

Is Galerkin Projection

The Right Scheme?

Edriss S. Titi

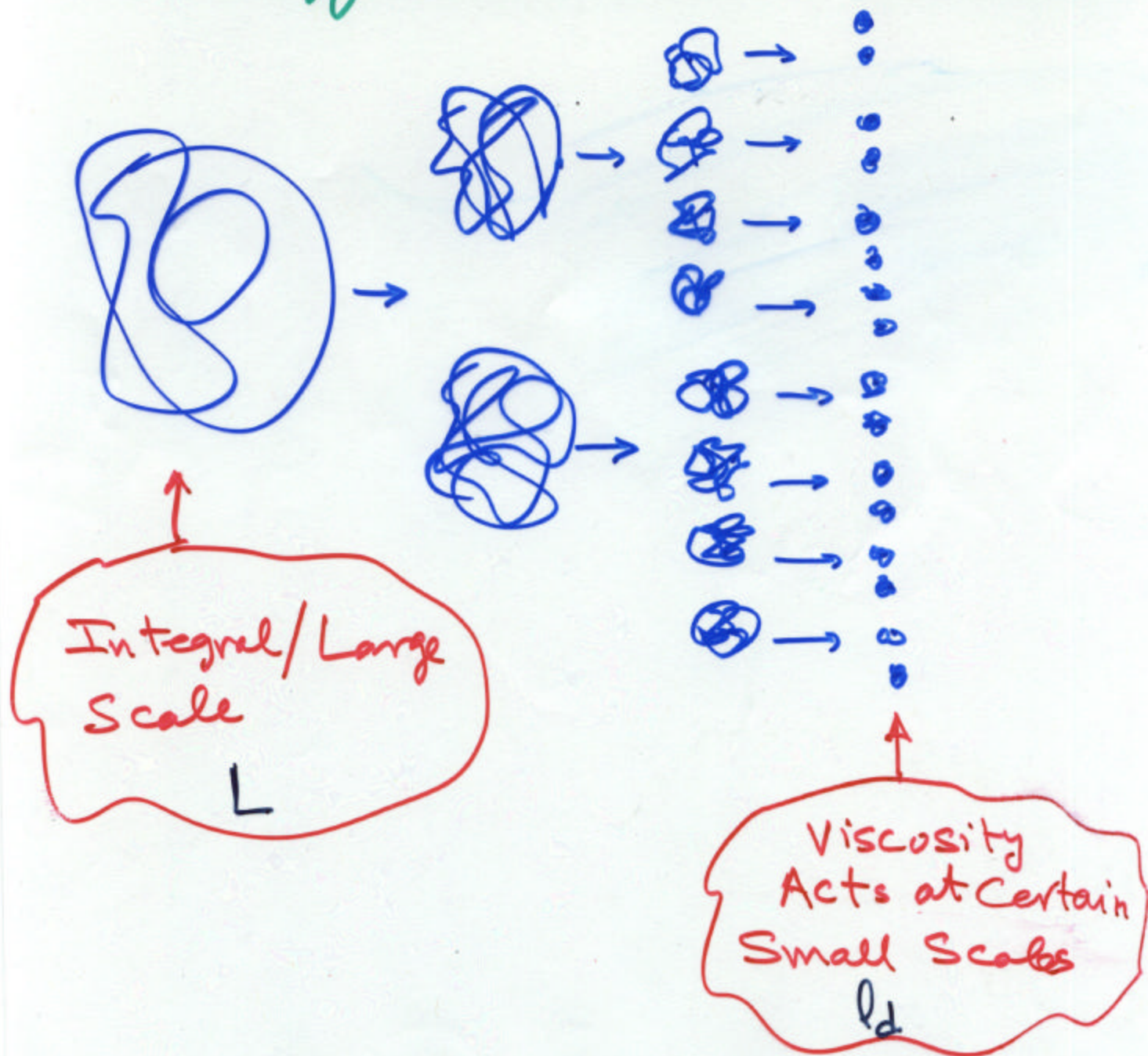
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Why Infinite Dimensional
Systems Can Be Reduced to
Finite Systems?

What is the Underlying
Mechanism(s) for Reduction?

Richardson/ Kolmogorov Scenario of Energy Cascade in Turbulence



The Number of Degrees of
Freedom in Turbulent Flows
is proportional to

$$\left(\frac{L}{l_0}\right)^3.$$

That is Finitely Many

Degrees of Freedom.

In Infinite Dimensional Dynamical
Systems.

The Navier-Stokes Equations of Viscous Incompressible Fluid:

$$\left\{ \begin{array}{l} \frac{\partial \vec{u}}{\partial t} - \nu \Delta \vec{u} + (\vec{u} \cdot \nabla) \vec{u} + \frac{1}{\rho_0} \nabla p = \vec{f} \\ \nabla \cdot \vec{u} = 0 \end{array} \right.$$

in $\Omega = [0, L]^3$ (Periodic Box)

$$\int_{\Omega} \nabla p \cdot \vec{u} \, dx = 0 \quad \& \quad \int_{\Omega} (\vec{u} \cdot \nabla) \vec{u} \cdot \vec{u} \, dx = 0$$

"Energy Balance"

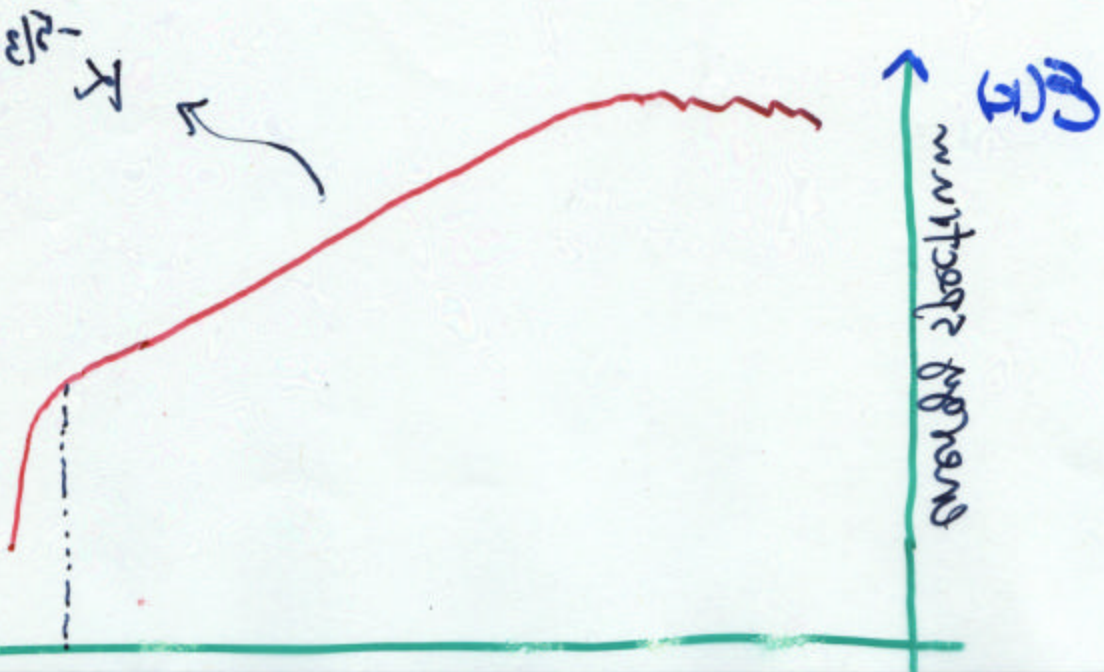
$$\frac{1}{2} \frac{d}{dt} \int_{\Omega} |\vec{u}(x, t)|^2 \, dx + \nu \int_{\Omega} |\nabla \vec{u}(x, t)|^2 \, dx = \int_{\Omega} \vec{f} \cdot \vec{u}(x, t) \, dx$$

Mean Rate of Dissipation of

Energy:

