Math 104C HW5

May 12, 2022

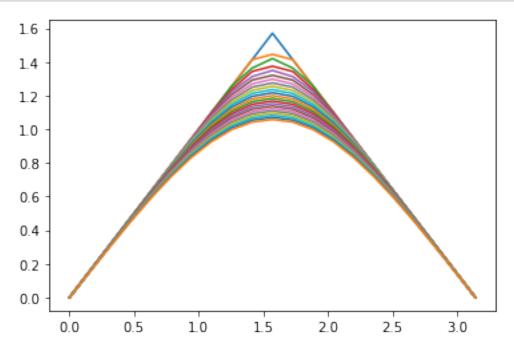
Math 104C Homework #5 Aidan Baker (perm:8141905)

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[84]: import numpy as np import matplotlib.pyplot as plt import math
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[85]: ##INITIAL BOUNDARY VALUE PROBLEM
      def initial_value(M,delta_x):
          result = []
          for i in range(M+1):
              x = i * delta_x
              u = x \text{ if } x < \text{math.pi/2 else math.pi} - x
              result.append(u)
          return np.array(result)
      ##EXPLICIT FINITE DIFFERENCE SCHEME
      def explicit_forward(u_init,alpha,delta_t,N):
          current = u_init
          for _ in range(1,N):
              next = np.zeros_like(current)
          for j in range(1,M):
              next[j] = current[j] + alpha *(current[j-1] - 2*current[j]+current[j+1])
          current = next
          return current
      ##OUR INITIAL CONDITIONS
      D=1
      M = 20
      delta_x = math.pi/M
      alpha = 0.4
      delta_t = alpha * (delta_x **2) / D
      N = 150
      x_values = [i*delta_x for i in range(M+1)]
      initial = initial_value(M,delta_x)
      plt.plot(x_values,initial)
```

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for i in range(M+1):
    u = explicit_forward(initial,alpha,delta_t,N)
    plt.plot(x_values,u)
    initial = u

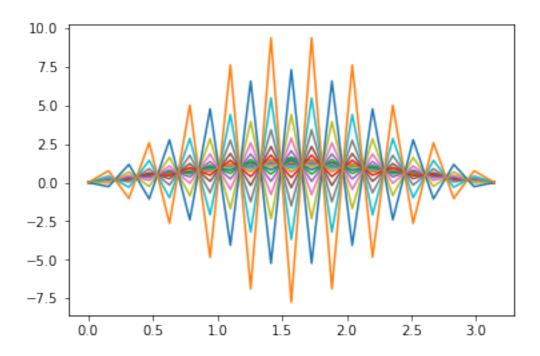
## WE SEE THAT WE HAVE A STABLE RESULT IF ALPHA <= .5. If we take a value over .
    →5 we lose stability</pre>
```



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[86]: ## REPEATING PROCESS WITH ALPHA OVER .5 Shows how we lose stability

D=1
M=20
delta_x = math.pi/M
alpha = 0.6
delta_t = alpha * (delta_x **2) / D
N=150
x_values = [i*delta_x for i in range(M+1)]
initial = initial_value(M,delta_x)
plt.plot(x_values,initial)

for i in range(M+1):
    u = explicit_forward(initial,alpha,delta_t,N)
    plt.plot(x_values,u)
    initial = u
```



```
[87]: def tridiag_mat(M,alpha):
          A = -2 * alpha * np.eye(M)
          for i in range(M-1):
              A[i][i+1] = 1.0
              A[i+1][i] = 1.0
          return np.eye(M) - A
      def system_solver(A,b):
          c = np.zeros(len(b)-1)
          d = np.zeros(len(b))
          for i in range(len(b)):
              if i == 0:
                  c[i] = A[i][i+1]/A[i][i]
                  d[i] = b[i]/A[i][i]
              elif i == len(b)-1:
                  d[i] = (b[i]-A[i][i-1]*d[i-1])/(A[i][i]-A[i][i-1]*c[i-1])
              else:
                  c[i] = A[i][i+1]/(A[i][i] - A[i][i-1]*c[i-1])
                  d[i] = (b[i]-A[i][i-1]*d[i-1])/(A[i][i] - A[i][i-1]*c[i-1])
          x = np.zeros(len(b))
          for i in reversed(range(len(b))):
              if i == len(b)-1:
                  x[i] = d[i]
              else:
                  x[i] = d[i] - c[i]*x[i+1]
```

```
return x
def backward_difference(u_init,alpha,delta_t,N):
    M = len(u_init)
    current = u_init
    tridiag = tridiag_mat(M,alpha)
    for i in range(N):
        current = system_solver(tridiag,current)
    return current
delta_x = 0.5
delta_t = delta_x
D = 1
alpha = D * (delta_t)/(delta_x **2)
N = 150
x_values = np.array([i*delta_x for i in range(M+1)])
approx = backward_difference(initial,alpha,delta_t,N)
plt.plot(x_values,approx)
## WE SEE THAT THE BACKWARDS DIFFERENCE METHOD IS ALSO STABLE for delta_t =
\rightarrow delta_x and D=1
```

[87]: [<matplotlib.lines.Line2D at 0x7fcef04c89d0>]

