

Flow control applied to transitional flows: input-output analysis, model reduction and control



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Outline

- Introduction with input-output configuration
- Matrix-free methods for flow stability using Navier-Stokes snapshots
Edwards *et al.* (1994), ...
- Global modes and transient growth
Cossu & Chomaz (1997), ...
- Input-output characteristics of Blasius BL and reduced order models based on balanced truncation
Rowley (2005), ...
- LQG feedback control based on reduced order model
- Conclusions



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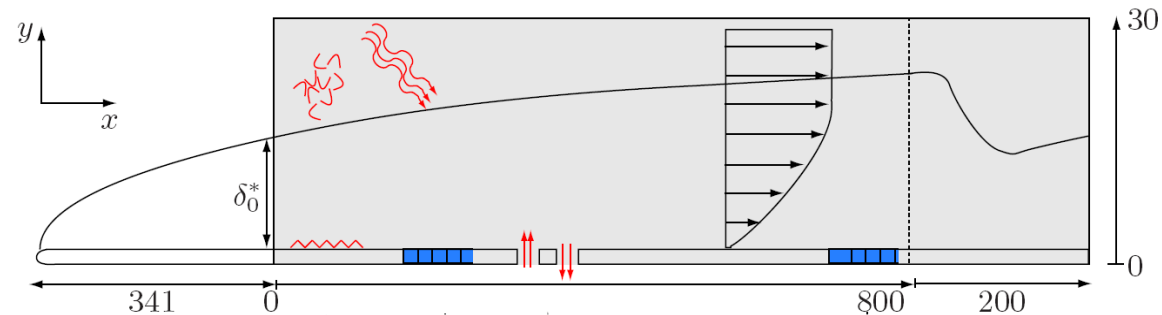
Message

- Need only snapshots from a Navier-Stokes solver (with adjoint) to perform stability analysis and control design for complex flows
- Main example Blasius BL, but many other more complex flows are now tractable ...



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Linearized Navier-Stokes for Blasius flow



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Continuous formulation

$$\begin{aligned} \frac{\partial u}{\partial t} &= \mathcal{A}u - \nabla p \\ 0 &= \nabla \cdot u \\ u &= u_0 \quad \text{at } t = 0 \end{aligned}$$

$$\mathcal{A} = -(U \cdot \nabla) - (\nabla U^T)^T + \frac{1}{Re} \nabla^2 + \lambda(x)$$

$$Re = \frac{U_\infty \delta_0^*}{\nu} = 1000$$

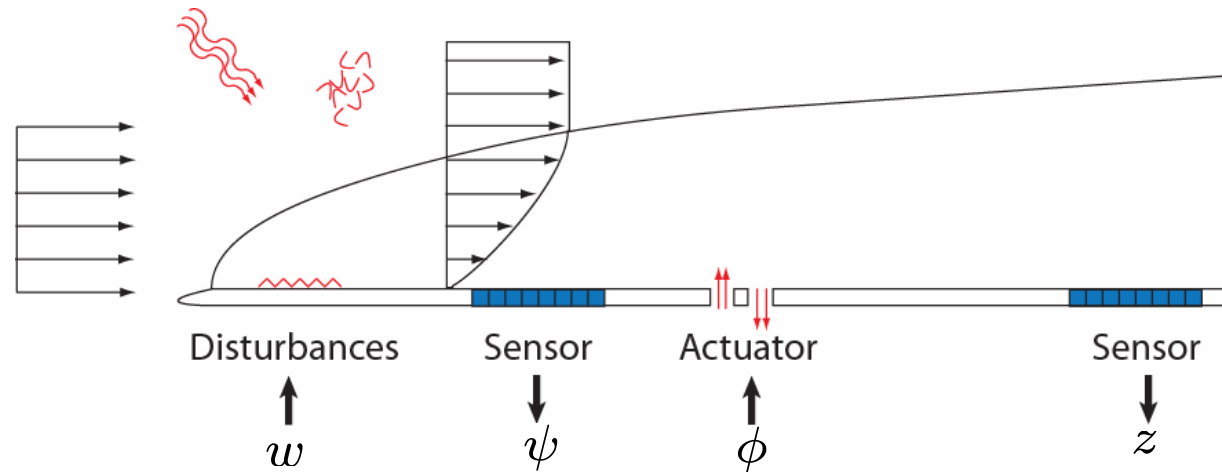
Discrete formulation

$$\begin{aligned} \frac{du}{dt} &= \mathcal{A}u \\ u &= u_0 \quad \text{at } t = 0 \end{aligned}$$

