

# Highlights of recent work:

## 1. Effects of shaping on the Dimits nonlinear shift

with Emily Belli

## 2. Implementation of equilibrium sheared flows in GS2

with Bill Dorland & Nuno Loureiro

## 3. Algorithm studies

(improved GMRES, Krylov solvers, TVD/CWENO shock-capturing algorithms, semi-implicit methods)

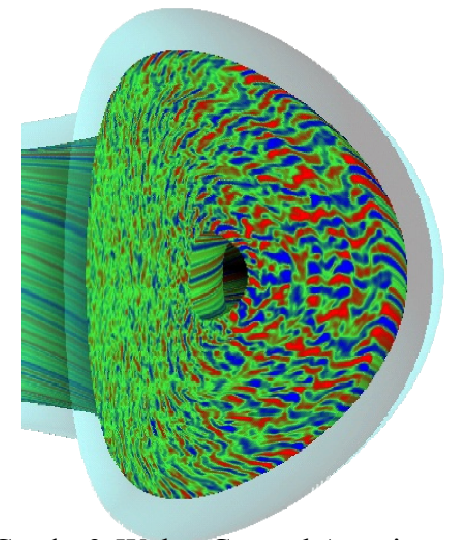
(interactions with Jin Chen, Steve Jardin, Ravi Samtaney, David Keyes, Mark Adams, Jim Stone, Prateek Sharma, Nuno Loureiro, Eitan Tadmor)

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Greg Hammett

PPPL theory microseminar 11/16/2006

(I only showed about half these slides for a ~10 min talk.)

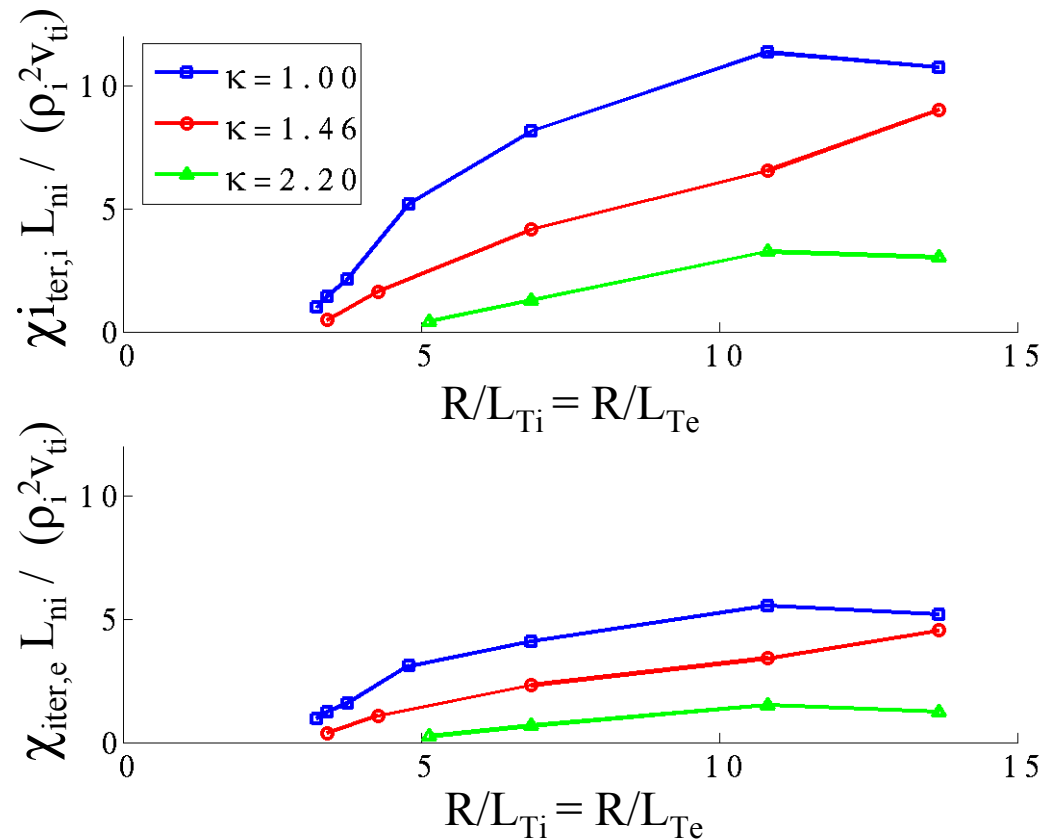


Candy & Waltz, General Atomics

# Unlike the linear results, the nonlinear critical temperature gradient increases with strong shaping.

Nonlinear ion and electron heat fluxes vs.  $R/L_T$

(Electrostatic)

