What SAGE Needs to be Useful for Applied Mathematics

(in my corner of this world, at least)

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or....

Can SAGE be

Software for Applied mathematics, Graphics, and Engineering?

- Teaching replace Matlab, Maple, etc.?
- Research e.g., Numerical analysis
- Scientific computing, data manipulation, visualization
- chebfun filling the gap between symbolic and numerical computing?

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- Students can use after leaving university,
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 - Also supports latex, webpages, etc.

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- Python scripting is popular for scientific computing
 - Manipulating large data sets in various formats,
 - Coupling together diverse software in different languages,
 - Creating modern interfaces and GUIs for old codes,
 - Organizing benchmark tests, validations, parameter studies, etc.

Many packages and modules already available for

. . .

- Numerical methods: NumPy, SciPy, MatPy, Pysparse, Signaltools, ...
- Graphics and visualization: Gnuplot, PythonPlot, MayaVi, gracePlot.py, NURBS, ...
- Interfacing with other languages: f2py, swig, pymat,

Reference: Hans Petter Langtangen, *Python Scripting for Computational Science*, Springer, 2004.

Sparse matrix operations essential for many applications.

Example: Solve u''(x) = f(x) on $0 \le x \le 1$ with boundary conditions u(0) = u(1) = 0.

Let
$$U_j \approx u(x_j)$$
, $j = 1, 2, ..., m$
where $x_j = jh$ with $h = 1/(m+1)$.

Replace ODE by finite difference equations

$$\frac{1}{h^2}(U_{j-1} - 2U_j + U_{j+1}) = f(x_j), \quad j = 1, 2, \dots, m.$$

This is a tridiagonal linear system of m equations.

The tridiagonal system is solved in O(m) operations, not $O(m^3)$ as needed for a dense $m \times m$ matrix.

2D Poisson problem

Consider $u_{xx}(x, y) + u_{yy}(x, y) = f(x, y)$ on unit square with u = 0 on boundaries.

Finite difference method with "5-point stencil":

$$\begin{split} &\frac{1}{h^2} \left[(U_{i-1,j} - 2U_{ij} + U_{i+1,j}) \right. \\ &+ \left(U_{i,j-1} - 2U_{ij} + U_{i,j+1} \right) \right], \quad i, \ j = 1:m. \end{split}$$

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Natural row-wise ordering of unknowns: (not the best for Gaussian elimination!)

$$U = \begin{bmatrix} U^{[1]} \\ U^{[2]} \\ \vdots \\ U^{[m]} \end{bmatrix}, \quad \text{where } U^{[j]} = \begin{bmatrix} U_{1j} \\ U_{2j} \\ \vdots \\ U_{mj} \end{bmatrix}.$$

Gives sparse $m^2 \times m^2$ matrix with bandwidth m.

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Gives sparse $m^2 \times m^2$ matrix with bandwidth m (block $m \times m$):

$$A = \frac{1}{h^2} \begin{bmatrix} T & I & & & \\ I & T & I & & \\ & I & T & I & \\ & & \ddots & \ddots & \ddots \\ & & & I & T \end{bmatrix}, \qquad U = \begin{bmatrix} U^{[1]} \\ U^{[2]} \\ U^{[3]} \\ \vdots \\ U^{[m]} \end{bmatrix},$$

Each block T or I is itself an $m \times m$ matrix,

$$T = \begin{bmatrix} -4 & 1 & & \\ 1 & -4 & 1 & & \\ & 1 & -4 & 1 & \\ & & \ddots & \ddots & \ddots \\ & & & 1 & -4 \end{bmatrix}$$

x = h * (1:m)'; y = x; [X,Y] = meshgrid(x,y);

```
I = eye(m); e = ones(m,1);
T = spdiags([e -2*e e], [-1 0 1], m, m);
A = 1/h^2 * (kron(I,T) + kron(T,I));
```

```
b = f(X,Y);
bvec = reshape(b,m<sup>2</sup>,1);
uvec = A\bvec;
u = reshape(uvec,m,m);
```

Matlab uses Gaussian elimination with smart ordering.

For Poisson problem, better approach is FFT.

For general sparse systems, iterative methods often used.

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• Parallel computing: Crucial for many large problems, even as processors get faster (especially now since most are multi-core).

Grid computing, GPU computing?

Need symbolic manipulation of Taylor series in several variables for general functions, e.g. in Maple:

Matlab is so popular largely because it combines

- simple programming capability,
- interfaces to high quality software,
- data manipulations tools,
- and simple and powerful graphics.

Numerical results often consist of approximations to functions at millions of grid points — graphics is the only way to view and interpret.

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- Lots of graphics packages exist don't want to reinvent all this! Need basic graphics well integrated and perhaps front end to more sophisticated packages.

