



Fast solution of stiff PDEs in 1D, 2D and 3D

Industrial and Applied Mathematics Seminar, University of Nottingham

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February 24, 2016



UNIVERSITY OF
OXFORD

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- **Problem:** Stiff PDEs of the form

$$u_t(t, X) = \mathcal{L}u + \mathcal{N}(u), \quad t \in [0, T], \quad X \in [0, 2\pi]^d \quad (d = 1, 2, 3),$$

with initial condition $u(0, X)$

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Higher-order terms of the equation are linear: exponential integrators (Section 3)





- A periodic (or rapidly decaying) function $u(x)$ on $[0, 2\pi]$ is represented by its trigonometric interpolant on a grid of N points $x_j = 2\pi j/N$, $0 \leq j \leq N-1$,

$$u(x) \approx u_N(x) = \sum_{k=-N/2}^{N/2} u_k e^{ikx}, \quad u_k = \frac{1}{N} \sum_{j=0}^{N-1} u(x_j) e^{-ikx_j} = \text{fft}(u(x_j))$$



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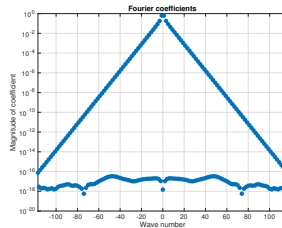
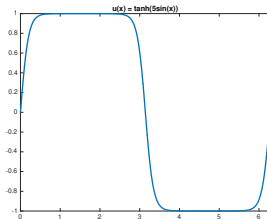
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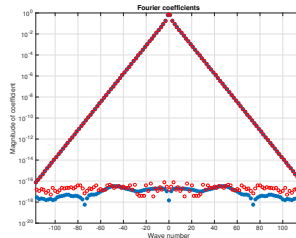
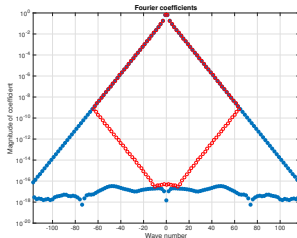
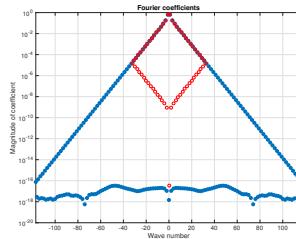
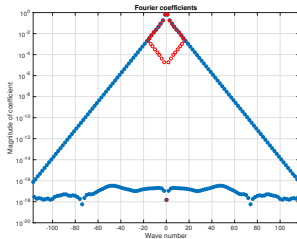
- Example: $u(x) = \tanh(5 \sin(x))$, $N = 235$
`u = chebfun(@(x) tanh(5*sin(x)), [0 2*pi], 'trig');`





Trigonometric interpolation (2/3)

- Constructor: evaluates the function on grids of sizes 17, 33, 65, 129, 257, ..., takes the fft and looks at the amplitude of the Fourier coefficients

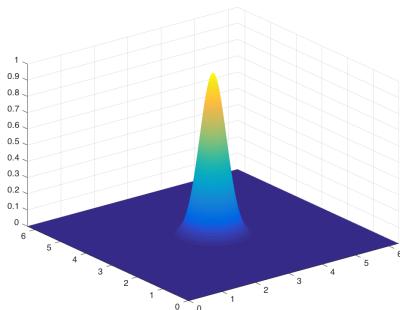






- Example in 2D: $u(x, y) = \exp(-5((x - \pi)^2 + (y - \pi)^2))$

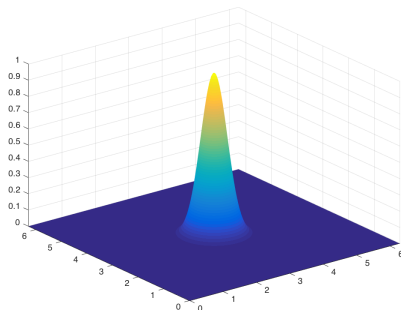
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u = chebfun2(@(x,y)exp(-5*((x-pi).^2+(y-pi).^2)), [0 2*pi 0 2*pi], 'trig');
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- In 3D, one can type, e.g.,

```
dom = [0 2*pi 0 2*pi 0 2*pi];  
u = chebfun3(@(x,y,z)exp(-5*((x-pi).^2+(y-pi).^2+(z-pi).^2)), dom, 'trig');
```






