# Time-dependent problems & PETSC TS time-stepping for ODEs and PDEs

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### outline for today

### Chapter 6 of book:

- ODE initial value problem examples and methods
- PETSc TS object does time-stepping
  - explicit Runge-Kutta methods
  - implicit methods (backward Euler,  $\theta$ , IMEX, gl)
    - require solving equations at each time step: SNES
- heat.c solves classical heat equation

$$u_t = D\nabla^2 u + f(x, y)$$

- diffusions are stiff, so implicit appropriate
- pattern.c solves system of coupled nonlinear diffusions for pattern formation:

$$u_t = D_u \nabla^2 u - uv^2 + F(1 - u)$$
  
$$v_t = D_v \nabla^2 v + uv^2 - (F + k)v$$

### systems of ODEs

we can solve ODE systems of form

$$\mathbf{y}' = \mathbf{g}(t, \mathbf{y})$$

- $\circ$   $\mathbf{y}(t) \in \mathbb{R}^N$
- ODE describes motion of point in N dimensions
- g can be linear or nonlinear in t or y
- initial value problem specifies solution at one time:

$$\mathbf{y}(t_0) = \mathbf{y}_0$$

- exists unique solution if  $\mathbf{g}(t, \mathbf{y})$  is continuous in both inputs and Lipschitz in  $\mathbf{y}$  ... that is, you can *predict*
- ∘ for  $N < \infty$  cases (i.e. *not* PDEs), one can generally go forward or backward from  $t = t_0$

## example with fixed size N = 2

• example with initial time  $t_0 = 0$ 

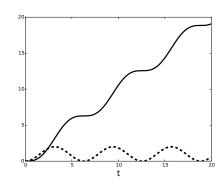
$$\mathbf{y}' = \begin{bmatrix} y_1 \\ -y_0 + t \end{bmatrix}, \quad \mathbf{y}(0) = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

exact solution is

$$\mathbf{y}(t) = \begin{bmatrix} t - \sin t \\ 1 - \cos t \end{bmatrix}$$

it is a linear system
g(t, y) = Ay + f(t) where

$$A = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}, \quad \mathbf{f}(t) = \begin{bmatrix} 0 \\ t \end{bmatrix}$$



### Euler numerical methods: forward and backward

- ▶  $t_0 < t_1 < t_2 < \cdots < t_L$  a sequence of times
- steps  $h = t_{\ell} t_{\ell-1}$ ; assume equal for convenience
- $\mathbf{Y}_{\ell} \approx \mathbf{y}(t_{\ell})$ , and  $\mathbf{Y}_{0} = \mathbf{y}_{0}$
- explicit forward Euler method

$$\frac{\mathbf{Y}_{\ell}-\mathbf{Y}_{\ell-1}}{h}=\mathbf{g}(t_{\ell-1},\mathbf{Y}_{\ell-1})$$

o as update formula:

$$\mathbf{Y}_{\ell} = \mathbf{Y}_{\ell-1} + h\mathbf{g}(t_{\ell-1}, \mathbf{Y}_{\ell-1})$$

implicit backward Euler method is just as easy to state

$$\frac{\mathbf{Y}_{\ell} - \mathbf{Y}_{\ell-1}}{h} = \mathbf{g}(t_{\ell}, \mathbf{Y}_{\ell})$$

but requires solving system of eqns at each time step

# Runge-Kutta methods

Runge-Kutta methods with s stages have form

$$\hat{\mathbf{Y}}_i = \mathbf{Y}_{\ell-1} + h \sum_{j=1}^s a_{ij} \, \mathbf{g}(t_{\ell-1} + c_j h, \hat{\mathbf{Y}}_j), \qquad 1 \leq i \leq s$$
 $\mathbf{Y}_\ell = \mathbf{Y}_{\ell-1} + h \sum_{i=1}^s b_i \, \mathbf{g}(t_{\ell-1} + c_i h, \hat{\mathbf{Y}}_i)$ 

some "tableau" for particular explicit instances:



$\begin{array}{c} 0\\ \frac{1}{2}\\ \frac{1}{2} \end{array}$	$\frac{1}{2}$	1			
1	0	0	1		
	$\frac{1}{6}$	$\frac{1}{3}$	$\frac{1}{3}$	6	



Table 6.1: Tableau for the explicit trapezoidal rule (RK2a; left), the classical fourth-order Runge-Kutta method (RK4; middle), and an embedded fourstage, third-order scheme (RK3bs, the PETSc default; right).