Sumatra event of December 26, 2004

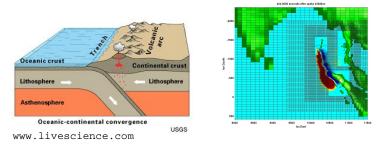
Magnitude 9.1 quake near Sumatra, where Indian tectonic plate is being subducted under the Burma platelet.

Rupture along subduction zone

pprox 1200 km long, 150 km wide

Propagating at \approx 2 km/sec (for \approx 10 minutes)

Fault slip up to 15 m, uplift of several meters. (Fault model from Caltech Seismolab.)



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Adaptive mesh refinement is essential

Zoom on Madras harbors with 4 levels of refinement:

- Level 1: 1 degree resolution ($\Delta x \approx 60$ nautical miles)
- Level 2 refined by 8.
- Level 3 refined by 8: $\Delta x \approx 1$ nautical mile (only near coast)
- Level 4 refined by 64: $\Delta x \approx 25$ meters (only near Madras)

Factor 4096 refinement in x and y.

Less refinement needed in time since $c \approx \sqrt{gh}$.

Runs in a few hours on a laptop.

For animation and other related results, please visit http://www.amath.washington.edu/~rjl/talks/hilo06/ On-going research problem led by Nick Trefethen (Oxford), with Zachary Battles, Ricardo Pachón,

Toby Driscoll (Delaware).

References:

Z. Battles and L. N. Trefethen, *An extension of Matlab to continuous functions and operators*, SIAM J. Sci. Comp. 25 (2004), pp. 1743–1770.

http://web.comlab.ox.ac.uk/projects/chebfun/

Example:
$$f(x) = \exp(-x^2)\sqrt{x+2}$$
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Suppose we need to work with $g(x) = \int_{-1}^{x} f(y) \, dy$.

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f.nintegrate(x, -1, 0)

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What if we need to ...

• compute
$$||g|| = \left(\int_{-1}^{1} |g(x)|^2 dx\right)^{1/2}$$
?

• compute
$$(g, f) = \int_{-1}^{1} g(x) f(x) dx$$
 ?

• work with $h(x) = \int_{-1}^{x} g(z) dz$?

```
>> x = chebfun('x');
>> f = exp(x.^2) .* sqrt(2+x);
>> q = cumsum(f);
>> f2 = diff(g);
>> norm(f-f2)
ans =
     2.760689672717827e-14
>> h = cumsum(q);
>> q'*f
ans =
     8.316267551154510e+00
>> disp([size(f) size(g) size(h)]
   -21 1 -22 1 -23 1
```

Approximation by polynomials

Weierstrass approximation theorem: If $f \in C[-1, 1]$ and p_n^* is the best approximation to f(x) by a polynomial of degree n, then

$$\|p_n^* - f\|_{\infty} = \max_{-1 \le x \le 1} |p_n^*(x) - f(x)| \to 0 \text{ as } n \to \infty.$$

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Interpolating polynomial: Given any n + 1 points there exists a unique polynomial $P_n(x)$ of degree $\leq n$ satisfying

$$P_n(x_j) = f(x_j), \quad j = 0, 1, \dots, n.$$

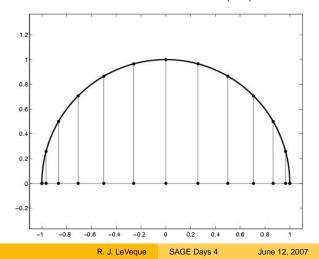
Many ways to compute, barycentric interpolation is best.

Choice of interpolating points x_0, \ldots, x_n .

Bad choice: Equally spaced, $x_j = -1 + jh$ with h = 2/n. Runge phenomenon, $||P_n - f||_{\infty}$ may blow up as $n \to \infty$

Polynomial Interpolation

Choice of interpolating points x_0, \ldots, x_n . Good choice: Chebyshev points $x_j = \cos\left(\frac{\pi j}{n}\right)$



Polynomial Interpolation at Chebyshev points

Suppose f(x) is analytic in ellipse enclosing [-1, 1] with major and minor axes of length L and ℓ .

Then

$$\max_{-1 \le x \le 1} |P_n(x) - f(x)| \le CK^{-n}$$

with $K = L + \ell$.

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Spectral accuracy: error goes to zero faster than n^{-p} for all p. (and some finite p depending on smoothness of f if it's not analytic near interval).

On-going research including:

- Backward error analysis,
- Extension to piecewise continuous functions,
- Continuous analogues of Householder and LU factorizations,
- Global optimization,
- Extension to 2D and 3D,
- · Krylov space iterative methods for operators,
- Spectral methods for PDEs

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