + 2][1]*f(x[m + 2])
54
55 def vec(X):
56     assert(X.shape == (N - 1, M - 2))
57     temp = X.flatten('F')
58     return temp.reshape(-1, 1)
59
60 def diag(X):
61     assert(len(X) == (N - 1)*(M - 2))
62     return np.diagflat(X)
63
64

```

In [6]:

```

1  '''
2      Initialising all the variables
3  '''
4  #Calculate U_init[0]
5  for j in range(1, M + 1):
6      U_init[0][1, j] = f(x[j])
7
8  #Calculate A_init
9  for i in range(1, N + 1):
10     for j in range(1, N + 1):
11         A_init[i, j] = cal_A(i, j)
12
13  #Calculate B_init[1]
14  for i in range(1, M + 1):
15     for j in range(1, M + 1):
16         B_init[1][i, j] = cal_B(i, j)
17
18  #Calculate B_init[2]
19  for i in range(1, M + 1):
20     for j in range(1, M + 1):
21         if( i != j):
22             B_init[2][i, j] = 2*(B_init[1][i, j]*B_init[1][i, i] - B_init[1][i, j]/(t[i] - t[j]))
23     for j in range(1, N + 1):
24         if(j != i):
25             B_init[2][i, i] -= B_init[2][i, j]
26
27  #Calculate A
28  for i in range(0, N - 1):
29     for j in range(0, N - 1):
30         A[i, j] = A_init[i + 2, j + 2]
31
32  #Calculate B[1], B[2]
33  for i in range(0, M - 2):
34     for j in range(0, M - 2):
35         B[1][i, j] = B_init[1][i + 2, j + 2]
36         B[2][i, j] = B_init[2][i + 2, j + 2]
37
38  #Calculate beta

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```
39 beta = -mu*B[2]
40
41 #Calculate alpha[0]
42 for i in range(0, N - 1):
43     for j in range(0, M - 2):
44         alpha[0][i, j] = cal_alpha(i, j, 0)
45
46 #Calculate F[0]
47 for i in range(0, N - 1):
48     for j in range(0, M - 2):
49         F[0][i, j] = cal_F(i, j, 0)
50
51 #Calculate beta_k
52 beta_k = np.kron(beta, np.eye(N - 1))
53
54 #Calculate A_k
55 A_k = np.kron(np.eye(M - 2), A)
56
57 #Calculate B_k
58 B_k = np.kron(B[1], np.eye(N - 1))
59
60
```

```
In [7]: 1 '''
2         This code finds the approximate numerical solution
3     '''
4     def numerical_soln():
5         for k in range(1, iteration + 1):
6             print("Iteration : ", k, "\r", end = "")
7             D1 = np.matmul(diag(vec(U[k - 1])), B_k)
8             D2 = diag(vec(alpha[k - 1]))
9
10            mat = np.zeros(((N - 1)*(M - 2), (N - 1)*(M - 2)))
11
12            for i in range((N - 1)*(M - 2)):
13                for j in range((N - 1)*(M - 2)):
14                    mat[i,j] = beta_k[i,j] + A_k[i,j] + D1[i,j] + D2[i,j]
15            X = np.matmul(np.linalg.inv(mat), vec(F[k-1]))
16            U[k] = X.reshape((N - 1, M - 2), order = 'F')
17
18            for i in range(1, N + 1):
19                for j in range(1, M + 1):
20                    if(i == 1 or j == 1 or j == M):
21                        U_init[k][i,j] = U_init[k - 1][i,j]
22                    else:
23                        U_init[k][i,j] = U[k][i - 2, j - 2]
24            for i in range(N - 1):
25                for j in range(M - 2):
26                    alpha[k][i,j] = cal_alpha(i,j,k)
27                    F[k][i,j] = cal_F(i,j,k)
28            print('\n')
29            return U_init[iteration]
30
31
```