- For discretizing

$$u_t(t, X) = \mathcal{L}u + \mathcal{N}(u), \quad u(0, X) = u_0(X),$$

we use a Fourier spectral method in coefficient space, i.e., we look for a solution of the form (with $N = N_x N_y N_z$ points)

$$u(t, X) \approx \sum_{k_x = -\frac{N_x}{2}}^{\frac{N_x}{2}}{}' \sum_{k_y = -\frac{N_y}{2}}^{\frac{N_y}{2}}{}' \sum_{k_z = -\frac{N_z}{2}}^{\frac{N_z}{2}}{}' u_{k_x k_y k_z}(t) e^{i(k_x x + k_y y + k_z z)}$$



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- The linear part \mathcal{L} is discretized with Fourier differentiation matrices \Rightarrow matrix \mathbf{L}
- This leads to a system of N ODEs for the coefficients $u(t) = \{u_{k_x k_y k_z}(t)\}$,

$$u'(t) = \mathbf{L}u + \mathbf{N}(u), \quad u(0) = u_0$$





- First three Fourier differentiation matrices (even N):

$$D = \begin{pmatrix} 0 \\ (-\frac{N}{2} + 1)i \\ \vdots \\ -2i \\ -i \\ 0 \\ i \\ 2i \\ \vdots \\ (\frac{N}{2} - 1)i \end{pmatrix}, \quad D^{(2)} = \begin{pmatrix} -\frac{N^2}{4} \\ \vdots \\ -4 \\ -1 \\ 0 \\ -1 \\ -4 \\ \vdots \\ -(\frac{N}{2} - 1)^2 \end{pmatrix}, \quad D^{(3)} = \begin{pmatrix} 0 \\ -(-\frac{N}{2} + 1)^3 i \\ \vdots \\ 8i \\ i \\ 0 \\ -i \\ -8i \\ \vdots \\ -(\frac{N}{2} - 1)^3 i \end{pmatrix}$$



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- **Method:** Time-stepping with exponential integrators, e.g.,

$$u_{n+1} = e^{h\mathbf{L}}u_n + h\varphi_1(h\mathbf{L})\mathbf{N}(u_n), \quad h = t_{n+1} - t_n, \quad \varphi_1(z) = \frac{e^z - 1}{z}$$



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- **How to derive this scheme?** Consider the linearized version of the system of ODEs on $[t_n, t_{n+1}]$,

$$u'(t) = \mathbf{S}(u_n) + (u - u_n)\mathbf{S}_u(u_n), \quad u(t_n) = u_n$$

with exact solution at t_{n+1}

$$u_{n+1} = u_n + h\varphi_1(h\mathbf{S}_u(u_n))\mathbf{S}(u_n),$$

and approximate $\mathbf{S}_u(u_n)$ by \mathbf{L}





For given starting values u_0, u_1, \dots, u_{q-1} at times $t = 0, h, \dots, (q-1)h$, the numerical approximation u_{n+1} at time $t_{n+1} = (n+1)h$, $n+1 \geq q$, is given by

$$u_{n+1} = e^{hL}u_n + h \sum_{i=1}^s B_i(hL)\mathbf{N}(v_i) + h \sum_{i=1}^{q-1} V_i(hL)\mathbf{N}(u_{n-i}),$$

with *steps* u_{n-i} and *stages* $v_1 = u_n$ and, for $2 \leq i \leq s$,

$$v_i = e^{C_i hL}u_n + h \sum_{j=1}^{i-1} A_{i,j}(hL)\mathbf{N}(v_j) + h \sum_{j=1}^{q-1} U_{i,j}(hL)\mathbf{N}(u_{n-j})$$



Method	Type	Order	Stages s	Steps q
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ABNørsett5	ETD Adams-Bashforth	5	1	5
ABNørsett6	ETD Adams-Bashforth	6	1	6
Friedli (VRK4)	ETD Runge-Kutta	4	4	1
Strehmel-Weiner	ETD Runge-Kutta	4	4	1
Cox-Matthews (ETDRK4)	ETD Runge-Kutta	4	4	1
Krogstad (ETDRK4-B)	ETD Runge-Kutta	4	4	1
Minchev	ETD Runge-Kutta	4	4	1
Hochbruck-Ostermann	ETD Runge-Kutta	4	5	1
Luan-Ostermann (EXPRK5S8)	ETD Runge-Kutta	5	8	1
(Mod)GenLawson41	(Mod.) Gen. Lawson	4	4	1
(Mod)GenLawson42	(Mod.) Gen. Lawson	4	4	2
(Mod)GenLawson43	(Mod.) Gen. Lawson	4	4	3
(Mod)GenLawson44	(Mod.) Gen. Lawson	5	4	4
(Mod)GenLawson45	(Mod.) Gen. Lawson	6	4	5
PEC423	Predictor-corrector	4	2	3
PECEC433	Predictor-corrector	4	3	3
PEC524	Predictor-corrector	5	2	4
PECEC534	Predictor-corrector	5	3	4
PEC625	Predictor-corrector	6	2	5
PECEC635	Predictor-corrector	6	3	5
PEC726	Predictor-corrector	7	2	6
PECEC736	Predictor-corrector	7	3	6