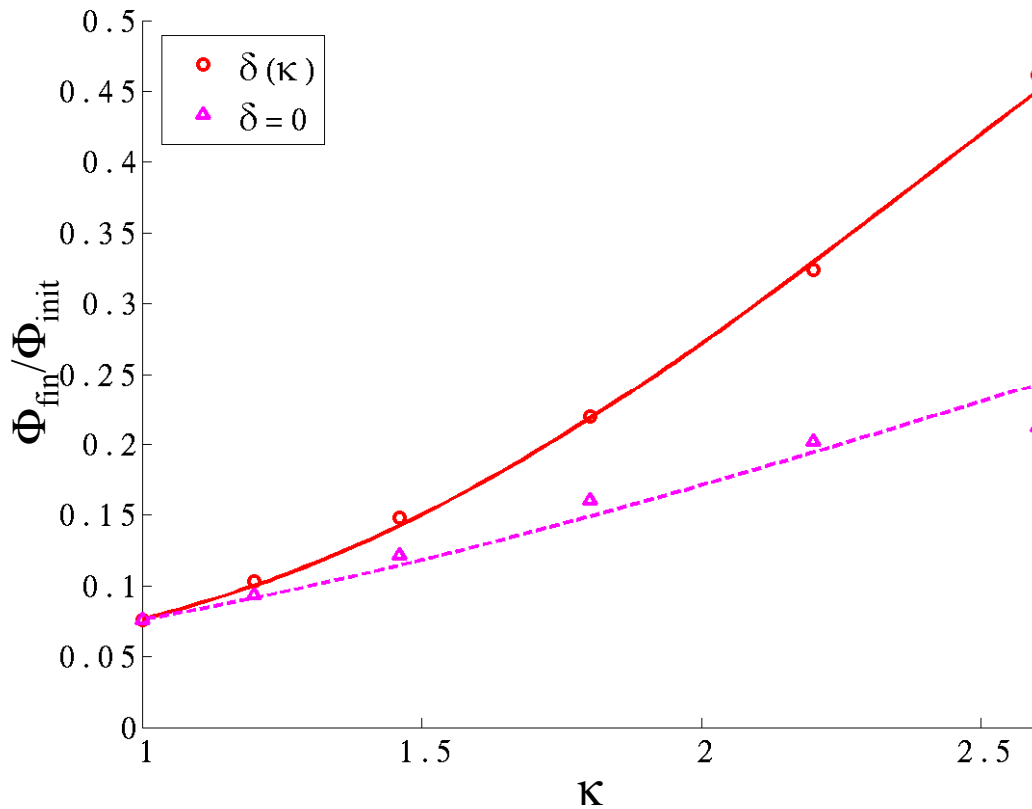


( $\partial_\rho \beta$  is varied with  $\kappa$  &  $R/L_T$ )

# The larger upshift of the nonlinear critical gradient with higher shaping may be due to enhanced zonal flows.

Amplitude of R-H residual zonal flows vs.  $\kappa$

(data points are the GS2 results;  
the lines are the analytic model prediction)



Overall, the geometry of residual zonal flows is important and may help explain why strong shaping is favorable in experiments.

A model prediction for the scaling of the R-H\* h with shaping was found empirically based on the GS2 data. We assume:

$$\frac{\Phi_{fin}}{\Phi_{init}} = \left( 1 + \frac{1.6C}{h_{shaping}} \right)^{-1}$$

where

$$h_{shaping} = \underbrace{\frac{\epsilon^{1/2}}{q^2}}_{\substack{\text{Rosenbluth} \\ \text{Hinton}}} \underbrace{f(\kappa, \delta)}_{\substack{\text{shaping} \\ \text{effects}}}$$

( $C=0.887$  to match GS2 circular case ( $\kappa=1$ ) with original R-H model.)

We found a good fit with:

$$f(\kappa, \delta) = \frac{1}{2} \left[ 1 + \kappa^2 \left( 1 + r \frac{\partial \delta}{\partial r} \right)^2 \right]$$

Analytic confirmation: Yong Xiao & Catto '06

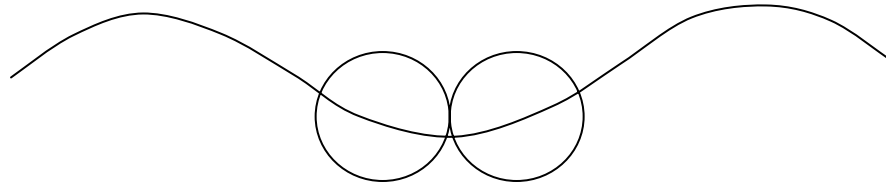
\* M. Rosenbluth & F. Hinton, Phys. Rev. Lett. **80**, 724 (1998).