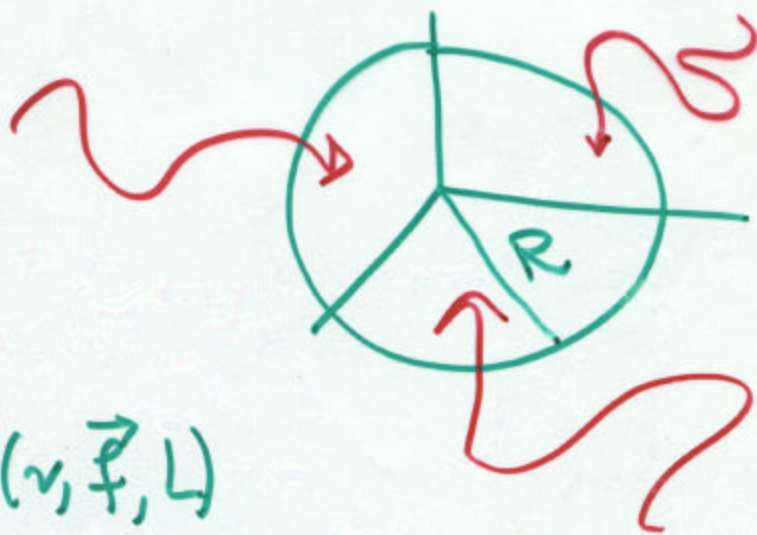
$$\epsilon = \left\langle \frac{1}{2} \rho \frac{d\langle u^2 \rangle}{dt} \right\rangle$$

$$\epsilon = \left(\frac{1}{2} \rho \right) \left(\frac{1}{4} \right)$$



Dynamical System Approach to Turbulence.

$$u(t) = S(t)u_0 \quad \text{Semi-group of
Solutions}$$



- $R(\nu, \vec{f}, L)$

- $B(0, R)$ An Absorbing Ball

- $A = \overline{\bigcap_{t \geq \tau} \bigcup_{t \geq \tau} S(t)B}$

$$d_H(A) \leq d_F(A) \leq N$$

$N \sim \#$ of degrees of freedom.

- In 2-D Navier-Stokes
[Constantin-Foias-Temam]
- In 3-D Navier-Stokes

???

2-D Navier-Stokes Equations

$$\begin{cases} \frac{\partial \vec{u}}{\partial t} - \nu \Delta \vec{u} + (\vec{u} \cdot \nabla) \vec{u} + \nabla p = f \\ \nabla \cdot \vec{u} = 0 \\ + \text{B.C.} \end{cases}$$

1-D Kuramoto-Sivashinsky Equation

$$\begin{cases} \frac{\partial u}{\partial t} + \frac{\partial^4 u}{\partial x^4} + \alpha \left(\frac{\partial^2 u}{\partial x^2} + u \frac{\partial u}{\partial x} \right) = 0 \\ \text{periodic B.C. on } [0, 2\pi], \\ \alpha > 0. \end{cases}$$

Complex Ginzburg-Landau Eq.

$$\frac{\partial u}{\partial t} + \alpha \frac{\partial^2 u}{\partial x^2} + \beta u + \gamma |u|^2 u = 0$$

$$\alpha, \beta, \gamma \in \mathbb{C}, \quad \operatorname{Re} \alpha < 0, \operatorname{Re} \beta < 0, \operatorname{Re} \gamma > 0.$$

$$\text{Periodic B.C. on } [0, 2\pi].$$

$$\begin{cases} \frac{du}{dt} + Au + R(u) = 0 \\ \text{in a Hilbert space } H. \end{cases}$$

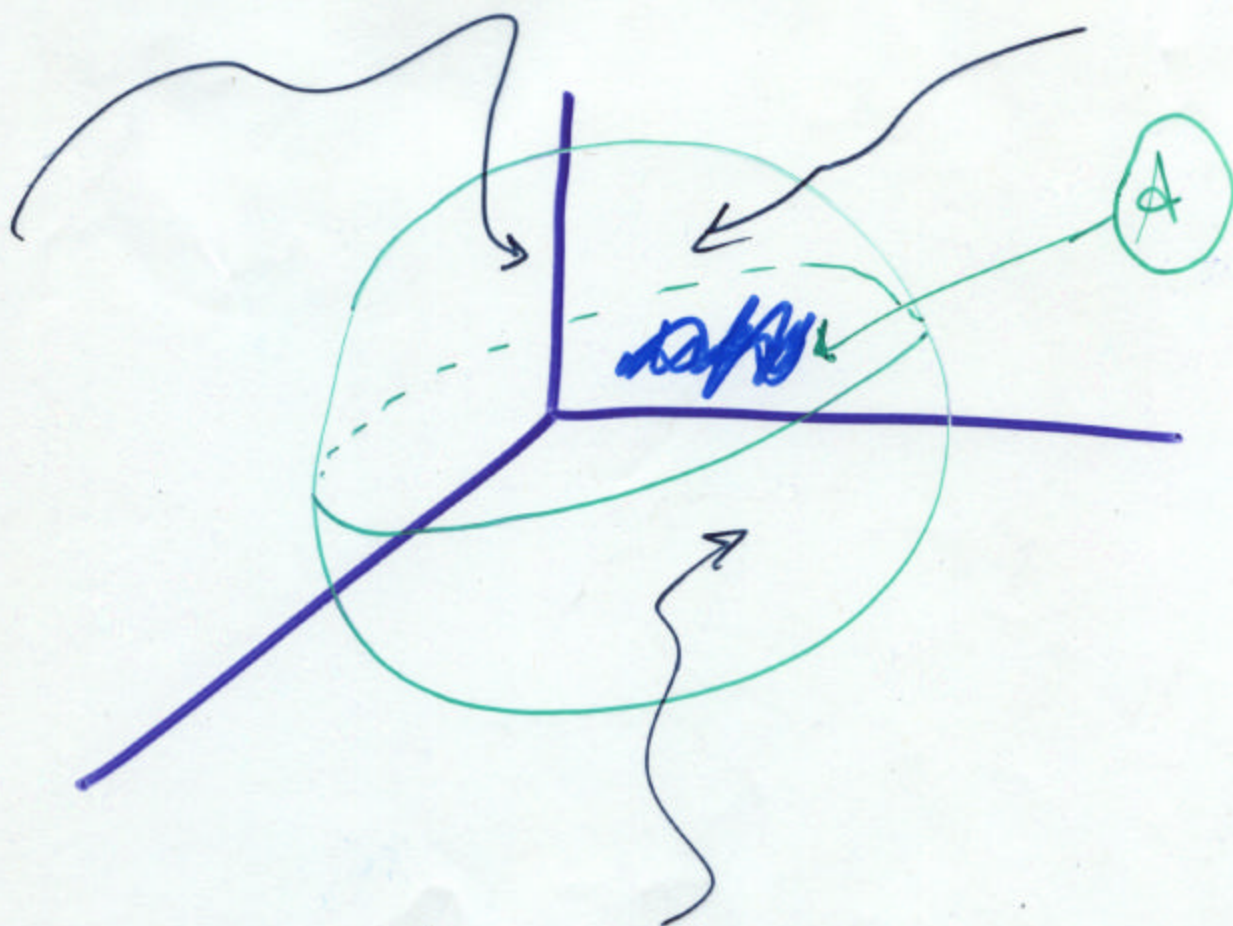
- $A > 0$, unbounded operator
- $Aw_j = \lambda_j w_j$, $\lambda_j \rightarrow \infty$, as $j \rightarrow \infty$
 $\{w_j\}$ basis of H
- $u(t) = S(t)u_0$ semigroup of nonlinear Compact operator.
- $\exists$ Ball $B \subset H$ s.t.
 for every E bounded set in H , $\exists T^*(E)$ for which
 $S(t) \in B$ for all $t \geq T^*$.