Input-output configuration for linearized N-S



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$$\frac{du}{dt} = Au + Bf$$

$$y = Cu$$

$$u = u_0 \quad \text{at } t = 0$$

$$B_1 f_1 = B_1 w : \text{incoming dist.}$$

$$B_2 f_2 = B_2 \phi : \text{actuator input}$$

$$y_1 = \psi = C_1 u : \text{sensor output}$$

$$y_2 = z = C_2 u : \text{objective function}$$

$$B_1 = h(x_i) = \text{"Gaussian"}$$

$$C_1 u = (h, u) = h^T u$$

Solution to the complete input-output problem

$$y(t) = \underbrace{C e^{At} u_0}_{\text{IVP}} + \underbrace{C \int_0^t e^{A(t-\tau)} B f(\tau) d\tau}_{\text{forced solution}}$$



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- Initial value problem: flow stability
- Forced problem: input-output analysis



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The Initial Value Problem

- Asymptotic stability analysis:
Global modes of the Blasius boundary layer
- Transient growth analysis:
Optimal disturbances in Blasius flow

Dimension of discretized system

	Base Flow	Inhomogeneous direction(s)	Dimension of $\mathbf{u}(t)$	Storage of A
Ginzburg-Landau	$U(x)$	1D	10^2	1 MB
Blasius	$U(x, y)$	2D	10^5	25 GB
Jet in crossflow	$U(x, y, z)$	3D	10^7	500 TB



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- Matrix A very large for complex spatially developing flows
- Consider eigenvalues of the matrix exponential, related to eigenvalues of A

$$\lambda_j = e^{\omega_j t}$$

- Use Navier-Stokes solver (DNS) to approximate the action of matrix exponential or evolution operator

$$\mathbf{u}(t) = e^{At} \mathbf{u}_0 = T(t) \mathbf{u}_0$$

Krylov subspace with Arnoldi algorithm

- Krylov subspace created using NS-timestepper
- Orthogonal basis created with Gram-Schmidt
- Approximate eigenvalues from Hessenberg matrix H



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Krylov subspace: $\{v_1, e^{At}v_1, \dots, (e^{At})^{m-1}v_1\}$

