Due: April 18, 2014 (Friday, because of Exam 2) · Out: April 6, 2014

Problem 1: Interpolation, Newton and Cubic Spline (20 points)

(a) (9 points) Prove that the formula using divided differences:

$$f[t_1, t_2, \dots, t_k] := \frac{f[t_2, t_3, \dots, t_k] - f[t_1, t_2, \dots, t_{k-1}]}{t_k - t_1}$$
$$f[t_j] := f(t_j)$$

indeed gives the coefficient of the jth basis function in the Newton interpolation polynomial. *Hint:* Use induction.

- (b) $(4 \times 1.5 = 6 \text{ points})$ Given the three data points (-1, 1), (0, 0), (1, 1), determine the interpolating polynomial of degree two:
 - (i) Using the monomial basis
 - (ii) Using the Lagrange basis
 - (iii) Using the Newton basis
 - (iv) Show that the three representations give the same polynomial.
- (c) (5 points) Consider interpolating a given set of data points (x_i, y_i) , i = 1, ..., n using natural cubic splines. Write a code to set up and solve the linear system that performs this interpolation. Plot the resulting cubic spline along with the data. For the data, pick n = 6 random points $(x_i)_{i=1}^n$ on [0, 1) with values $(y_i)_{i=1}^n$ in [0, 1).

Hint: Make sure you sort the x_i 's after you draw the random numbers and before you start constructing the spline, to avoid confusing your spline construction code.

Problem 2: Numerical Quadrature (25 points)

The goal of this problem is to compute π using numerical integration using the following equation:

$$\int_0^1 \frac{4}{1+x^2} \, dx = \pi$$

and several different quadrature schemes. Your task is to write a function that accomplishes this using each of the following different schemes.

(a) Composite midpoint rule.

Hint: Perform your own tests by comparing using scipy.integrate.fixed_quad(f, a, b, n=1). No need to report these.

- (b) Composite trapezoid rule. (Compare with scipy.integrate.trapz(f(X), X) and don't report the test results.)
- (c) Composite Simpson rule. (Compare with scipy.integrate.simps(f(X), X).)
- (d) Monte Carlo method.

The Monte Carlo methods work by drawing n uniform samples $(x_i)_{i=1}^n$ using a uniform distribution on the integration domain ([0, 1] in this case) and using the approximation

$$\int_0^1 f(x) \, dx \approx \frac{1}{n} \sum_{i=1}^n f(x_i).$$

For the purpose of comparing with the other methods, use h = 1/n as a rough analog of the 'mesh spacing' h.

For parts (a) through (d), answer the following prompts for each method:

- (i) Compute the approximate value for π using the method with various values of h.
- (ii) Characterize the error for each method as a function of h.

Use two different approaches:

- Create a clearly labeled log-log plot. Use the same plot (with a legend). Use the same plot for all four parts.
- Compute the *empirical order of convergence*. Use this procedure to estimate this quantity:

We assume that the error depends on the mesh spacings h as $E(h) \approx Ch^p$ for some unknown power p. Taking the log of this approximate equality reveals a linear function in p:

$$E(h) \approx Ch^p \quad \Longleftrightarrow \quad \log E(h) \approx \log(C) + p \log(h).$$

You can now either do a least-squares fit for $\log C$ and p from a few data points (h, E(h)) (more accurate, more robust), or you can use just two grid sizes h_1 and h_2 , and estimate the slope: (less accurate, less robust)

$$p \approx \frac{\log(E(h_2)/E(h_1))}{\log(h_2/h_1)}.$$

This is called the *empirical order of convergence* or *EOC*.

(iii) Is there a point beyond which decreasing h yields no further improvement?

Problem 3: Gaussian Quadrature (10 + 5 + 5 = 20 points)

(a) Let p be a real polynomial of degree n such that:

$$\int_{a}^{b} p(x) x^{k} dx = 0, \quad k = 0, \dots, n - 1.$$

- (i) Show that the *n* zeros of *p* are real, simple and lie in the open interval (a, b). *Hint*: Consider the polynomial $q_k(x) = (x - x_1)(x - x_2) \cdots (x - x_k)$, where x_i , $i = 1, \ldots, k$ are the roots of *p* in (a, b). Where does $p(x)q_k(x)$ change signs?
- (ii) Show that the n-point interpolatory quadrature on [a, b] whose nodes are the zeros of p has degree 2n 1.

Hint: Consider the quotient and remainder polynomials when a given polynomial is divided by p.

(b) Write a function that computes a given integral using Gaussian quadrature. Your Gaussian quadrature function should nominally have the following signature:

```
def gauss_quad(f, n, ...):
# add code here
```

You can obtain the nodes using the following snippet:

```
nodes = scipy.scpecial.legendre(n).weights[:, 0]
```

- (c) Use gauss_quad to integrate the following two functions for various quadrature orders n = 1, 2, ..., 100. For the integrals of both these functions, create a log-plot of error in the Gaussian quadrature versus the order. Does the error obey $E(h) \approx Ch^p$ for some p, in each of the cases?
 - (i) $f(x) = \sin 2\pi x$ on [-1, 1]
 - (ii) g(x) = |x| on [-1, 1]

Problem 4: Numerical Differentiation (15 points)

- (a) (5 points) Given a sufficiently smooth function $f : \mathbb{R} \to \mathbb{R}$, use Taylor series to derive a second order accurate, one-sided difference approximation to f'(x) in terms of the values of f(x), f(x+h) and f(x+2h).
- (b) (5 points) Write a Python function to implement this difference scheme for a function f defined on [-1, 1].
- (c) (5 points) Discretize (-1, 1) using a uniformly spaced mesh with spacing $h = 2^{-k}, k = 3, ..., 20$ and obtain the derivative at the sampled points using the above function for $f(x) = \sin x$. Obtain the error in the derivative using the max norm (i.e. error is measured to be the maximum of the absolute differences) and make a plot of error versus h. What is the expected order of convergence? What is the convergence rate that you obtain? Is the error monotonically decreasing? Explain your observations.

Problem 5: Initial Value Problems (20 points)

Consider the initial value problem

$$y' = -200 t y^2$$
, with $y(0) = 1$.

This IVP has the analytical solution

$$y(t) = \frac{1}{1 + 100 t^2}$$

Implement the following methods to numerically integrate the ODE from t = 0 to t = 1:

- (a) Forward Euler method,
- (b) Backward Euler method,
- (c) Fourth-order Runge-Kutta method.

For each method do the following:

- Use step sizes of h = 0.125, 0.25, 0.5, 1.
- Plot the numerical solution y_h versus t for these various values of h. You may plot y on same plot for comparison.
- Compute the error at t = 1 between y_h and y for each h and plot versus h on a loglog plot.
- Based on the error data, what is the order of accuracy for each method?
- Explain whether each method is stable or not.

Your writeup should consist of plots of y_h against t (one plot for each method and each of the values of h), and three error plots—one per method. You should also provide comments on stability and accuracy of each method.