Dissipative System

- $$A = \bigcap_{t>0} \overline{\bigcup_{\tau \geq t} B}$$

finite dimensional Global Attractor

Indication of finite-dimension
long-term behavior



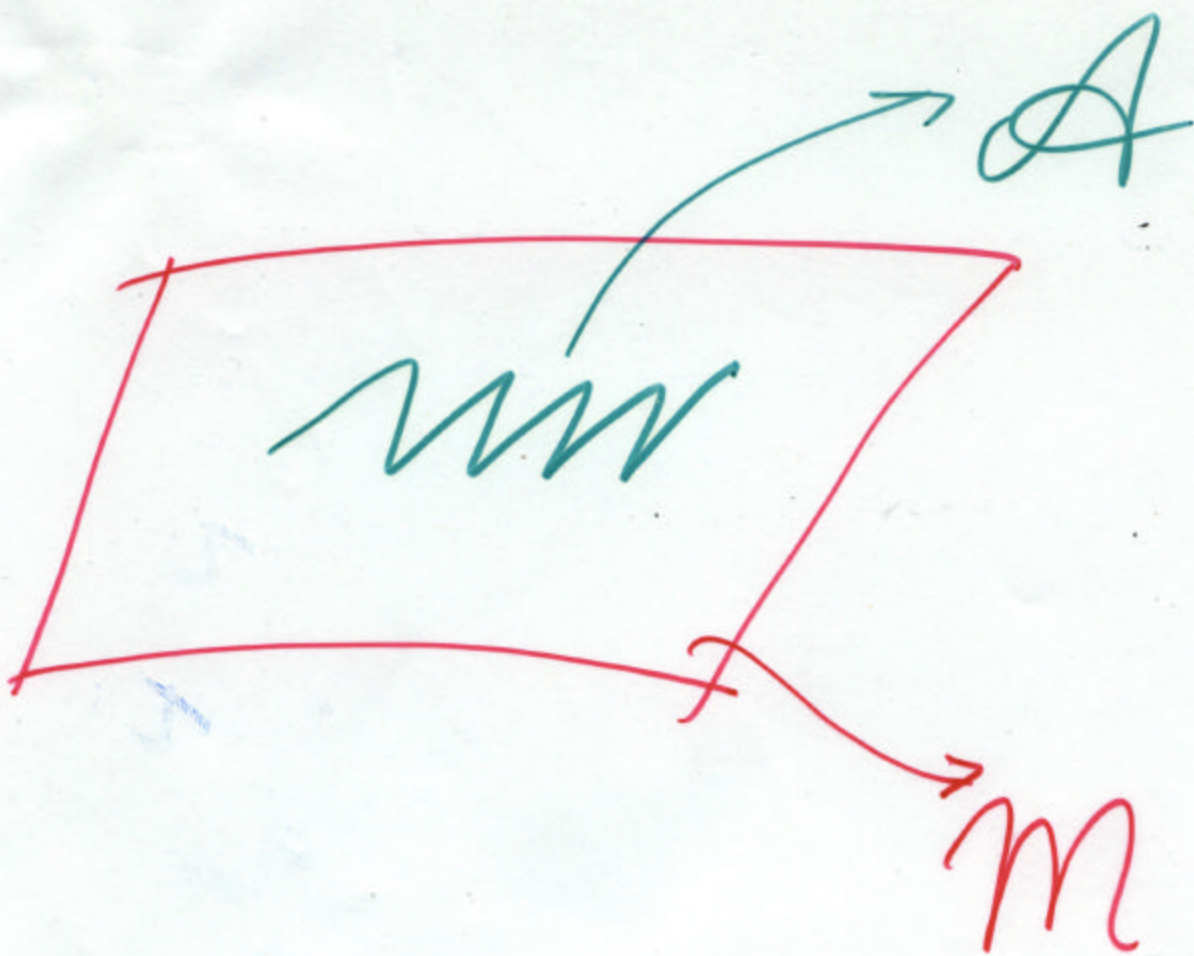
Inertial Manifold

(i) $M \subset H$ Lipschitz Finite-dimension
Manifold.

(ii) $S(t)M \subset M$

(iii) $\text{dist}(S(t)E, M) \rightarrow 0, \text{ as } t \rightarrow \infty$

Exponentially fast, $\forall E \subset H$ bounded.



$$M = \text{graph } \Phi$$

- $H_m = \text{Span} \{w_1, w_2, \dots, w_m\}$

- P_m Orthogonal projection of H onto H_m .

- $Q_m = I - P_m$.

- $\frac{du}{dt} + Au + R(u) = 0$

Equivalent to

- $\begin{cases} \frac{dp}{dt} + Ap + P_m R(p+q) = 0 \\ \frac{dq}{dt} + Aq + Q_m R(p+q) = 0 \end{cases}$

$p = P_m u, \quad q = Q_m u, \quad u = p + q$

- Search for Inertial Manifolds

$$\mathcal{M} = \text{graph}(\Phi)$$

$$\Phi: H_m \rightarrow H_m^\perp$$

for $m \gg 1$

- That is $q(t) = \Phi(p(t))$ as $t \rightarrow \infty$

- That is

$$\begin{cases} \frac{dp}{dt} + Ap + P_m R(p + \Phi(p)) = 0 \\ q = \Phi(p) \end{cases} \quad \begin{matrix} \text{[Inertial Form /} \\ \text{ODE]} \end{matrix}$$

is equivalent to you

$$\bullet \frac{du}{dt} + Au + R(u) = 0, \quad u = p + \Phi(p)$$

Sufficient Condition
for Existence of
An Inertial Manifold
Of Dimension m Is
That

$$\lambda_{m+1} - \lambda_m \gg 1$$

I.e.

Separation In Scales

FOIAS-PRODI 1987, DETERMINING
MODES.

THERE EXISTS $M_0(G)$ SUCH THAT
IF u AND v ARE TWO EXACT
SOLUTIONS TO THE NAVIER-STOKES
EQUATIONS SATISFYING

$$\|P_m u - P_m v\| \rightarrow 0 \text{ As } t \rightarrow \infty$$

FOR SOME $m \geq M_0(G)$ THEN

$$\|u - v\| \rightarrow 0 \text{ As } t \rightarrow \infty.$$

i.e.

$$\|(I - P_m)u - (I - P_m)v\| \rightarrow 0$$

$$\text{As } t \rightarrow \infty.$$

Application of the Determining Modes to Continuous Data Assimilations:

Consider NSE

$$\frac{du}{dt} + \gamma Au + B(u, u) = f$$

equivalent to

$$\frac{dp}{dt} + \gamma Ap + PB(p+q, p+q) = Pf$$

$$\frac{dq}{dt} + \gamma Aq + QB(p+q, p+q) = Qf$$

Suppose $p(t)$ is given
from measurements.

Find $q(t)$ (small
scales)!

The problem is $q(0)$?
(Initialization).

Theorem (Olson-T.)

Let $\tilde{q}$ solves

$$\frac{d\tilde{q}}{dt} + \gamma A \tilde{q} + Q B(p + \tilde{q}, p + \tilde{q}) = Qf$$
$$\tilde{q}(0) = 0$$

. If $m \gg 1 \Rightarrow \|q(t) - \tilde{q}(t)\| \rightarrow 0.$
 $t \rightarrow \infty.$

OTHER PARAMETERIZATION OF INERTIAL MANIFOLD

Determining Nodes

Foias-Terman



$$\mathcal{E} = \{x_1, \dots, x_N\}$$

Suppose

$$|u(x_j, t) - v(x_j, t)| \xrightarrow[t \rightarrow \infty]{} 0$$

$$\Rightarrow \|u(x, t) - v(x, t)\| \longrightarrow 0 \\ \text{as } t \rightarrow \infty.$$

Determining - Volume ELEMENTS

Foias-Titi



$$\Omega = \bigcup_{j=1}^N Q_j$$

$$\text{If } \int_{Q_j} u(x,t) - v(x,t) \xrightarrow{t \rightarrow \infty} 0$$

$$\Rightarrow \|u - v\| \xrightarrow{t \rightarrow \infty} 0$$

2-D Navier-Stokes best
estimates on N are due
to Jones-Titi

Questions:

Can one have similar results
concerning continuous data
assimilations from point value
measurements?

Namely, if $U(x_j, t)$, $j=1, \dots, N$
are given can we find an
algorithm for computing $U(x, t)$,
say, asymptotically in time?