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- ETD Runge-Kutta of order four have similar accuracy and stability properties, **EXPRK5S8** the most accurate but sometimes unstable



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- **Predictor-Corrector methods of order ≥ 5** more accurate but often unstable



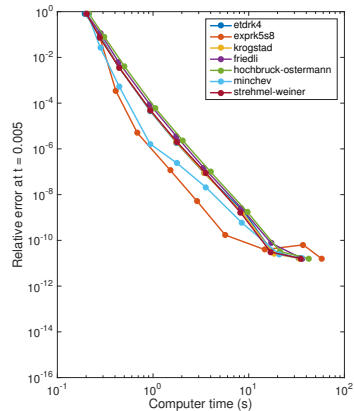
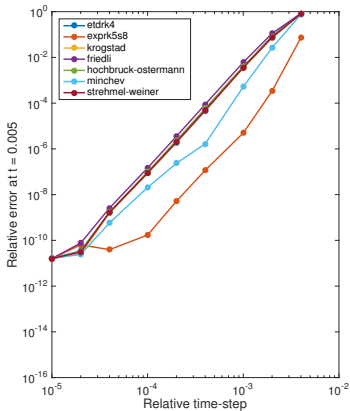
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We plot relative error = $\frac{\|u_{\text{approx}} - u_{\text{exact}}\|_{\infty}}{\|u_{\text{exact}}\|_{\infty}}$ vs relative time-step = $\frac{dt}{0.005}$ and time



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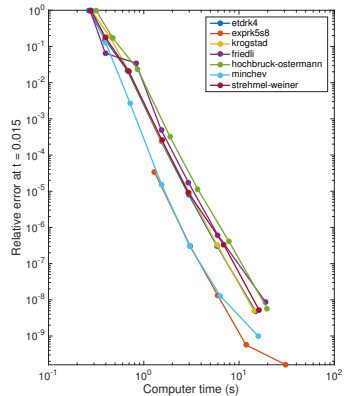
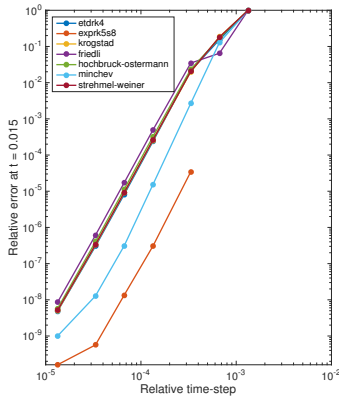




Example 2: ETD RK schemes for KdV equation from $t = 0$ to $t = 0.015$



Example 2: ETD RK schemes for KdV equation from $t = 0$ to $t = 0.015$

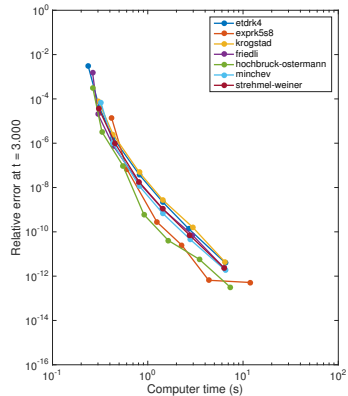
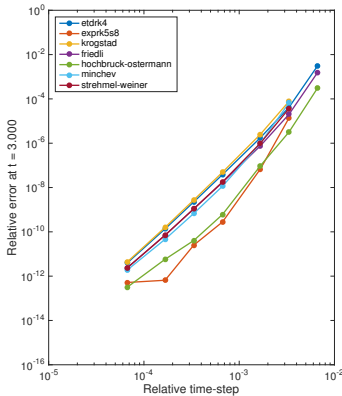




