The Finite Difference Method

A finite difference for a function f(x) is an expression of the form f(x + s) - f(x + t). Finite differences can give a good approximation of derivatives.

Suppose we have a function u(x), defined on an interval [a, b]. Let $a = x_{-1}, x_0, x_1, \ldots x_{n-1} = b$ be a grid of n + 1 evenly spaced points, with $x_i = a + (i + 1)h$, h = (b - a)/n.

You are used to seeing the derivative u'(x) which can be written as

$$u'(x) = \lim_{h \to \infty} \frac{u(x+h) - u(x)}{h} = \lim_{h \to \infty} \frac{u(x+h) - u(x-h)}{2h}.$$

Since we are interested in the derivative at certain fixed points x_i , we can consider the approximation of u'(x) using finite differences. We first write the Taylor polynomial expansion of u(x+h) and u(x-h) centered at x. This gives

$$u(x+h) = u(x) + u'(x)h + \frac{1}{2}u''(x)h^2 + \frac{1}{6}u'''(x)h^3 + \mathcal{O}(h^4)$$
(7.1)

$$u(x-h) = u(x) - u'(x)h + \frac{1}{2}u''(x)h^2 - \frac{1}{6}u'''(x)h^3 + \mathcal{O}(h^4)$$
(7.2)

Subtracting (7.2) from (7.1) and rearranging gives

$$u'(x) = \frac{u(x+h) - u(x-h)}{2h} + \mathcal{O}(h^2).$$

From the Taylor expansion, this term has error $E(h) = \mathcal{O}(h^2)$. In terms of our grid points $\{x_i\}$, we can rewrite u'(x) as $u'(x_i)$ and

$$u'(x_i) = \frac{u(x_i+h) - u(x_i-h)}{2h} = \frac{u(x_{i+1}) - u(x_{i-1})}{2h}$$

We won't worry about the derivative at the endpoints, $u'(x_{-1})$ and $u'(x_{n-1})$. This allows us to write the set of points $\{u'(x_i)\}$ as the solution to a system of equations

$$\frac{1}{2h} \begin{bmatrix} -1 & 0 & 1 & & \\ & -1 & 0 & 1 & & \\ & & & \ddots & \ddots & \\ & & & -1 & 0 & 1 \\ & & & & -1 & 0 & 1 \end{bmatrix} \cdot \begin{bmatrix} u(x_{-1}) \\ u(x_{0}) \\ \vdots \\ u(x_{n-2}) \\ u(x_{n-1}) \end{bmatrix} = \begin{bmatrix} u'(x_{0}) \\ u'(x_{1}) \\ \vdots \\ u'(x_{n-3}) \\ u'(x_{n-2}) \end{bmatrix} .$$
(7.3)

This can be rewritten with an $(N-1) \times (N-1)$ tridiagonal matrix on the left.

$$\frac{1}{2h} \begin{bmatrix} 0 & 1 & & \\ -1 & 0 & 1 & & \\ & -1 & 0 & 1 & & \\ & & -1 & 0 & 1 & \\ & & & -1 & 0 & 1 \\ & & & & -1 & 0 \end{bmatrix} \cdot \begin{bmatrix} u(x_0) \\ u(x_1) \\ \vdots \\ u(x_{n-3}) \\ u(x_{n-2}) \end{bmatrix} + \begin{bmatrix} -u(x_{-1})/(2h) \\ 0 \\ \vdots \\ 0 \\ u(x_{n-1})/(2h) \end{bmatrix} = \begin{bmatrix} u'(x_0) \\ u'(x_1) \\ \vdots \\ u'(x_{n-3}) \\ u'(x_{n-2}) \end{bmatrix} . \quad (7.4)$$