LONG-TERM DYNAMICS & POST-PROCESSING

GALERKIN METHOD

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- B. Garcia - Archilla
 - J. Novo
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- L. Margolin
 - S. Wynne
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- Devulder
- C. Foias
- I. Kevrekidis
- M. Jolly
- D. Jones
- M. Marion
- G. Sell
-
-
-

- M. Graham
- P. Steen
-
-
-

WORK WITH

B. GARCIA-ARCHILLA

&

J. NOVO

(SPAIN)

Motivated by Inertial
Manifolds where

$$q(t) = \Phi(p(t)).$$

Consider an Approximate
Inertial Manifolds

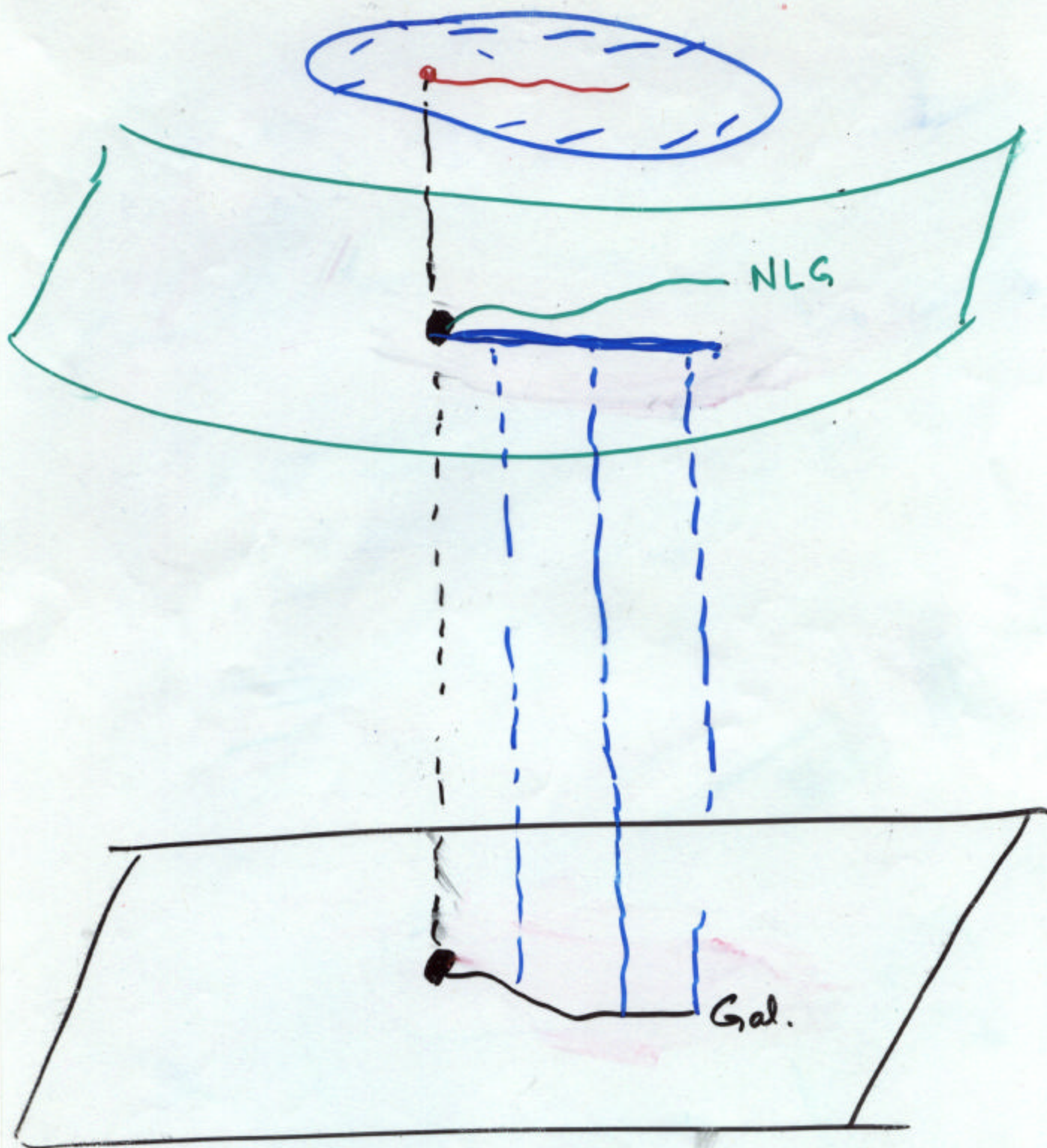
$$q \approx \Psi(p)$$

[Parametrization of Small
Scales in terms of Large Scales].

• Nonlinear Galerkin:

$$\frac{dp}{dt} + \nu A p + P R(p + \Psi(p)) = 0$$

$$u_{app} = p + \Psi(p).$$



Galerkin:

$$\frac{dy_m}{dt} + A y_m + P_m R(y_m) = 0$$

$$u_{app} = y_m$$

Nonlinear Galerkin

$$\frac{du_m}{dt} + A u_m + P_m R(u_m + \psi(u_m)) = 0$$

$$u_{app} = u_m + \psi(u_m)$$

New Idea

$$\frac{dy_m}{dt} + Ay_m + P_m R(y_m) = 0$$

Galerkin

- $u_{app} = y_m + \psi(y_m)$

post-processing.

Properties:

- The "Same" CPU Like Galerkin
- ACCURACY: NOT WORSE THAN GALERKIN, BUT POTENTIALLY AS GOOD AS THE NONLINEAR GALERKIN.
- CAN BE EASILY IMPLEMENTED FOR FINITE ELEMENTS OR OTHER SPECTRAL METHOD

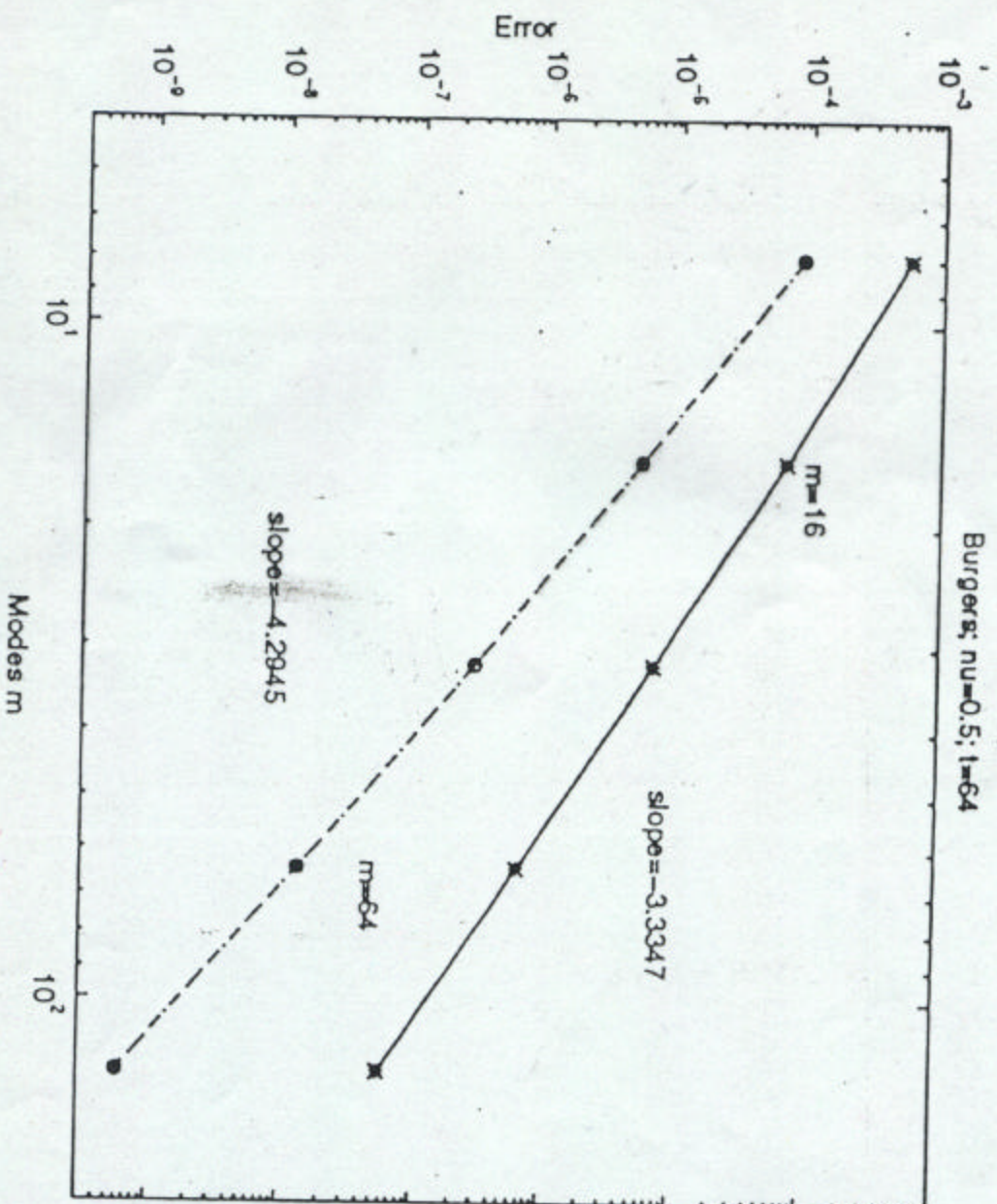


Figure 1: Errors for Burgers' equation with $\nu = 0.5$: * standard Galerkin
 o NLG, + Postprocessed Galerkin