```
In [8]: 1 '''
2         Exact solution as described in the paper
3     '''
4     def exact_soln():
5         c = [cal_coef(i, mu) for i in range(0 ,100)]
6         u = np.zeros((N + 1, M + 1))
7
8         for i in range(1, N + 1):
9             for j in range(1, M + 1):
10                xx = x[j]
11                tt = t[i]
12                numerator = 0
13                denominator = 0
14                for n in range(1, 100):
15                    numerator += c[n]*math.exp(-n*pi*pi*mu*tt)*n*np.sin(n*pi*xx)
16                    denominator += c[n]*math.exp(-n*pi*pi*mu*tt)*np.cos(n*pi*xx)
17                denominator += c[0]
18                u[i][j] = 2*pi*mu*numerator/denominator
19            return u
20
21
```

```
In [9]: 1 u = exact_soln()
2 u_num = numerical_soln()
3
4 print("Error : ", (u_num - u).max())
5
6
```

Iteration : 5

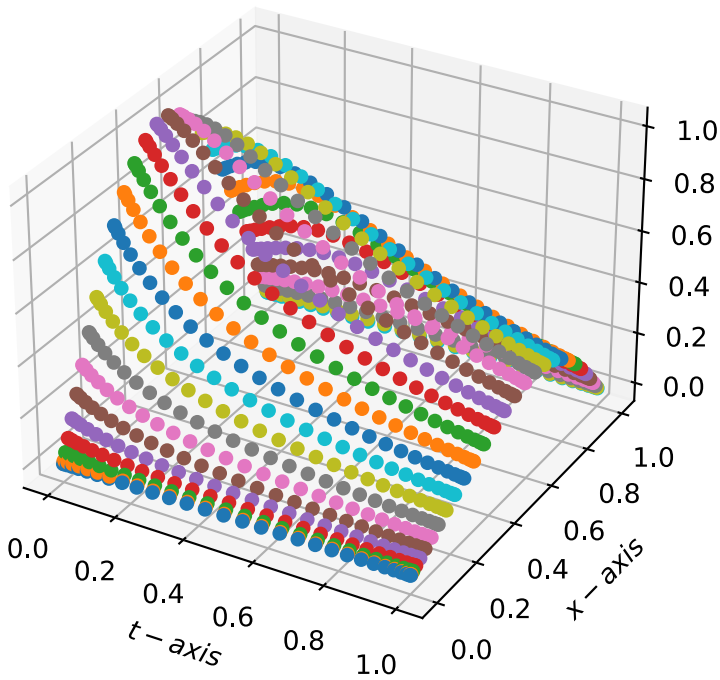
Error : 4.612299431272504e-11

```
In [10]: 1 '''
2         plotting the numerical solution obtained
3     '''
4     %config InlineBackend.figure_formats = ['svg']
5     fig = plt.figure()
```

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6 ax = fig.add_subplot(111, projection='3d')
7
8 for i in range(1, N + 1):
9     for j in range(1, M + 1):
10         ax.scatter(t[i], x[j], u_num[i,j])
11
12 ax.set_xlabel('$t$-axis$')
13 ax.set_ylabel('$x$-axis$')
14 ax.set_zlabel('$u$')
15
16 plt.show()
17
18

```



Problem 1.2. Analyze the free vibration of a simply supported beam, using differential quadrature.

With Octave as the tool, a program authored by Professor Francesco Tornabene is presented. Further information (on shell structures, not beams) can be found in a paper by the same author, *Static analysis of doubly-curved anisotropic shells and panels using CUF approach, differential geometry and differential quadrature method*.

```

In [ ]: 1 % Beam Problem - Simply-supported Euler-Bernoulli Beam
2 clear
3 clc
4 % data
5 L=2;
6 b=0.01;
7 h=0.01;
8 A=b*h;
9 I=1/12*b*h^3;
10 E=2.1e11;
11 ni=0.3;
12 rho=7800;
13 q0=100;
14 % numerical data
15 N=31;
16 m=4;
17 % Weighting coefficients for derivation
18 r=zeros(1,N);
19 for k=1:N
20     r(k)=cos(((2*k-1)*pi)/(2*N));
21 end
22 r=sort(r);
23 % first derivative of Chebyshev polynomial of the first kind
24 dCheb_1=(N*sin(N*acos(r)))./(1 - r.^2).^(1/2);
25 c=zeros(N,N,m);
26 for p=1:m
27     if p==1
28         % first order derivative weighting coefficients
29         for i=1:N
30             for j=1:N
31                 if i==j
32                     else
33 c(i,j,p)=dCheb_1(i)/((r(i)-r(j))*dCheb_1(j));
34 end
35 end
36 end
37 else
38 % higher order derivative weighting coefficients
39 for i=1:N

