Assignment #8

Due Friday, 14 April at the start of class

Please read Chapters 7 and 8 of LeVeque. Most of Chapter 7 is important, but I am de-emphasizing multistep methods. In Chapter 8, but I will not cover 8.6.

P29. Consider the " θ -method" for u' = f(u, t), namely

$$U^{n+1} = U^n + k[(1-\theta)f(U^n, t_n) + \theta f(U^{n+1}, t_{n+1})],$$

where $0 \le \theta \le 1$ is a fixed parameter.

- a) Cases $\theta = 0, 1/2, 1$ are all familiar methods. Name them. Then show that the θ -method is A-stable for $\theta \ge 1/2$.
- **b)** Plot the (absolute) stability regions for $\theta = 0, 1/4, 1/2, 3/4, 1$ and briefly comment on how the stability region will look for other values of θ (in [0,1]).
- P30. Consider this one-step (Runge-Kutta) method, the implicit midpoint method,

$$U^* = U^n + \frac{k}{2} f(U^*, t_n + k/2),$$

$$U^{n+1} = U^n + k f(U^*, t_n + k/2).$$

The first equation (stage) is Backward Euler to determine an approximation to the value at the midpoint in time and the second stage is the midpoint method using this value.

- a) Determine the order of accuracy of this method.
- **b)** Determine the stability region.
- c) Is this method A-stable? Is it L-stable?
- **P31.** This problem is about a "typical" non-stiff ODE system, the *Lotka-Volterra predator-prey model*. The equations are

$$\frac{dx}{dt} = \alpha x - \beta xy$$
$$\frac{dy}{dt} = \delta xy - \gamma y$$

where x(t) is the number of prey at time t and y(t) the predators. To keep things simple, we use $\alpha = 2/3, \beta = 4/3, \gamma = 1 = \delta$ and initial conditions x(0) = y(0) = 2.

Suppose we want to find x(t) and y(t) for the interval $0 \le t \le 20$, i.e. $t_0 = 0$ and $t_f = 20$. This question is about the choice of numerical methods for such a problem.

a) Using forwardeuler.m from the solutions to Assignment #7, or any other implementation of forward Euler (FE), confirm that N=50 step results, i.e. with $k=t_f/N=20/50=0.4$, produces exploding garbage. (How do you know it is garbage? Observe that with larger N values the results are consistent and qualitatively-different.)

¹The famous case is snowshoe hares and lynx: http://www.pnas.org/content/94/10/5147.full.

b) Despite the explosion in part **a)**, no expert would describe this problem as "stiff." We will show that, on this problem, implicitness does not yield a good solution with reduced computational cost; compare **P32** below for the opposite result. For concrete comparison of computational costs, fix N = 50 (and thus k = 0.4) for the rest of the problem, both parts **b)** and **c)**.

Implement both the backward Euler (BE) and RK2 methods³ and apply them to this problem with N=50. Plot the results and comment on their quality. (*Hints*. For BE you will need to solve nonlinear equations at each step, so a Newton iteration will be part of your code. You will find that using the current values as the initial iterate leads to convergence of the Newton iteration in a small, fixed number of steps; report what number of steps suffices, and why, and fix that number in your code. To evaluate "quality", compare to results for a larger N.)

c) Now count function evaluations for the N=50 runs you just plotted, for all three methods (FE, BE, RK2). A fair way to count is to count "1" for each component of the right-hand-side of the ODE system, and count "1" for each evaluation of a Jacobian entry. Show that BE is actually more expensive than RK2 and that RK2 results are just as good.

In conclusion, on *this* ODE system, one's time is better spent improving order (FE \rightarrow RK2) than adding implicitness (FE \rightarrow BE).

P32. For a definitely-stiff problem, we apply the method-of-lines (MOL) to the heat equation,

$$u_t \stackrel{*}{=} u_{xx},$$

a PDE. Here u(t,x) is the temperature in a rod of length one $(0 \le x \le 1)$ and we set boundary temperatures to zero (u(t,0) = 0 and u(t,1) = 0). For an initial temperature distribution we set one part hotter than the rest:

$$u(0, x) = \begin{cases} 1, & 0.25 < x < 0.5, \\ 0, & \text{otherwise.} \end{cases}$$

Suppose we seek u(1, x), i.e. we set $t_f = 1$.

We discretize the spatial derivatives using the notation from Chapter 2. That is, suppose there are m+1 subintervals, let h=1/(m+1), and let $x_j=jh$ for $j=0,1,2,\ldots,m+1$. Let $U_j(t)\approx u(t,x_j)$. By eliminating unknowns $U_0=0$ and $U_{m+1}=0$, and noting that the time derivatives remain as derivatives, from * we get a linear ODE system of dimension m,

$$U(t)' = AU(t)$$

where $U(t) \in \mathbb{R}^m$ and A is exactly the matrix in equation (2.10) in the textbook. For a given m, the components of U(0) can be computed from the above formula for u(0,x).

Finally the exercise itself: Implement both FE and BE on the above MOL heat equation system. In particular, store A using sparse storage and solve the equation (in BE) using backslash, which will automatically detect that the matrix is tridiagonal and then use the well-known efficient solution method for tridiagonal systems.

Now consider the m=100 case. For BE, compute and show the solution using N=100 time steps. For FE, N=100 will generate extraordinary explosion. Instead, determine the time step k from the eigenvalues of A and the stability region of FE. Then compare the computational costs by counting floating-point multiplications. You will conclude that an implicit is indeed effective in this case.

²The explosion does reflect a "stability problem," but nonlinearity confounds any easy classification.

³Note BE is *L*-stable while RK2 is not even *A*-stable.

⁴For an $m \times m$ tridiagonal matrix A, a Av costs 3m multiplications while $A \setminus v$ costs 5m multiplications.