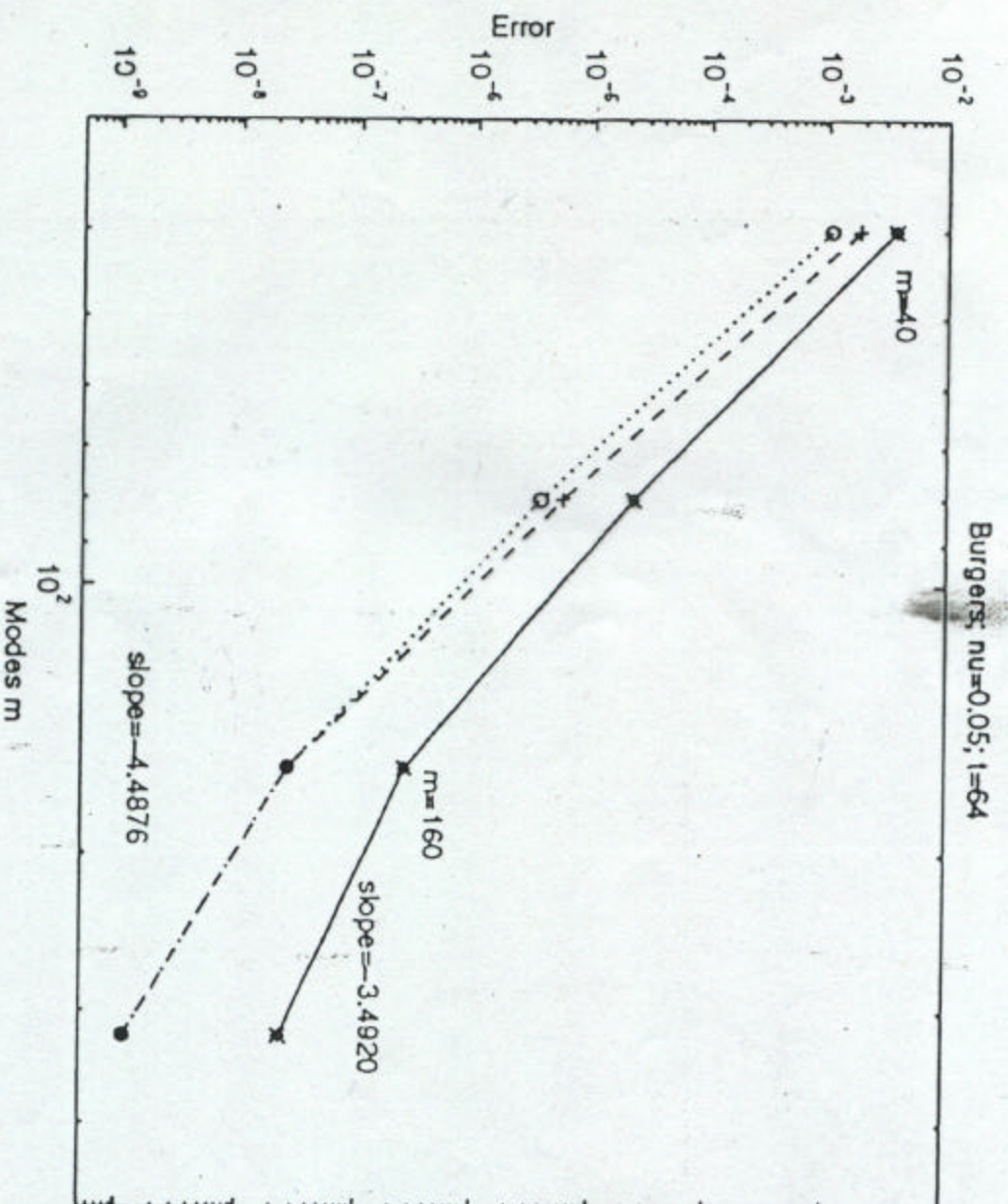
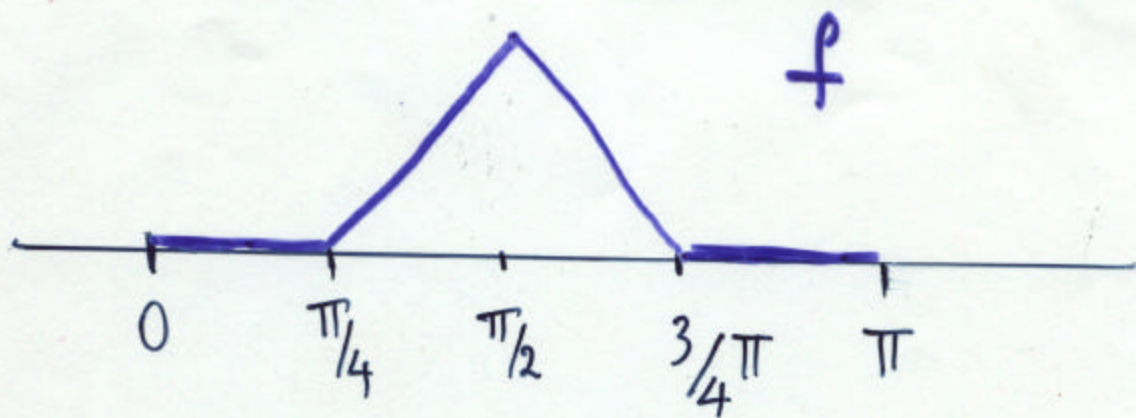
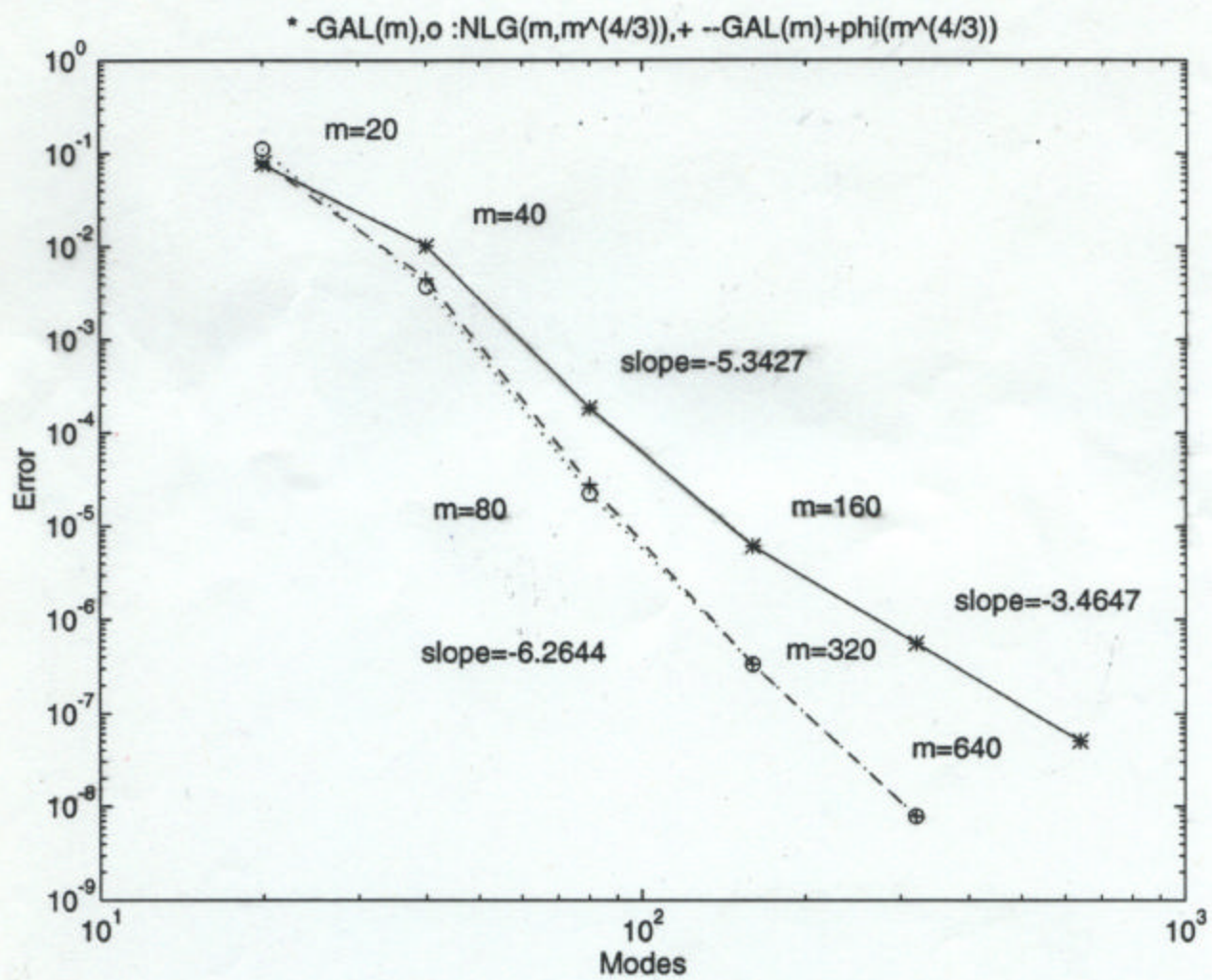