orthogonal basis: $V = \{v_1, v_2, \dots, v_m\}$

$$\Rightarrow e^{At} \approx VHV^T \quad H : m \times m$$

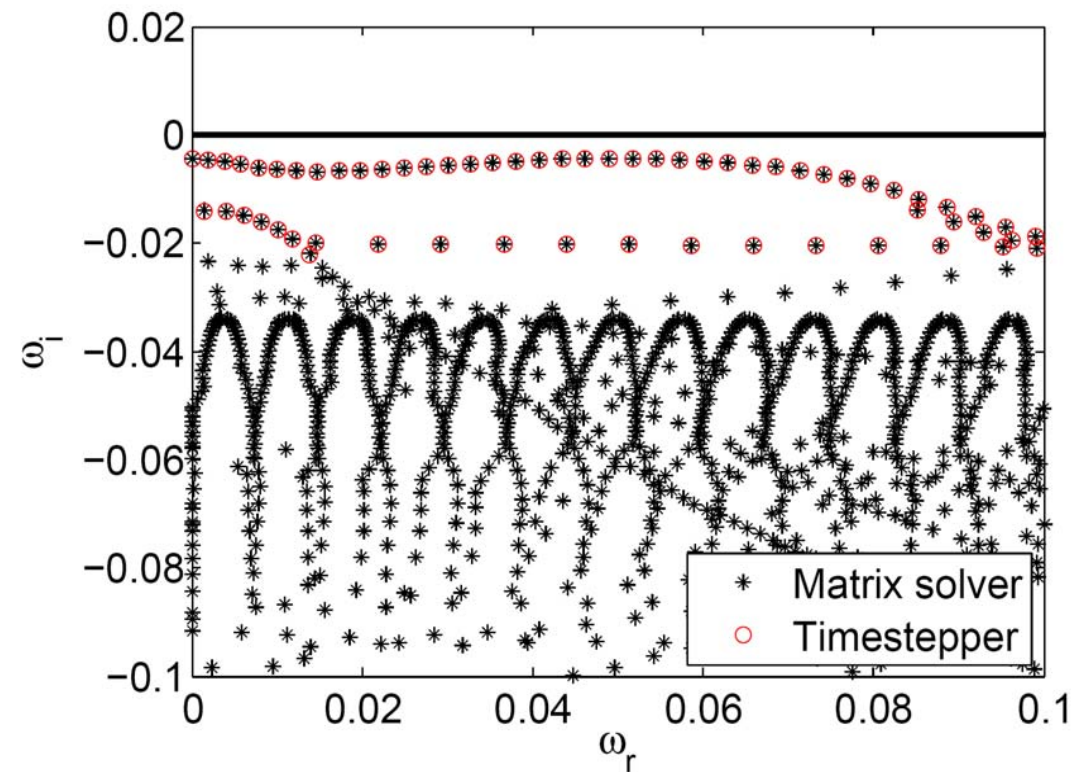
$$\text{eigenvalues: } H = E\tilde{\Lambda}E^{-1} \Rightarrow e^{At} \approx VE\tilde{\Lambda}E^{-1}V^T$$

Global spectrum for Blasius flow

- Least stable eigenmodes equivalent using time-stepper and matrix solver
- Least stable branch is a global representation of Tollmien-Schlichting (TS) modes



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Global TS-waves

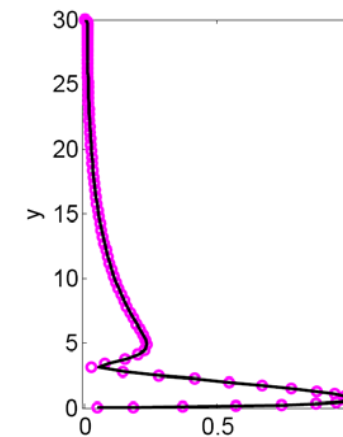
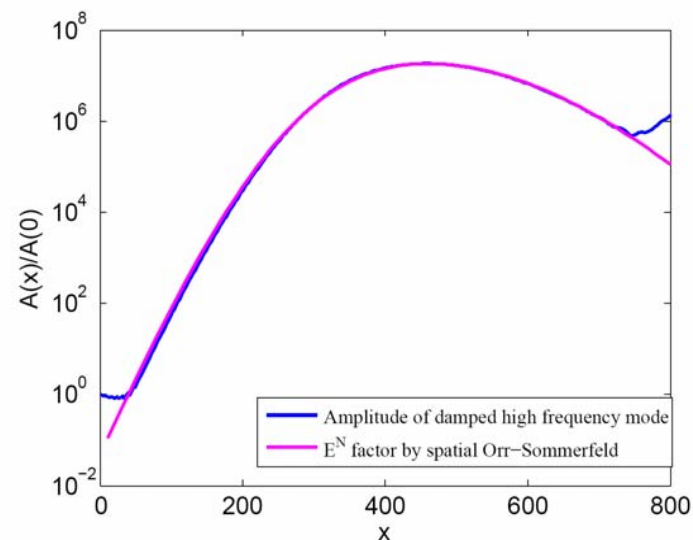
- Streamwise velocity of least damped TS-mode



- Envelope of global TS-mode identical to local spatial growth
- Shape functions of local and global modes identical