```

```

40 for j=1:N
41 if i==j
42 else
43 c(i,j,p)=p*(c(i,j,1)*c(i,i,p-1)-c(i,j,p-1)*((r(i)-r(j))^( -1)));
44 end
45 end
46 end
47 end
48 % diagonal weighting coefficients for each order derivative
49 for i=1:N
50 c(i,i,p)=-sum(c(i,:,p));
51 end
52 end
53 % domain length
54 L1=r(N)-r(1);
55 % coordinate transformation
56 x=(r-r(1))/(r(N)-r(1))*L;
57 % weighting coefficients transformation
58 for p=1:m
59 c(:, :,p)=(L1/L)^p*c(:, :,p);
60 end
61 % Strong formulation
62 Ks=E*I*c(:, :,4);
63 Fs=q0*b*ones(N,1);
64 Ms=rho*A*eye(N);
65 % boundary conditions
66 Ks(1,1)=1;
67 Ks(1,2:N)=zeros(1,N-1);
68 Ks(2:N,1)=zeros(N-1,1);
69 Ks(N,N)=1;
70 Ks(N,1:N-1)=zeros(1,N-1);
71 Ks(1:N-1,N)=zeros(N-1,1);
72 Ks(2,:)=E*I*c(1,:,2);
73 Ks(N-1,:)=E*I*c(N,:,2);
74 % boundary conditions for loads
75 Fs(1)=0;
76 Fs(2)=0;
77 Fs(N-1)=0;
78 Fs(N)=0;
79 % static analysis
80 us=Ks\Fs;
81 % dynamic analysis
82 Kdds=Ks(3:N-2,3:N-2);
83 Kdbs=[Ks(3:N-2,1:2) Ks(3:N-2,N-1:N)];
84 Kbbs=[Ks(1:2,1:2) Ks(1:2,N-1:N); Ks(N-1:N,1:2) Ks(N-1:N,N-1:N)];
85 Kbds=[Ks(1:2,3:N-2); Ks(N-1:N,3:N-2)];
86 Mddgs=Ms(3:N-2,3:N-2);
87 % eigenvalue problem
88 Hs=sqrt(eig(Mddgs\(-Kdbs*(Kbbs\Kbds)+Kdds)));
89 % circular frequencies
90 omegas=sort(Hs);
91 % frequencies
92 freqs=real(omegas/(2*pi));
93 % Exact solution
94 u_ex=q0*b*L^4/(24*E*I)*(x.^4/L^4-2*x.^3/L^3+x/L);
95 % Exact solution: dynamic analysis
96 freq_ex=zeros(N-4,1);
97 for i=1:N-4
98 freq_ex(i,1)=(((i^2)*(pi/2))/(L^2))*sqrt((E*I)/(rho*A));
99 end
100 Freqs=freqs./freq_ex;
101 % static deformation
102 figure
103 plot(x,u_ex,'r-',x,us,'kd')
104 set(gca,'Ydir','reverse')
105 % dynamic frequencies comparison
106 figure
107 plot(((1:N-4)-1)/(N-5),0*(Freqs-1)*100,'r-',((1:N-4)-1)/(N-5),(Freqs-1)*100,'kd-')
108 ylim([-10 10])
109 xlim([0 1])
110
111 #add some space to allow the PDF to be printed without glaring gap
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