Figure 2: Errors for Burgers' equation with $\nu = 0.05$: * standard Galerkin
 o NLG, + Postprocessed Galerkin

Computations for:

$$\frac{\partial u}{\partial t} - \frac{\partial^2 u}{\partial x^2} + u^3 - u = f$$





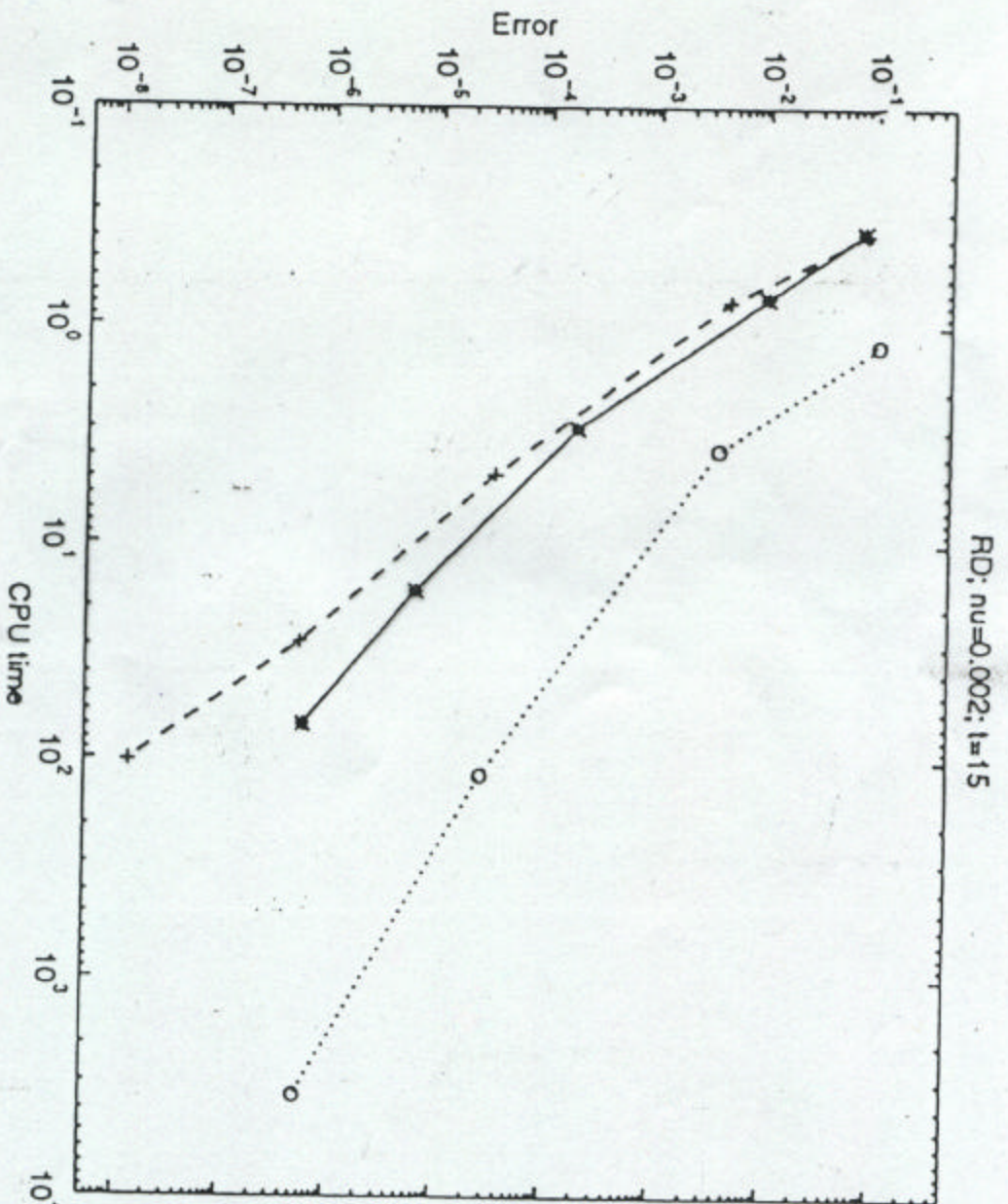


Figure 4: Efficiency diagram for the RD equation with $\nu = 0.002$: * standard Galerkin, o NLG, + postprocessed Galerkin.

THEOREM 1 (Gauss-Markov Theorem)

CONSIDER THE EQUATION

$$0 = (y)^\top + N A + \frac{N b}{\tau}$$

$$A^\top W y = A^\top W b$$

$$\infty \leftarrow y_1 \geq \dots \geq y_s \geq 1, y_s > 0$$

$$(y_{m+1}^\top)^\top 0 = \sqrt{s} \|A^\top N - N\|$$

$$(y_{m+1}^\top)^\top 0 = \sqrt{s} \|A^\top N - N\|$$

$$(y_{m+1}^\top)^\top 0 = \sqrt{s} \|A^\top N - N\|$$

FINITE ELEMENT CASE

(Garcia-Archilla & T.)

$$\frac{du}{dt} + \nu Au + F(u) = 0$$

$$\nu > 0.$$

$$A = -\Delta \quad \text{Dirichlet B.C.}$$

$$a(u, v) = (A^{\frac{1}{2}} u, A^{\frac{1}{2}} v) = (\nabla u, \nabla v)$$

$S_{h,r}$ - Finite Element Space

P_h - L^2 Orthogonal Proj.
onto $S_{h,r}$.

- $a_h(x, \psi) = (\nabla x, \nabla \psi)$
 $x, \psi \in S_{h,r}$

- Inner Product on $S_{h,r}$

- A_h

$$a_h(x, \psi) = (A_h x, \psi) = (x, A_h \psi)$$

- R_h ORTHOGONAL PROJECTION
 ON $S_{h,r}$ WITH RESPECT TO
 $a_h(\cdot, \cdot)$.

- GALERKIN:

$$\frac{du_h}{dt} + \nu A_h u_h + P_h F(u_h) = 0$$

$$u_h(0) = R_h u(0)$$

$$\nabla A u(T) = -F(u(T)) - \frac{du}{dt}(T)$$

IN THE R.H.S. APPROX. $u(T) = u_h(T)$

$$\nabla A u(T) \approx \boxed{\nabla A \tilde{u} = -F(u_h(T)) - \frac{du_h}{dt}(T)}$$

- SOLVE AN ELLIPTIC PROBLEM @ FINAL TIME.

THEOREM 1: Let $r \geq 3$. $\exists C > 0$ s.t.