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Optimal disturbance growth

- Optimal growth from eigenvalues of $T^*(t)T(t)$ $T(t) = e^{At}$

$$G(t) = \max_{\|u_0\|=1} (u(t), u(t)) = \max_{\|u_0\|=1} (u_0, T^*(t)T(t)u_0)$$

$$T^*(t)T(t)u_0 = \lambda_E u_0$$

- Krylov sequence built by forward-adjoint iterations

$$T^*(t)T(t) \approx VHV^T = VE\Lambda_E E^{-1}V^T$$



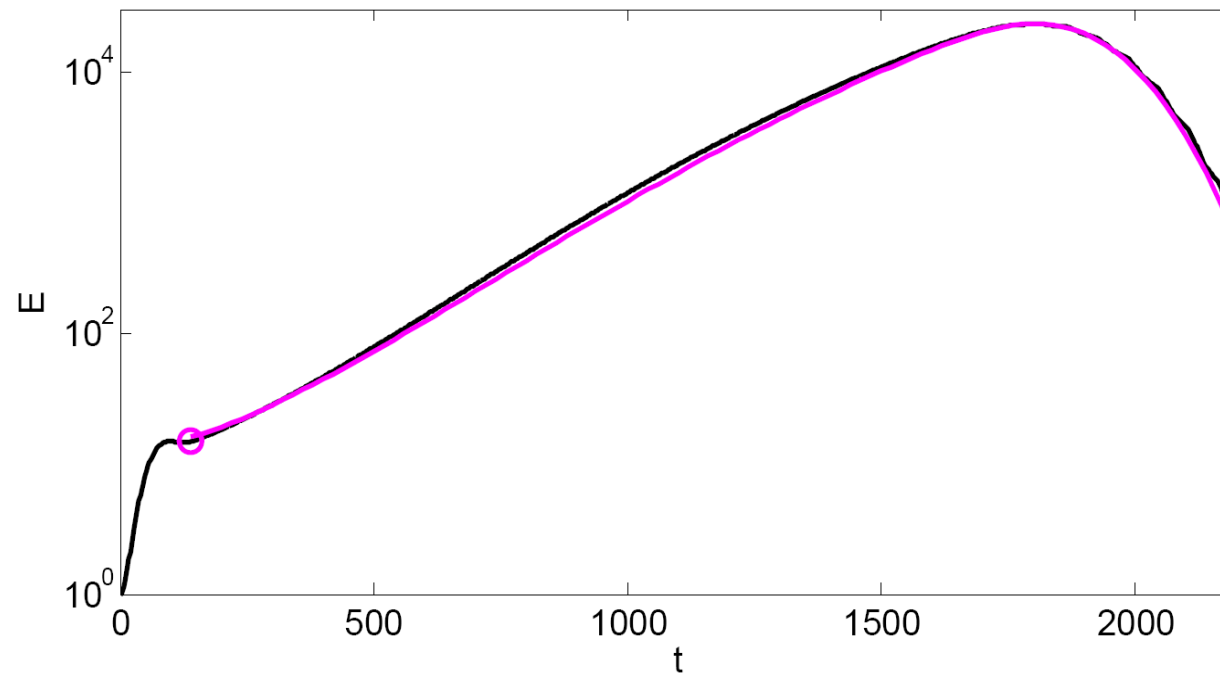
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Evolution of optimal disturbance in Blasius flow

- Full adjoint iterations (black)
sum of TS-branch modes only (magenta)
- Transient since disturbance propagates out of domain