### PETSc TS basic setup

```
TSCreate(COMM, &ts);
TSSetProblemType(ts,TS_NONLINEAR);
TSSetRHSFunction(ts,NULL,FormRHSFunction,NULL);
TSSetType(ts,TSRK);
TSSetInitialTimeStep(ts,t0,dt);
TSSetDuration(ts,maxsteps,tf-t0);
TSSetFromOptions(ts);
TSSolve(ts,y);
```

- setup similar to SNES, KSP, etc.
- ▶ setting TS\_NONLINEAR says ODE is in form  $\mathbf{y}' = \mathbf{g}(t, \mathbf{y})$
- calling TSSetRHSFunction() sets a call-back to our code FormRHSFunction(), which evaluates g(t, y)
- ► TSSetFromOptions() allows overrides:
  - the TSSetType() choice by -ts\_type
  - o t0,dt,maxsteps,tf by ...; see -help |grep ts\_
- y must be an allocated Vec of size N containing initial value Y<sub>0</sub>

### first PETSC TS example

recall our example:

$$\mathbf{g}(t,\mathbf{y}) = \begin{bmatrix} y_1 \\ -y_0 + t \end{bmatrix}$$

▶ thus implementation of  $\mathbf{g}(t, \mathbf{y})$ :

### running c/ch6/ode.c

- ode.c is essentially just the stuff on the last two slides ...simple!
- ▶ default run with adaptive, explicit RK  $O(h^3)$  method:

```
$ make
$ ./ode -ts_monitor
0 TS dt 0.1 time 0.
1 TS dt 0.170141 time 0.1
...
87 TS dt 0.11548 time 19.8845
88 TS dt 0.205616 time 20.
error at tf = 20.000 : |y-y_exact|_inf = 0.00930352
```

#### variations:

- changing solver and time axis:
  - | \$ ./ode -ts\_type euler -ts\_final\_time 1.0 -ts\_dt 0.5
- generating run-time movie of trajectory:

```
$ ./ode -ts_monitor_lg_solution -draw_pause 0.1
```

### implicit solvers need SNES and Jacobian

backward Euler:

```
$ ./ode -ts_type beuler # error
$ ./ode -ts_type beuler -snes_fd # works
```

odejac.c adds Jacobian function to ode.c:

```
$ make odejac
$ ./odejac -ts_type beuler # works
```

 a larger class of implicit methods are "θ"-methods with 0 < θ ≤ 1:</li>

$$\frac{\mathbf{Y}_{\ell} - \mathbf{Y}_{\ell-1}}{h} = (1 - \theta) \mathbf{g}(t_{\ell-1}, \mathbf{Y}_{\ell-1}) + \theta \mathbf{g}(t_{\ell}, \mathbf{Y}_{\ell})$$

•  $\theta = 1/2$  case is  $O(h^2)$  accurate: "trapezoid" or "Crank-Nicolson"

```
$ ./odejac -ts_type theta -ts_theta_theta 0.4
$ ./odejac -ts_type cn
```

### "stiff" ODE systems

- consider ODE systems of the form y' = g(y), i.e. "autonomous" without t-dependence
- ▶ ODE system is asymptotically stable if eigenvalues  $\lambda$  of Jacobian  $J = \frac{\partial g_i}{\partial y_j}$  have negative real part: Re  $\lambda < 0$
- such a system is stiff if

$$min(Re \lambda) \ll max(Re \lambda) < 0$$

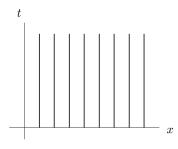
- or:  $\kappa(J) \gg 1$  in symmetric case
- ideas:
  - "stiff" implies that the min(Re  $\lambda$ ) eigenvalues will control time steps of explicit ODE methods, causing very short time steps, even though it is the max(Re  $\lambda$ ) eigenvalues which control the macroscopic behavior of solution
  - o implicit methods will follow the max(Re  $\lambda$ ) part as long as the min(Re  $\lambda$ ) part stays absolutely small