Next we will consider the matrix representation for u''(x). If we let

$$u'(x) = \frac{u(x + \frac{h}{2}) - u(x - \frac{h}{2})}{h}$$

then

$$u''(x) = \frac{u'(x+\frac{h}{2}) - u'(x-\frac{h}{2})}{h} = \frac{\frac{u((x+\frac{h}{2})+\frac{h}{2}) - u((x+\frac{h}{2})-\frac{h}{2})}{h} - \frac{u((x-\frac{h}{2})+\frac{h}{2}) - u((x-\frac{h}{2})-\frac{h}{2})}{h}}{h}$$
$$= \frac{u(x+h) - 2u(x) + u(x-h)}{h^2},$$

with error $E(h) = \mathcal{O}(h^3)$. You can achieve the same result by again consider the Taylor polynomial expansion and adding (7.1) and (7.2) and rearranging. Thus

$$u''(x_i) = \frac{u(x_i+h) - 2u(x_i) + u(x_i-h)}{h^2} = \frac{u(x_{i+1}) - 2u(x_i) + u(x_{i-1})}{h^2}, \quad i = 0, \dots, n-2.$$

Again ignoring the second derivative at the endpoints, this can be written in matrix form as

$$\frac{1}{h^{2}} \begin{bmatrix}
1 & -2 & 1 & & \\
& 1 & -2 & 1 & & \\
& & \ddots & \ddots & \ddots & \\
& & & 1 & -2 & 1 \\
& & & & 1 & -2 & 1
\end{bmatrix} \cdot
\begin{bmatrix}
u(x_{-1}) \\ u(x_{0}) \\ \vdots \\ u(x_{n-2}) \\ u(x_{n-1})\end{bmatrix} =
\begin{bmatrix}
u''(x_{0}) \\ u''(x_{1}) \\ \vdots \\ u''(x_{n-3}) \\ u''(x_{n-2})\end{bmatrix}.$$
(7.5)

This can also be written as an $(N-1) \times (N-1)$ tridiagonal matrix on the left.

$$\frac{1}{h^{2}} \begin{bmatrix} -2 & 1 & & \\ 1 & -2 & 1 & & \\ & 1 & -2 & 1 & \\ & & 1 & -2 & 1 \\ & & & 1 & -2 \end{bmatrix} \cdot \begin{bmatrix} u(x_{0}) \\ u(x_{1}) \\ \vdots \\ u(x_{n-3}) \\ u(x_{n-2}) \end{bmatrix} + \begin{bmatrix} u(x_{-1})/h^{2} \\ 0 \\ \vdots \\ 0 \\ u(x_{n-1})/h^{2} \end{bmatrix} = \begin{bmatrix} u''(x_{0}) \\ u''(x_{1}) \\ \vdots \\ u''(x_{n-3}) \\ u''(x_{n-2}) \end{bmatrix}$$
(7.6)

Problem 1. Let $u(x) = \sin((x + \pi)^2 - 1)$. Use (7.3) - (7.6) to approximate $\frac{1}{2}u'' - u'$ at the grid points where a = 0, b = 1, and n = 10. Graph the result.

Suppose that instead of knowing the function u(x), we know that $\frac{1}{2}u'' - u' = f$, where the function f(x) is given. How do we solve for u at the grid points?

Finite Difference Methods

Numerical methods for differential equations seek to approximate the exact solution u(x) at some finite collection of points in the domain of the problem. Instead of analytically solving the original differential equation, defined over an infinite-dimensional function space, they use a simpler finite system of algebraic equations to approximate the original problem.

Consider the following differential equation:

$$\epsilon u''(x) - u(x)' = f(x), \quad x \in (0,1), u(0) = \alpha, \quad u(1) = \beta.$$
(7.7)

Equation (7.7) can be written Du = f, where $D = \epsilon \frac{d^2}{dx^2} - \frac{d}{dx}$ is a differential operator defined on the infinite-dimensional space of functions that are twice continuously differentiable on [0, 1] and satisfy $u(0) = \alpha$, $u(1) = \beta$.

We look for an approximate solution $\{U_i\}_{i=-1}^{N-1}$, where

$$U_i = u(x_i)$$

on an evenly spaced grid of N subintervals, $a = x_{-1}, x_0, \ldots, x_{N-1} = b$ with $h = x_{i+1} - x_i$ for each *i*. Our finite difference method will replace the differential operator $D = \epsilon \frac{d^2}{dx^2} - \frac{d}{dx}$, defined on an infinite-dimensional space of functions, with difference operators defined on a finite vector space (the space of grid functions $\{U_i\}_{i=-1}^{N-1}$). To do this, we replace derivative terms in the differential equation with appropriate difference expressions.