# Physical picture of Hinton-Rosenbluth residual flow:

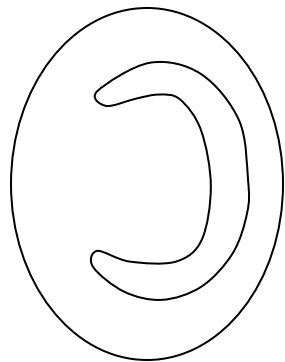
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Classical polarization density:

$$n_{i,pol} = -\int d^3v F_{M0} \frac{e}{T_i} \left( \Phi - \left\langle \left\langle \Phi \right\rangle_{gyroaverage} \right\rangle_{gc\_to\_particle} \right) = -n_{i0} k_{\perp}^2 \rho_i^2 \frac{e\Phi}{T_i}$$



Neoclassical polarization density (enhanced by larger banana widths):



$$n_{i,pol} = -n_{i0} \underbrace{\sqrt{\frac{r}{R}}}_{\text{fraction of trap. particles}} k_{\perp}^2 \rho_b^2 \frac{e\Phi}{T_i}$$

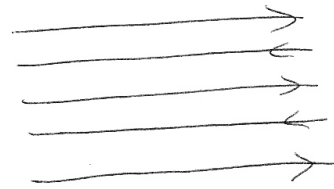
Hinton-Rosenbluth residual flow:

$$\frac{\Phi_{final}}{\Phi_{initial}} = \frac{k_r^2 \rho_i^2}{\sqrt{r/R} k_r^2 \rho_b^2 + k_r^2 \rho_i^2} = \frac{1}{1 + \frac{q^2}{\sqrt{r/R}} \frac{1}{f_{shaping}}}$$

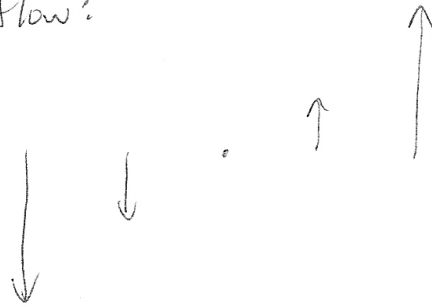
# Implementation of Equilibrium-Scale ExB shear in GS2

Hammett, Dorland, Loureiro

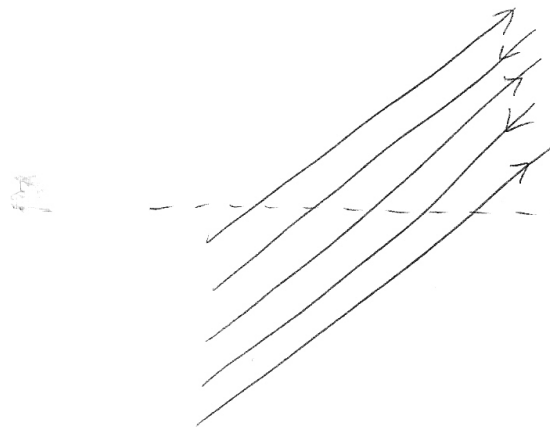
Eddies initially with  $h_x = 0$ :



+ Sheared flow:



At a later time:



Now has finite  
 $h_x = h_{x0} - h_y \frac{\partial v_y}{\partial x} t$

GS2 solves Eq. of form:

$$\frac{\partial}{\partial t} f(k_x, k_y, \theta, E, \mu, t) = L(k_x, k_y, \theta, E, \mu) f(k_x, k_y, \dots, t) + \text{NonlinTerms}$$

Include equilibrium scale ExB shear by substitution:

$$k_x \rightarrow k_x(t) = k_{x0} - k_y \frac{\partial v_y}{\partial x} t$$

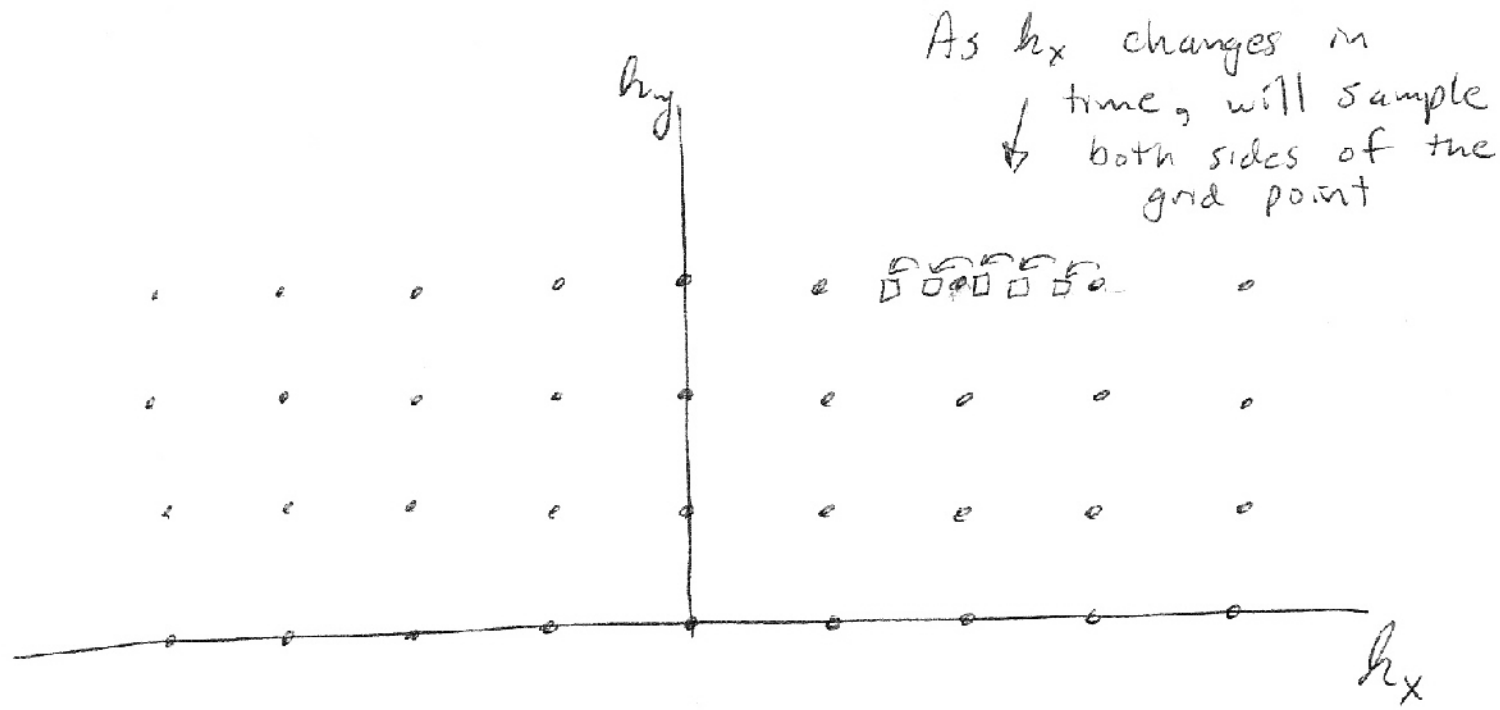
$$\frac{\partial}{\partial t} f(k_x(t), k_y, \theta, \dots, t) = \underbrace{L(k_x(t), k_y, \theta, \dots)}_{\text{implicit}} f(k_x(t), k_y, \dots, t) + \text{NonlinTerms}$$

GS2 treats all linear terms implicitly. Recalculating all of the implicit response matrices every time step when  $k_x$  changes a little bit would be very expensive.

Instead of recalculating response matrices, just use Nearest-Grid-Point interpolations. Most easily implemented in reverse:

Whenever  $k_x(t)$  differs from  $k_{x0}$  by more than  $\Delta k_x/2$ , shift  $f$ :

$$f(k_{xi}, k_y) \Leftarrow f(k_{xi} + \Delta k_x, k_y)$$



Errors will tend to ~~cancel~~ cancel over time, making the method quasi-2<sup>nd</sup> order accurate.

(Like Nearest Grid Point NGP that is widely used in Gyrokinetic PIC codes, or somewhat like Godunov splitting).

(Could generalize to true 2<sup>nd</sup> order accuracy by doing 2 iterations per time step and interpolating.)

With sheared magnetic field:

$$h_x(t) = h_{xi} + h_y \hat{s} \theta - h_y \frac{\partial v_y}{\partial x} t$$

|||  
-h\_y \hat{s} \theta\_0

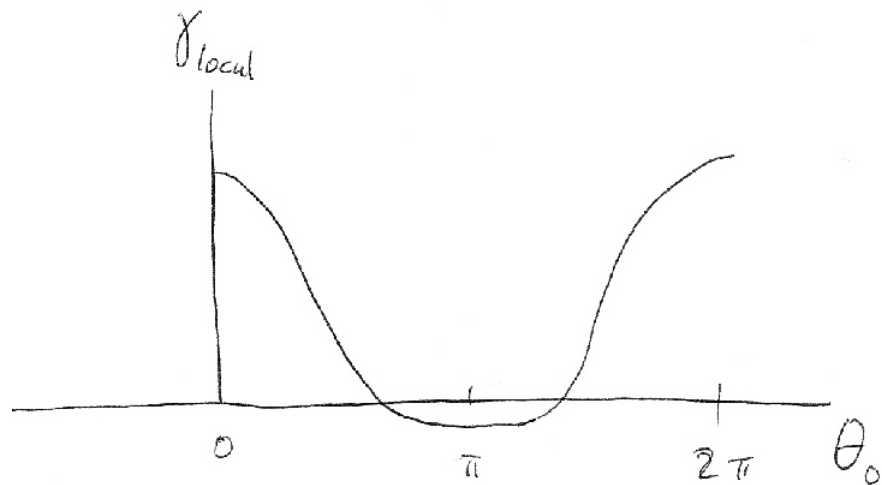
$$h_x(t) = h_y \hat{s} (\theta - \theta_0) - h_y \frac{\partial v_y}{\partial x} t$$



EXB shear twists  
the eddy.