### "method of lines" (MOL) for PDEs

- ▶ time-dependent PDEs are "ODEs in ∞ dimensions"
  - o discretize with this in mind
  - this is how to solve time-dependent PDEs using PETSC
- consider time-dependent PDEs of form

$$u_t = G(t, u, \nabla u, \nabla^2 u)$$

idea: spatial discretization only gives system of ODEs



### MOL for heat equation

• example: if  $\{x_i\}$  is an N-point spatial grid then

$$u_t = u_{xx}$$
  $\rightarrow$   $U'_j(t) = \frac{U_{j-1}(t) - 2U_j(t) + U_{j+1}(t)}{\Delta x^2}$ 

gives system in form " $\mathbf{y}' = \mathbf{g}(t, \mathbf{y})$ " in  $\mathbb{R}^N$ , where  $U_j(t) \approx u(t, x_j)$  and  $\mathbf{U}(t) = \{U_j(t)\} \in \mathbb{R}^N$ 

symmetric matrix for RHS is

$$A = \frac{1}{\Delta x^2} \begin{pmatrix} -2 & 1 & 0 & & 0 \\ 1 & -2 & 1 & & 0 \\ 0 & 1 & -2 & & 0 \\ & & & \ddots & \\ 0 & 0 & 0 & & -2 \end{pmatrix}$$

has negative real eigenvalues, some  $\approx$  0 and some  $\lambda \to -\infty$ 

this is generic:

MOL on diffusion equations gives stiff ODE systems

### TS example codes in c/ch6/

• using form  $\mathbf{y}' = \mathbf{g}(t, \mathbf{y})$ :

ode.c solves above example

odejac.c adds Jacobian so that implicit methods can be used heat.c MOL for classical heat equation

$$u_t = D\nabla^2 u + f$$

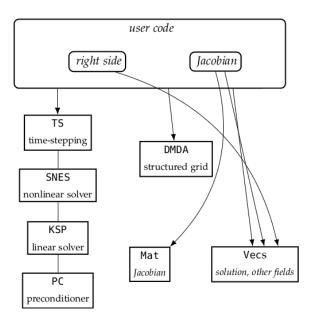
▶ using "implicit-explicit" form  $\mathbf{f}(t, \mathbf{y}, \mathbf{y}') = \mathbf{g}(t, \mathbf{y})$ :

pattern.c MOL for system of two coupled nonlinear heat-like PDEs

$$u_t - D_u \nabla^2 u = -uv^2 + F(1-u)$$
  
 $v_t - D_v \nabla^2 v = +uv^2 - (F+k)v$ 

- a model for two chemical species with reaction, diffusion, and feed (Pearson, 1993)
- "adaptive Runge-Kutta implicit-explicit" appropriate (-ts\_type arkimex)

### structure of heat.c



### running heat.c

- lacktriangle c/ch6/heat.c solves  $u_t = D 
  abla^2 u + f(x,y)$ 
  - on unit square  $\Omega = (0,1) \times (0,1)$
  - non-homogeneous Neumann boundary conditions in x
  - periodic boundary conditions in y
  - energy is conserved
- high-res Crank-Nicolson run on 4 processes with saving of time axis and solution:

```
mpiexec -n 4 ./heat -ts_type cn -da_refine 7 \
    -ts_dt 0.0001 -ts_final_time 0.05 \
    -ts_monitor binary:theat.dat \
    -ts_monitor_solution binary:uheat.dat
```

output can be transformed into an off-line movie:

```
./plotTS.py -mx 385 -my 384 theat.dat uheat.dat \ -oroot heat ffmpeg -r 8 -i heat%03d.png heat.m4v
```

### running pattern.c

c/ch6/pattern.c solves

$$u_t - D_u \nabla^2 u = -uv^2 + F(1-u)$$
  
 $v_t - D_v \nabla^2 v = +uv^2 - (F+k)v$ 

- periodic boundary conditions in both directions
- o creates 2D DMDA with dof=2
- ▶ problem in form  $\mathbf{f}(t, \mathbf{y}, \mathbf{y}') = \mathbf{g}(t, \mathbf{y})$ :
  - -ts type arkimex is default
  - FormIFunctionLocal() for f; with Jacobian
  - FormRHSFunctionLocal() for g; no Jacobian
- ▶ high-res run on 4 procs, and off-line movie of *u* component:

```
mpiexec -n 4 ./pattern -ts_adapt_type none \
    -ts_final_time 5000 -ts_dt 10 -da_refine 8 \
    -ts_monitor binary:t8.dat \
    -ts_monitor_solution binary:uv8.dat
./plotTS.py -mx 768 -my 768 -dof 2 -c 0 \
    t8.dat uv8.dat -oroot uhigh8
ffmpeg -r 8 -i uhigh8%03d.png uhigh8.m4v
```