$\forall h$ small enough:

$$\|u - \tilde{u}\|_{L^2(\Omega)} + h \|u - \tilde{u}\|_{H^1(\Omega)} \leq C h^{r+\bar{\mu}} |\log h|^{\bar{\nu}}$$

$$\bar{\mu} = \begin{cases} 2 & \text{if } r \geq 4 \\ 1 & \text{if } r = 3 \end{cases}$$

$$\bar{\nu} = \begin{cases} 0 & \text{if } r \geq 4 \\ 1 & \text{if } r = 3 \end{cases}$$

$$K_{\infty}(u) = \max_{0 \leq t \leq T} (\|u(\cdot, t)\|_{W^{r, \infty}} + \|u_t(\cdot, t)\|_{W^{r, \infty}})$$

THEOREM 2:

$\mathbb{R}^d, d=2,3; r \geq 3. \exists C(K_{\infty}(u)) > 0$
 s.t. $\forall h$ small enough:

$$(i) \|u(T) - \tilde{u}\|_{L^{\infty}} \leq C h^{r+2-\frac{d}{2}} |\log h|^{4-\frac{d}{2}} \quad \text{if } r=3$$

$$(ii) \|u(T) - \tilde{u}\|_{L^{\infty}} \leq C h^{r+2} |\log h|^3 \quad \text{if } r \geq 4$$

for some $C = C(\|v\|_{H^r(\Omega)})$. Thus, from (60), (61) and (62), it follows that

$$\|A^{-1}(b(v) \cdot \nabla \epsilon)\|_{L^2(\Omega)} \leq C \|\nabla A^{-1} \epsilon\|_{L^2(\Omega)}. \quad (63)$$

Finally, the second term on the right-hand side of (58) can be treated similarly as the first-one (i.e. after applying A^{-1} it can be bounded by the right-hand side of (63). This, together with (59) lead to

$$\|A^{-\mu/2}(b(v) \cdot \nabla v - F(\chi) \cdot \nabla \chi)\|_{L^2(\Omega)} \leq C(\|A^{(1-\mu)/2}(v-\chi)\|_{L^2(\Omega)} + \|v-\chi\|_{H^1(\Omega)}^2), \quad (64)$$

for $\mu = 2$.

In the case $\mu = 1$, notice first that $\|A^{-1/2} \cdot\|_{L^2(\Omega)} = \|\nabla A^{-1} \cdot\|_{L^2(\Omega)}$. Then, go back to the decomposition (58) and estimate again the first two terms by duality (recall that the third one has already been dealt with). For example, for $\phi \in C_0^\infty(\Omega)^2$, integrating by parts twice one gets $(\nabla A^{-1}(b(v) \cdot \nabla \epsilon), \phi) = (\epsilon, \operatorname{div}((A^{-1} \operatorname{div}(\phi))b(v)))$ from where it is easy to proceed as in the case $\mu = 2$. Finally, the bound (64) in the case $\mu = 0$ is a direct consequence of (58).

The arguments above can be (more easily) applied to the reaction term, yielding

$$\|A^{-\mu/2}(g(v) - g(\chi))\|_{L^2(\Omega)} \leq C(\|A^{-\mu/2}(v-\chi)\|_{L^2(\Omega)} + \|v-\chi\|_{L^q(\Omega)} \|v-\chi\|_{L^2(\Omega)}),$$

for $\mu = 0, 1, 2$. This and (64) conclude (57). $\square$

3 Numerical Experiments

We consider the reaction-diffusion system

$$\left. \begin{aligned} \frac{\partial u_1}{\partial t} &= 1 - 4u_1 + u_1^2 u_2 + \nu \Delta u_1 \\ \frac{\partial u_2}{\partial t} &= 3u_1 - u_1^2 u_2 + \nu \Delta u_2 \end{aligned} \right\} \quad (65)$$

known as the Brusselator (see e.g. [23]), in $\Omega \times [0, T]$, where $\Omega = [0, 1]^2$ and $T = 10$, subject to

$$\frac{\partial u_1}{\partial n} = \frac{\partial u_2}{\partial n} = 0.$$

We set $\nu = 0.002$. We complement this system with the initial condition

$$\begin{aligned} u_1(x, y, 0) &= 1 - \frac{1}{4}(2y - 1)((2y - 1)^2 - 3), \\ u_2(x, y, 0) &= 3 - (2x - 1)((2x - 1)^2 - 3). \end{aligned}$$

The corresponding solution becomes nearly periodic in time, and develops moderately large gradients. Fig. 1, shows the evolution of $\|u\|_{L^2(\Omega)^2}$ up to