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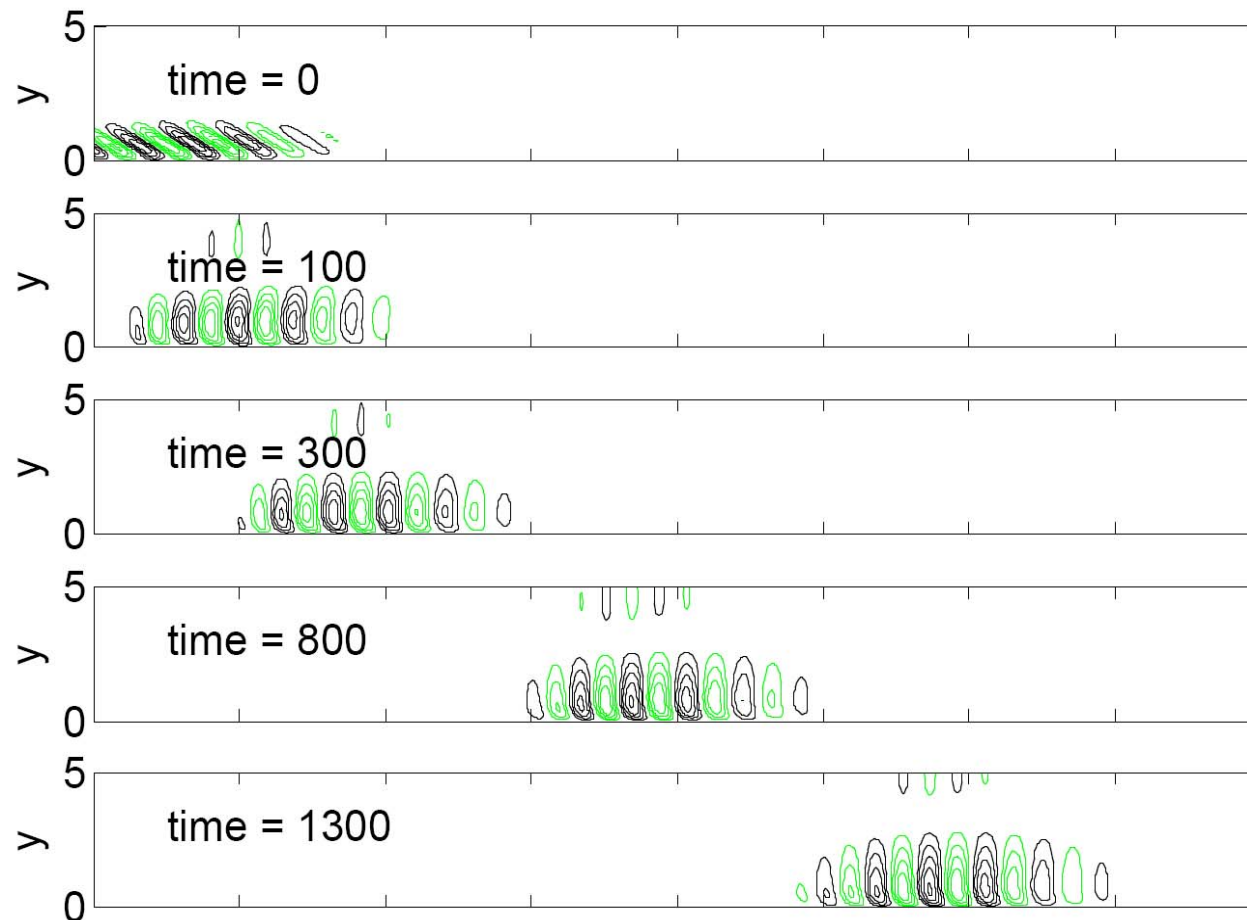


Snapshots of optimal disturbance evolution

- Initial disturbance leans against the shear raised up by Orr-mechanism into propagating TS-wavepacket



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The forced problem: input-output



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- Ginzburg-Landau example
- Input-output for 2D Blasius configuration
- Model reduction