Recalling that

$$\frac{d^2}{dx^2}u(x_i) = \frac{u(x_{i+1}) - 2u(x_i) + u(x_{i-1})}{h^2} + \mathcal{O}(h^2),$$
$$\frac{d}{dx}u(x_i) = \frac{u(x_{i+1}) - u(x_{i-1})}{2h} + \mathcal{O}(h^2).$$

we define the finite difference operator D_h by

$$D_h U_i = \epsilon \frac{1}{h^2} \left(U_{i+1} - 2U_i + U_{i-1} \right) - \frac{1}{2h} \left(U_{i+1} - U_{i-1} \right).$$
(7.8)

Thus we discretize equation (7.7) using the equations

$$\frac{\epsilon}{h^2}(U_{i+1} - 2U_i + U_{i-1}) - \frac{1}{2h}(U_{i+1} - U_{i-1}) = f(x_i), \quad i = 0, \dots, N-2,$$

along with boundary conditions $U_{-1} = \alpha$, $U_{N-1} = \beta$.

This gives N + 1 equations and N + 1 unknowns, and can be written in matrix form as

$$\frac{1}{h^2} \begin{bmatrix} h^2 & 0 & 0 & \dots & 0 \\ (\epsilon+h/2) & -2\epsilon & (\epsilon-h/2) & \dots & 0 \\ \vdots & & \ddots & & \vdots \\ 0 & \dots & (\epsilon+h/2) & -2\epsilon & (\epsilon-h/2) \\ 0 & \dots & 0 & h^2 \end{bmatrix} \cdot \begin{bmatrix} U_{-1} \\ U_0 \\ \vdots \\ U_{N-2} \\ U_{N-1} \end{bmatrix} = \begin{bmatrix} U_{-1} \\ f(x_0) \\ \vdots \\ f(x_{N-2}) \\ U_{N-1} \end{bmatrix} .$$

We can further modify the system to obtain an $(N-1) \times (N-1)$ tridiagonal matrix on the left:

$$\frac{1}{h^{2}} \begin{bmatrix}
-2\epsilon & (\epsilon - h/2) & 0 & \dots & 0 \\
(\epsilon + h/2) & -2\epsilon & (\epsilon - h/2) & \dots & 0 \\
\vdots & & \ddots & & \vdots \\
0 & \dots & (\epsilon + h/2) & -2\epsilon & (\epsilon - h/2) \\
0 & \dots & (\epsilon + h/2) & -2\epsilon
\end{bmatrix} \begin{bmatrix}
U_{0} \\
U_{1} \\
\vdots \\
U_{N-3} \\
U_{N-2}
\end{bmatrix} \\
(N-1)\times(N-1) & (N-1)\times1 \\
= \begin{bmatrix}
f(x_{0}) - U_{-1}(\epsilon + h/2)/h^{2} \\
f(x_{1}) \\
\vdots \\
f(x_{N-3}) \\
f(x_{N-2}) - U_{n-1}(\epsilon - h/2)/h^{2}
\end{bmatrix} .$$
(7.9)

Problem 2. Use equation (7.9) to solve the singularly perturbed BVP (7.7) with $\epsilon = 1/10$, f(x) = -1, $\alpha = 1$, and $\beta = 3$. Graph the solution. This BVP is called singularly perturbed because of the location of the parameter ϵ . For $\epsilon = 0$ the ODE has a drastically different character - it then becomes first order, and can no longer support two boundary conditions.



Figure 7.1: The solution to Problem 2. The solution gets steeper near x = 1 as ϵ gets small.

A heuristic test for convergence

The finite differences used above are second order approximations of the first and second derivatives of a function. It seems reasonable to expect that the numerical solution would converge at a rate of about $\mathcal{O}(h^2)$. How can we check that a numerical approximation is reasonable?