Can undo this twist  
by moving in  $\theta$   
along the sheared  
magnetic field.

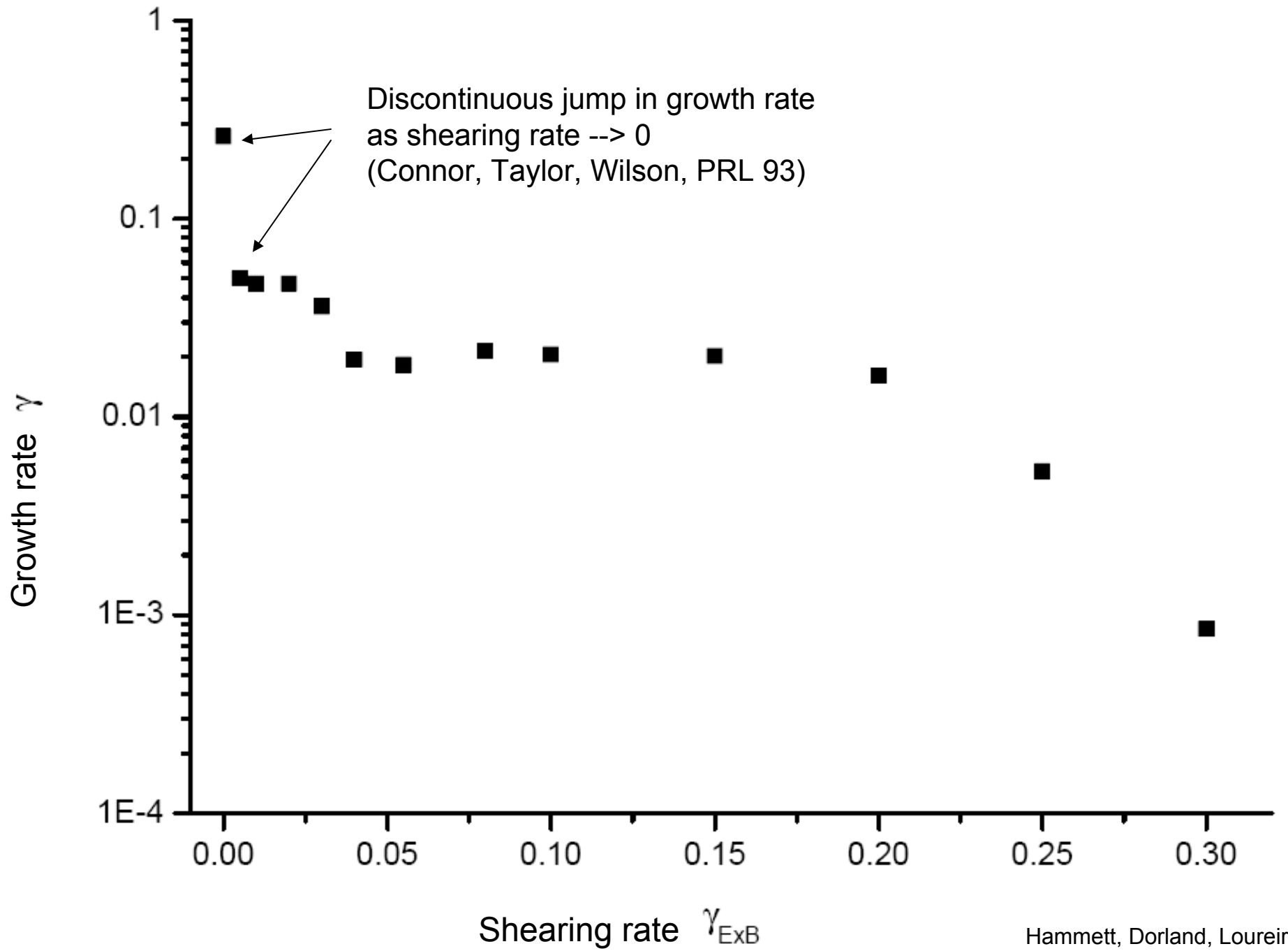


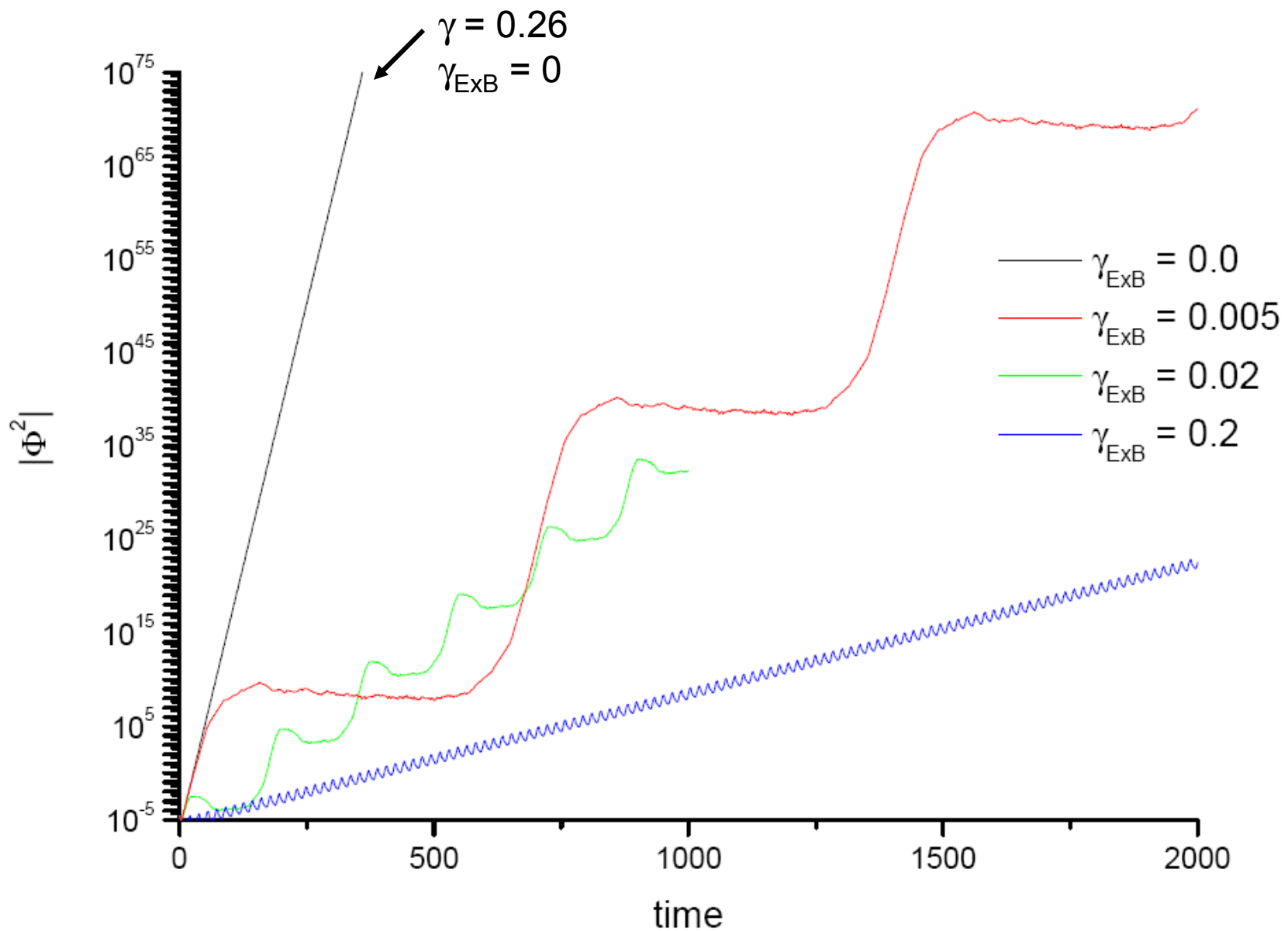


As mode moves along  $\theta_0$ , the time-averaged growth rate changes discontinuously between

$$\gamma_{\text{EXB}} = \frac{\partial V_y}{\partial X} = 0 \quad \text{if } V_{\text{EXB}} \neq 0$$

But not really relevant if mode grows to nonlinear amplitude before moving out of the bad curvature region.



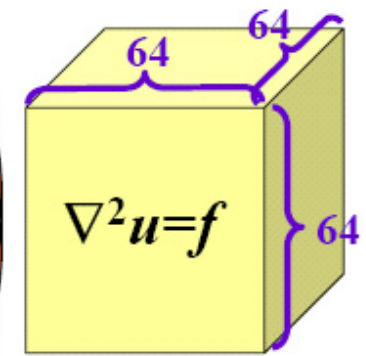


Floquet mode representation (Waltz, Dewar, Garbet 98) shows large growth of wave packet while in bad curvature region, even though time-averaged growth rate is small.