Ginzburg-Landau example

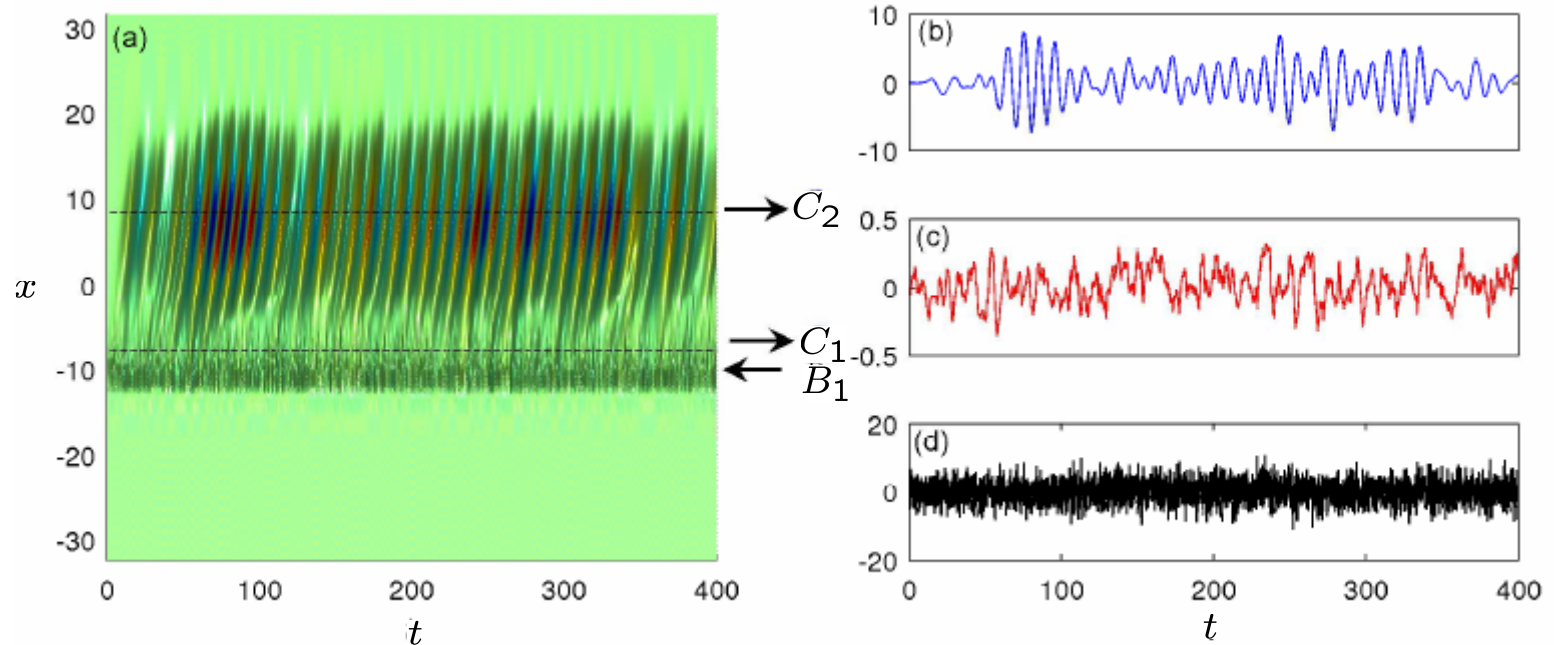
- Entire dynamics vs. input-output time signals

$$\begin{aligned} \frac{du}{dt} &= Au + Bf \\ y &= Cu \end{aligned}$$

$$y(t) = \int_0^t C e^{A(t-\tau)} B f(\tau) d\tau$$



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Input-output operators

- Past inputs to initial state: class of initial conditions possible to generate through chosen forcing