Example 3: ETD RK schemes for Cahn-Hilliard equation from $t = 0$ to $t = 3$



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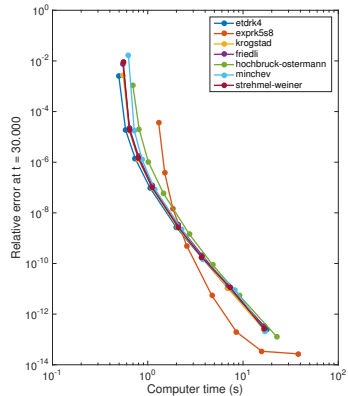
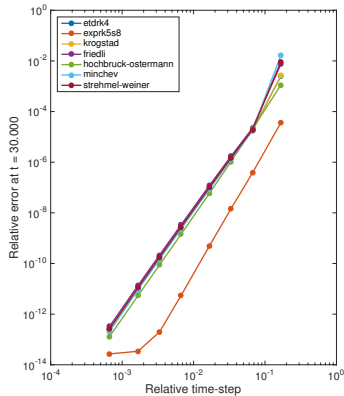




Example 4: ETD RK schemes for Gray-Scott equations in 2D from $t = 0$ to $t = 30$



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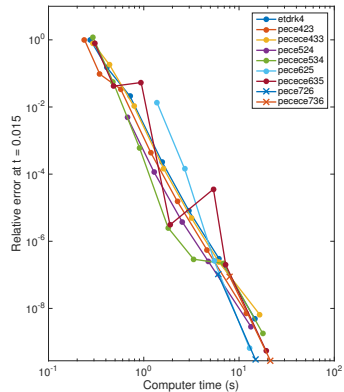
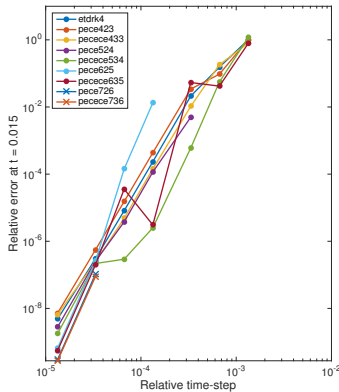




Example 5: PEC schemes for KdV equation from $t = 0$ to $t = 0.015$



Example 5: PEC schemes for KdV equation from $t = 0$ to $t = 0.015$

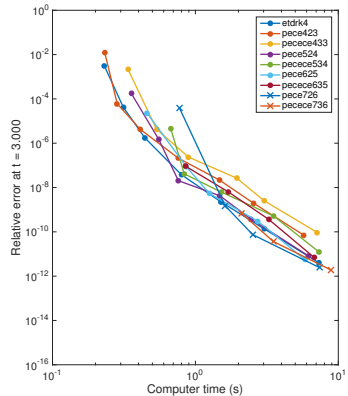
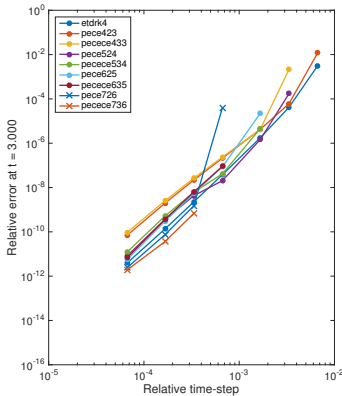




Example 6: PEC schemes for Cahn-Hilliard equation from $t = 0$ to $t = 3$



Example 6: PEC schemes for Cahn-Hilliard equation from $t = 0$ to $t = 3$

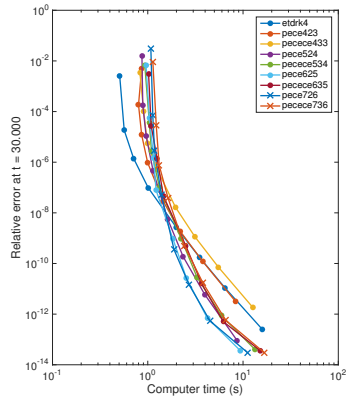
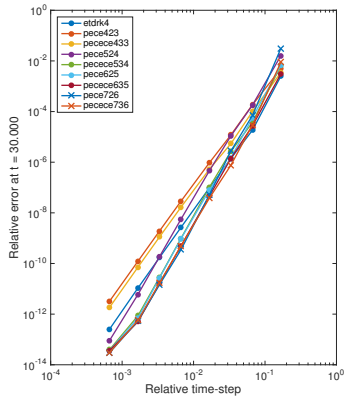




Example 7: PEC schemes for Gray-Scott equations in 2D from $t = 0$ to $t = 30$



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- Future work includes PDEs on the sphere (with `spherefun`)