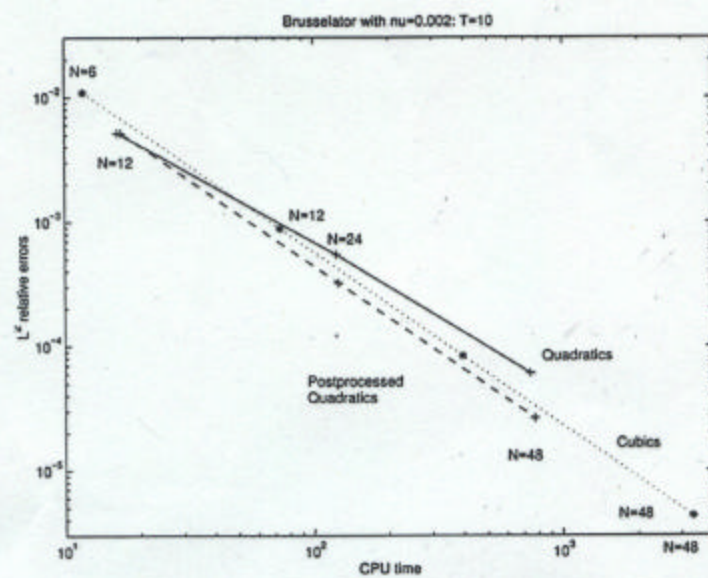


Figure 5: Efficiency diagram

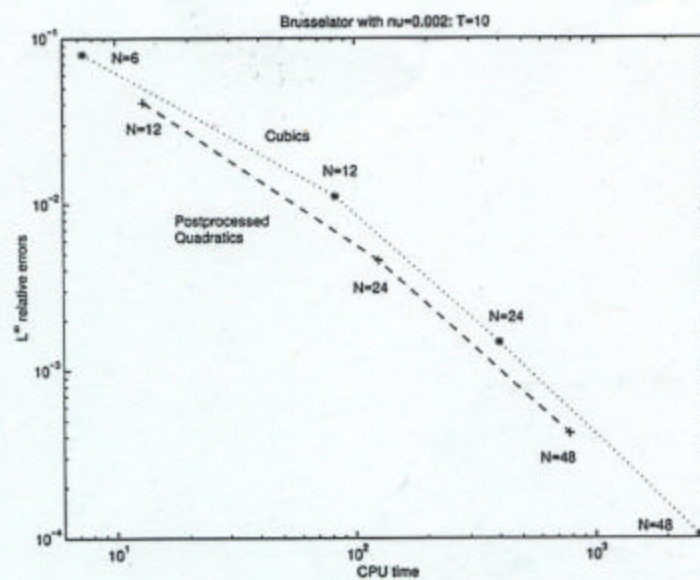


Figure 6: Efficiency diagram in L^∞

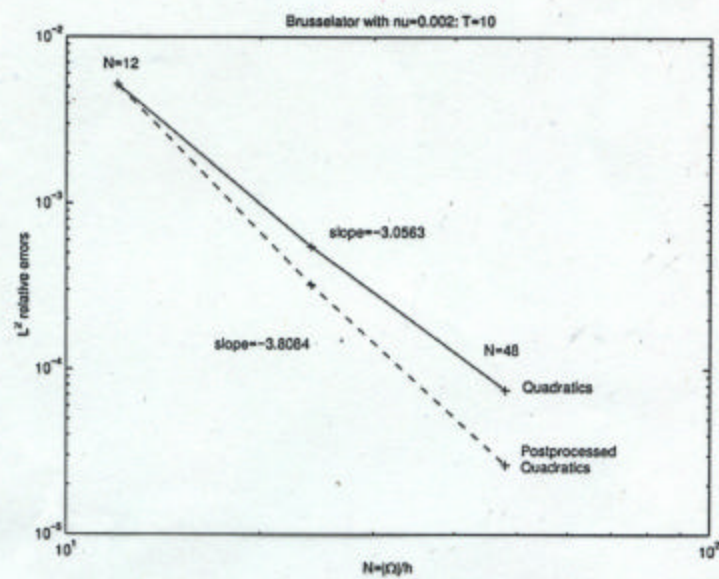


Figure 3: Convergence diagram

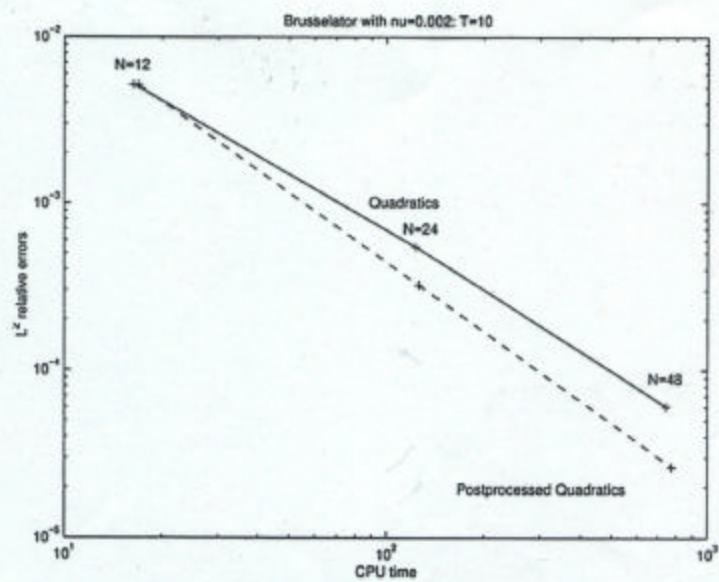


Figure 4: Efficiency diagram

Remark

This is different than
any two-level multigrid
method.

TRUNCATION ANALYSIS

- L. Margolin
- S. Wynne
- E. S. T.

Recall The Navier-Stokes Equations

$$\left\{ \begin{array}{l} \frac{\partial \vec{u}}{\partial t} - \nu \Delta \vec{u} + \vec{u} \cdot \nabla \vec{u} + \nabla p = f \\ \nabla \cdot \vec{u} = 0 \\ + \text{B.C.} \end{array} \right.$$

is equivalent to:

- $\frac{du}{dt} + \nu Au + B(u, u) = f$
- $u = p + q$, $p \in H_m$, $q \in H_m^\perp$

$$\|q\|_{L^2} = O(\lambda_{m+1}^{-1})$$

$$\left\| \frac{dq}{dt} \right\|_{L^2} = O(\lambda_{m+1}^{-1})$$

$$\|\nabla q\|_{L^2} = O(\lambda_{m+1}^{-1/2})$$

$$\|\Delta q\|_{L^2} = O(1)$$

- Let $\varepsilon = \lambda_{m+1}^{-1/2}$

Small parameter

- $\|q\|_{L^2} = O(\varepsilon^2)$, $\|\frac{dq}{dt}\|_{L^2} = O(\varepsilon^2)$

- $\|\nabla q\|_{L^2} = O(\varepsilon)$, $\|\Delta q\|_{L^2} = O(1)$

The above estimates are
valid on the Global
Attractor.

$$\begin{aligned} \underline{\frac{dP}{dt}} + \underline{rA_P} + \underline{P_m B(p,p)} + P_m B(p,q) + \\ + P_m B(q,p) + P_m B(q,q) = \underline{P_m f} \end{aligned}$$

$$\begin{aligned} \boxed{\frac{dq}{dt}} + \underline{rA_q} + \underline{Q_m B(p,p)} + Q_m B(p,q) + \\ + Q_m B(q,p) + Q_m B(q,q) = \underline{Q_m f} \end{aligned}$$

Thus to leading Order:

$$\frac{dp}{dt} + \nu A p + P_m B(p, p) = P_m f$$

$$\boxed{\frac{dq}{dt} +} \nu A q + Q_m B(p, p) = Q_m f$$

Post-processing &

NOT Galerkin.

General Postprocessing Schemes With Applications to Data Assimilations.

- We have the equation

$$\frac{du}{dt} + rAu + R(u) = 0.$$

- We are interested in $u(t=T)$.

$$rAu(T) = - \left. \frac{du}{dt} + R(u) \right|_{t=T}$$

- The Right Hand is usually given on a coarse mesh.
We use the Coarse data to compute the Left hand by inverting (rA) as accurate as possible.

• Dynamical Post-processing Galerkin

- In the case of transient solutions or time-dependent oscillatory forcing

$\left\| \frac{dq}{dt} \right\|_{L^2}$ might be large

$$\begin{cases} \frac{dp}{dt} + rAp + P_m B(p,p) = P_m f & \leftarrow \text{Galerkin} \\ \frac{dq}{dt} + rAq + Q_m B(p,p) = Q_m f & \leftarrow \end{cases}$$

Dynamical Post-processing
Galerkin

Dynamical post-processing Galerkin - Finite Elements Case:

① Galerkin step

$$\begin{cases} \frac{\partial u_h}{\partial t} + A u_h + P_h R(u_h) = 0 \\ u_h(0) = \text{projection of } u_0 \end{cases}$$

② Post-processing step (Dynamical)

$$\begin{cases} \frac{\partial \tilde{u}}{\partial t} + A \tilde{u} + R(u_h) = 0 \\ \tilde{u}(0) = u_0 \end{cases}$$

$$u_{\text{app}}(t) = \tilde{u}(t)$$

$$u_{\text{exact}} = \sum_{k=1}^{\infty} \frac{a_k(t)}{k^3} \sin kx$$

$$a_k(t) = \begin{cases} 1 + \gamma \sin k^2 t & 1 \leq k \leq 10^5 \\ 1 & 100 < k \end{cases}$$

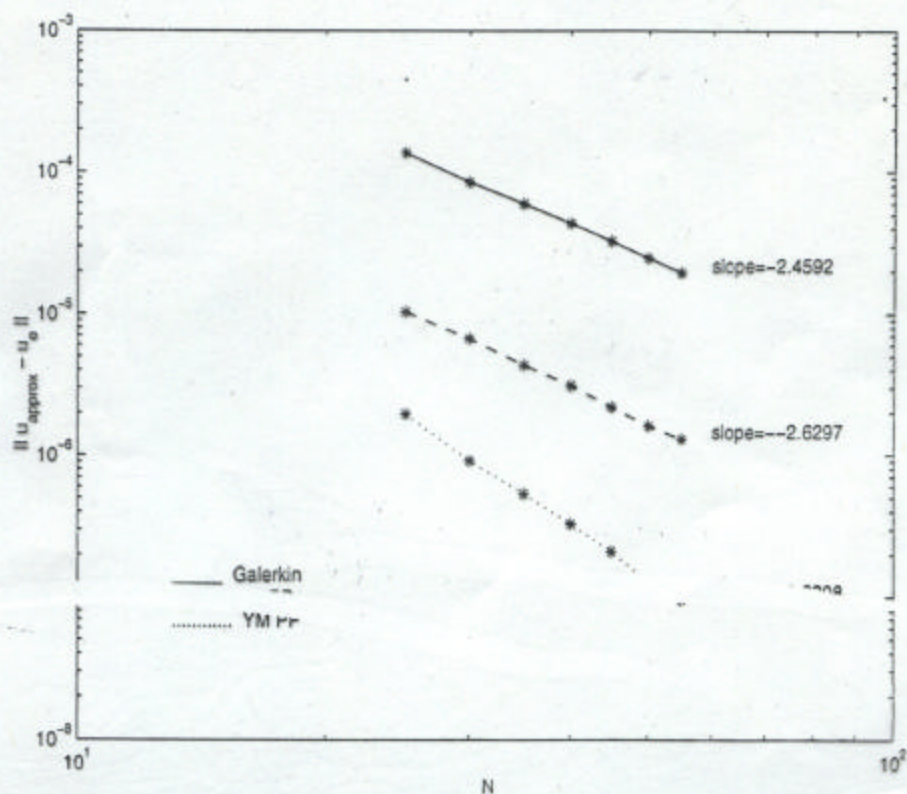


Figure 1: Total errors $\|u_{\text{approx}} - u_e\|$

$$t = 2.0$$

$$\gamma = 0.1$$

Forced-viscous Burgers