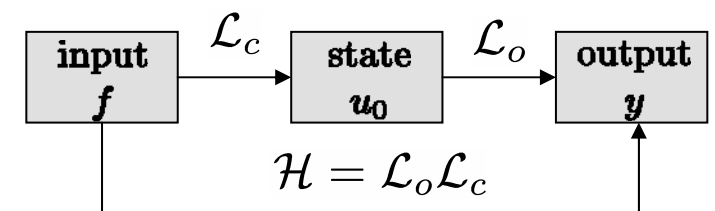
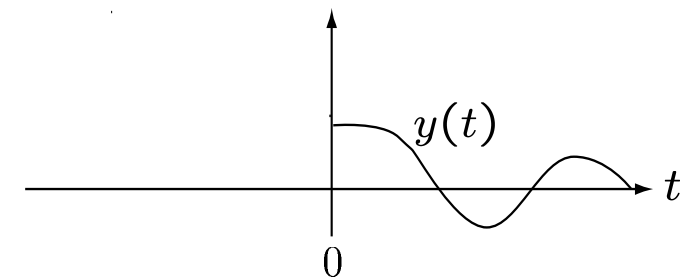
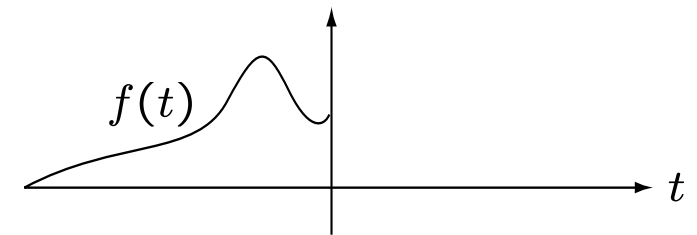
$$u_0 = \mathcal{L}_c f = \int_{-\infty}^0 e^{-A\tau} B f(\tau) d\tau$$

- Initial state to future outputs: possible outputs from initial condition

$$y(t) = \mathcal{L}_o u = C e^{At} u_0$$

- Past inputs to future outputs:

$$y = \mathcal{L}_o \mathcal{L}_c f(t) = \mathcal{H} f$$



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Most dangerous inputs and the largest outputs

- Eigenmodes of Hankel operator – balanced modes

$$\max_{\|f(t)\|=1} \|y(t)\|^2 = \max_{\|f(t)\|=1} (\mathcal{H}f, \mathcal{H}f) = \max_{\|f(t)\|=1} (f, \underbrace{\mathcal{L}_c^* \mathcal{L}_o^* \mathcal{L}_o \mathcal{L}_c}_{\mathcal{H}^* \mathcal{H}} f)$$

$$\underbrace{\mathcal{L}_c \mathcal{L}_c^*}_P \underbrace{\mathcal{L}_o^* \mathcal{L}_o}_Q \underbrace{\mathcal{L}_c f}_u = \sigma^2 \underbrace{\mathcal{L}_c f}_u$$



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- Controllability Gramian

$$P = \mathcal{L}_c \mathcal{L}_c^* = \int_0^{\infty} e^{A\tau} B B^T e^{A^T \tau} d\tau$$

- Observability Gramian

$$Q = \mathcal{L}_o^* \mathcal{L}_o = \int_{-\infty}^0 e^{-A^T \tau} C^T C e^{-A\tau} d\tau$$