Figure 7.2: Demonstration of second order convergence for the finite difference approximation (7.8) of the BVP given in (7.7) with $\epsilon = .5$.

Suppose a finite difference method is $\mathcal{O}(h^p)$ accurate. This means that the error $E(h) \approx Ch^p$ for some constant C as $h \to 0$ (in other words, for h > 0 small enough).

So compute the approximation y_k for each stepsize h_k , $h_1 > h_2 > \ldots > h_m$. y_m should be the most accurate approximation, and will be thought of as the true solution. Then the error of the approximation for stepsize h_k , k < m, is

$$E(h_k) = \max(|y_k - y_m|) \approx Ch_k^p,$$

$$\log(E(h_k)) = \log(C) + p\log(h_k).$$

Thus on a log-log plot of E(h) vs. h, these values should be on a straight line with slope p when h is small enough to start getting convergence. We should note that demonstrating second-order convergence does NOT imply that the numerical approximation is converging to the correct solution.

Problem 3. Return to problem 2. How many subintervals are needed to obtain 4 digits of accuracy?

This is a question about the convergence of your solution. The following code generates the log-log plot in Figure 7.2, and demonstrates second-order convergence for our finite difference approximation of (7.7). Use this code to determine what h (and hence what N) is needed for the error to be less than 10^{-4} . You don't need to return the value of h, but make sure you understand by looking at the plot.

NOTE: The function bvp is not provided; you need to use your code from problem 2 to

define it. Make sure your function is compatible with the code below. It must take 5 parameters as input and return the solution.

```
num_approx = 10 # Number of Approximations
N = 5*np.array([2**j for j in range(num_approx)])
h, max_error = (1.-0)/N[:-1], np.ones(num_approx-1)
# Best numerical solution, used to approximate the true solution.
# bvp returns the grid, and the grid function, approximating the solution
# with N subintervals of equal length.
num_sol_best = bvp(lambda x:-1, epsilon=.1, alpha=1, beta=3, N=N[-1])
for j in range(len(N)-1):
    num_sol = bvp(lambda x:-1, epsilon=.1, alpha=1, beta=3, N=N[j])
    max_error[j] = np.max(np.abs(num_sol_num_sol_best[::2**(num_approx-j-1)]))
plt.loglog(h,max_error,'.-r',label="$E(h)$")
plt.loglog(h,h**(2.),'-k',label="$h^{\, 2}$")
plt.xlabel("$h$")
plt.legend(loc='best')
plt.show()
print("The order of the finite difference approximation is about ",
    (np.log(max_error[0])-np.log(max_error[-1]))/(np.log(h[0])-np.log(h[-1])),
    ".")
```

Problem 4. Extend your finite difference code to the case of a general second order linear BVP with boundary conditions (These boundary conditions are sometimes called Dirichlet conditions):

$$a_1(x)y'' + a_2(x)y' + a_3(x)y = f(x), \quad x \in (a, b),$$

 $y(a) = \alpha, \quad y(b) = \beta.$

Use your code to solve the boundary value problem

$$\epsilon y'' - 4(\pi - x^2)y = \cos x,$$

$$y(0) = 0, \quad y(\pi/2) = 1,$$

for $\epsilon = 0.1$. Be sure to modify the finite difference operator D_h in (7.8) correctly.

The next few problems will help you troubleshoot your finite difference code.



Figure 7.3: The solution to Problem 4.

Problem 5. Numerically solve the boundary value problem $\epsilon y'' + xy' = -\epsilon \pi^2 \cos(\pi x) - \pi x \sin(\pi x),$ $y(-1) = -2, \quad y(1) = 0,$ for $\epsilon = 0.1, 0.01$, and 0.001.

Problem 6. Numerically solve the boundary value problem

$$(\epsilon + x^2)y'' + 4xy' + 2y = 0,$$

$$y(-1) = 1/(1+\epsilon), \quad y(1) = 1/(1+\epsilon),$$

for $\epsilon = 0.05, 0.02$.



Figure 7.4: The solution to Problem 5.



Figure 7.5: The solution to Problem 6.