# The power of optimal algorithms

- Advances in algorithmic efficiency can rival advances in hardware architecture
- Consider Poisson's equation on a cube of size  $N=n^3$

<i>Year</i>	<i>Method</i>	<i>Reference</i>	<i>Storage</i>	<i>Flops</i>
1947	GE (banded)	Von Neumann & Goldstine	$n^5$	$n^7$
1950	Optimal SOR	Young	$n^3$	$n^4 \log n$
1971	CG w/ modified ILU	Reid / Gustafson (BIT 78)	$n^3$	$n^{3.5} \log n$
1984	Full MG	Brandt Multigrid	$n^3$	$n^3$



If  $n=64$ , implies  
X16 Million speedup:  
6 months  $\rightarrow$  1 second

Restarted GMRES

$n^5$

Loose GMRES/CG (Baker 2005)

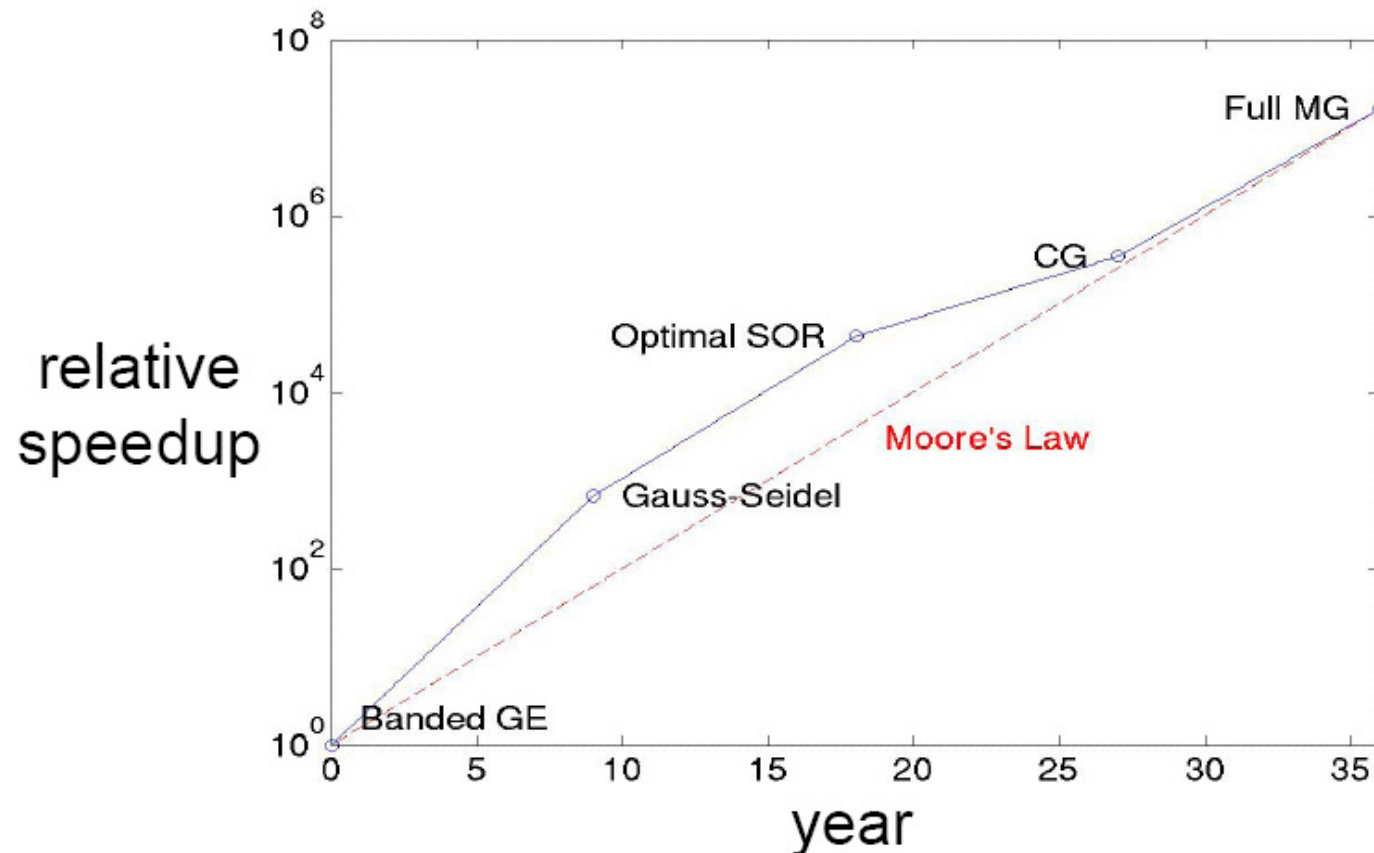
$\sim n^4$

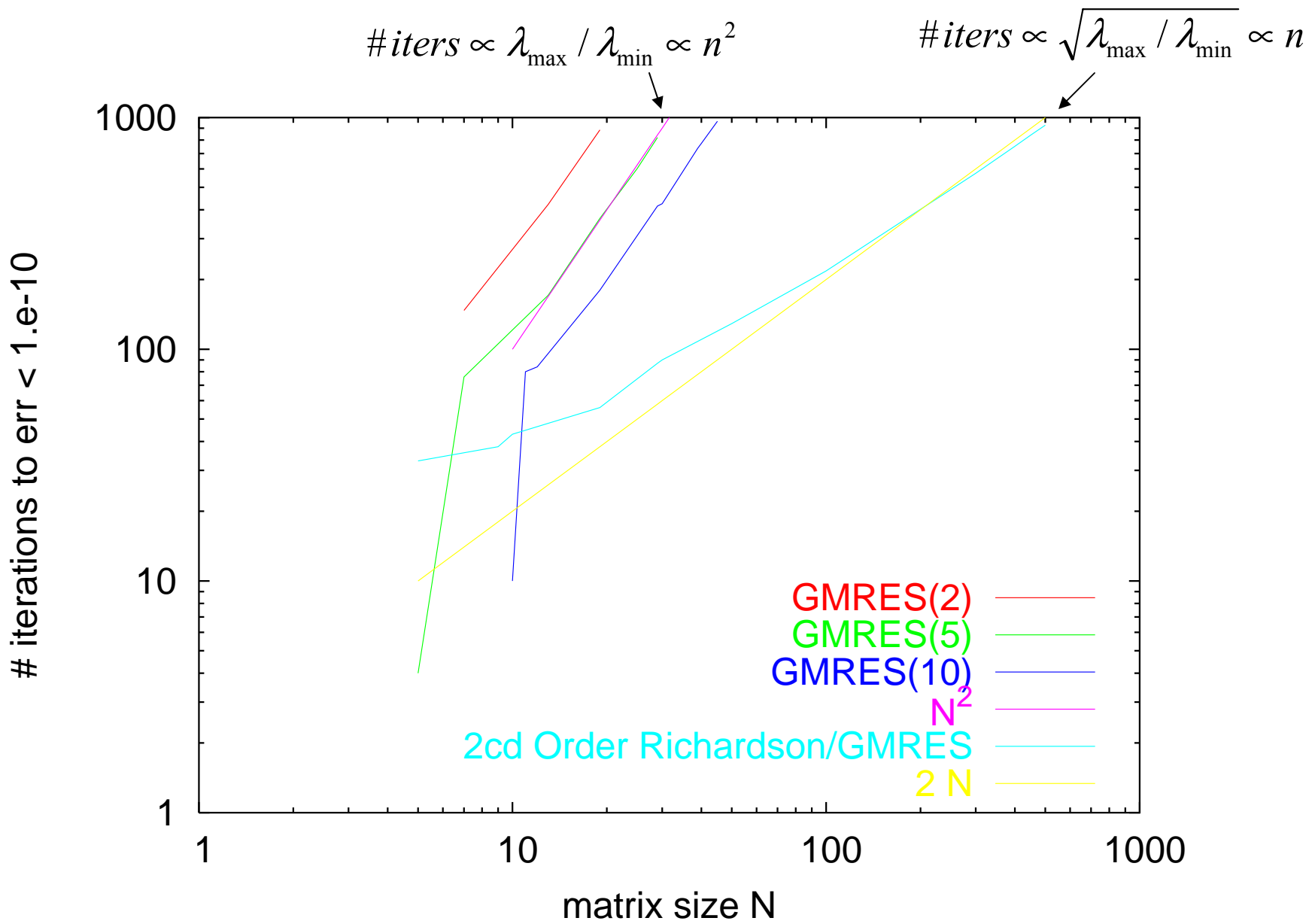
Loose GMRES/CG with  
modified ILU preconditioning

$\sim n^{3.5}$

# Algorithms and Moore's Law

- This advance took place over a span of about 36 years, or 24 doubling times for Moore's Law
- $2^{24} \approx 16$  million  $\Rightarrow$  the same as the factor from algorithms alone!





1-D Poisson Problem:  $\mathcal{L} \Phi / \partial x^2 = \rho$ , with  $n$  grid points

## Improved method for restarting GMRES

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- “The Tortoise and the Hare Restart GMRES,” Mark Embree (2003) <http://epubs.siam.org/sam-bin/dbq/article/39961> shows examples where GMRES(1) converges, GMRES(2) stagnates
- I developed a way to significantly improve restarted GMRES, by keeping information from previous search directions. Like “Dynamic Relaxation” (see Jardin’s course notes), but time step & damping rate chosen optimally each iteration to minimize error. (“Dynamic Relaxation” = 2cd order Richardson iteration, related to 2cd order Chebyshev iteration). Nonlinear/Newton-Krylov.

`/u/hammett/codes/gmres/dr_gmres`

- Turns out to be equivalent to LGMRES(1,1) “Loose GMRES”:  
“A Technique for Accelerating the Convergence of Restarted GMRES”, A. H. Baker, E. R. Jessup, T. Manteuffel, SIAM Journal on Matrix Analysis and Applications Volume 26, Number 4, pp. 962-984 (2005)
- Standard ILU preconditioning is of no benefit. Essential to use “Modified ILU” preconditioning, Gustafsson BIT 18, 142 (1978) (but MILU broken in Matlab)