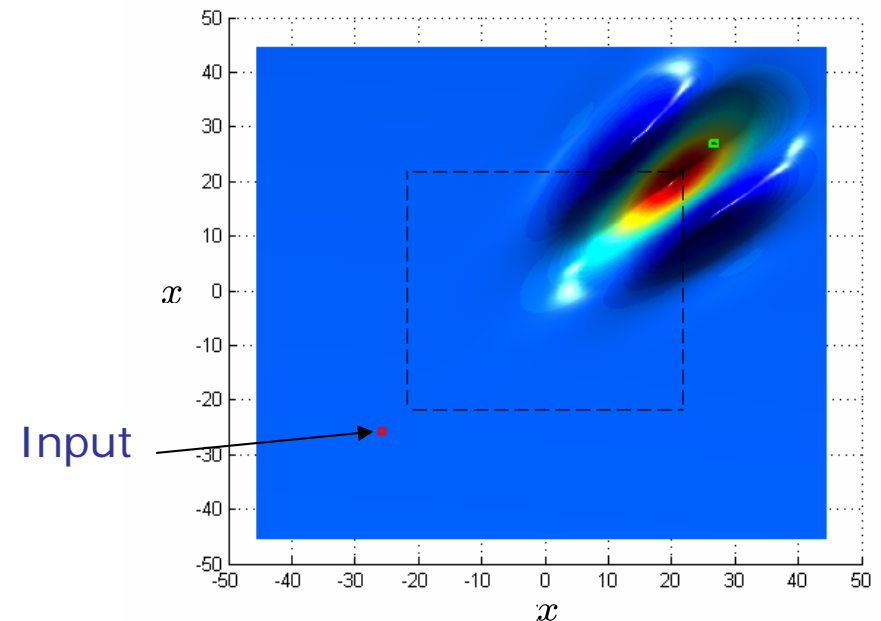
Controllability Gramian for GL-equation

$$P = \mathcal{L}_c \mathcal{L}_c^* = \int_0^\infty e^{A\tau} B B^T e^{A^T \tau} d\tau \approx X X^T, \quad X = [e^{At_1} B, \dots, e^{At_m} B]$$



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- Correlation of actuator impulse response in forward solution
- POD modes: $\text{eig}\{X X^T\}$
- Ranks states most easily influenced by input
- Provides a means to measure controllability



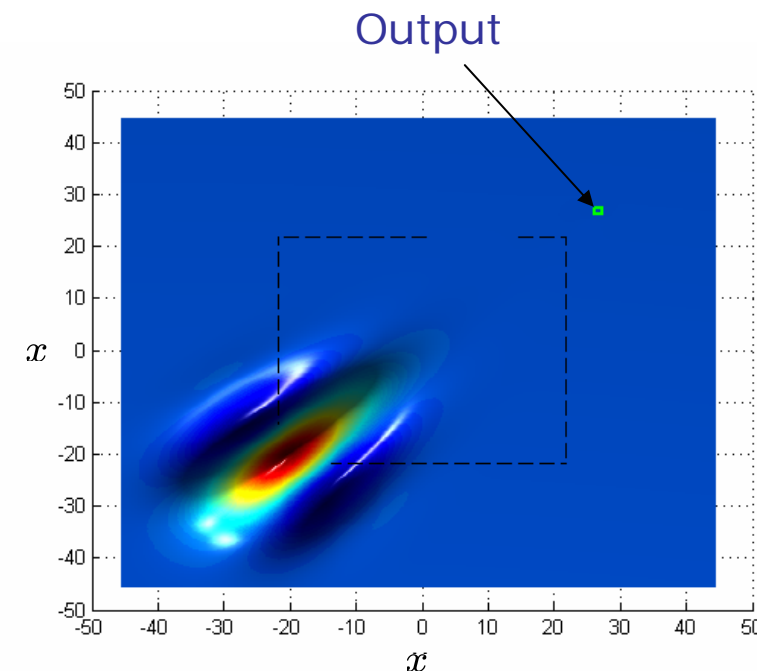
Observability Gramian for GL-equation

$$Q = \mathcal{L}_o^* \mathcal{L}_o = \int_{-\infty}^0 e^{-A^T \tau} C^T C e^{-A \tau} d\tau \approx YY^T, \quad Y = [e^{-A^T t_m} C^T, \dots, e^{-A^T t_1} C^T]$$



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- Correlation of sensor impulse response in adjoint solution
- Adjoint POD modes: $\text{eig}\{YY^T\}$
- Ranks states most easily sensed by output
- Provides a means to measure observability



Balanced modes: eigenvalues of the Hankel operator



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- Combine snapshots of direct and adjoint simulation

$$\underbrace{\mathcal{L}_c \mathcal{L}_c^*}_P \underbrace{\mathcal{L}_o^* \mathcal{L}_o}_Q u = \sigma^2 u \quad \Rightarrow \quad \underbrace{X X^T Y Y^T}_{n \times n} u = \sigma^2 u$$

- Expand modes in snapshots to obtain smaller eigenvalue problem

$$u = X a \quad \Rightarrow \quad X \underbrace{(X Y Y^T X)}_{m \times m} a - \sigma^2 a = 0$$

Snapshots of direct and adjoint solution in Blasius flow

Direct simulation:

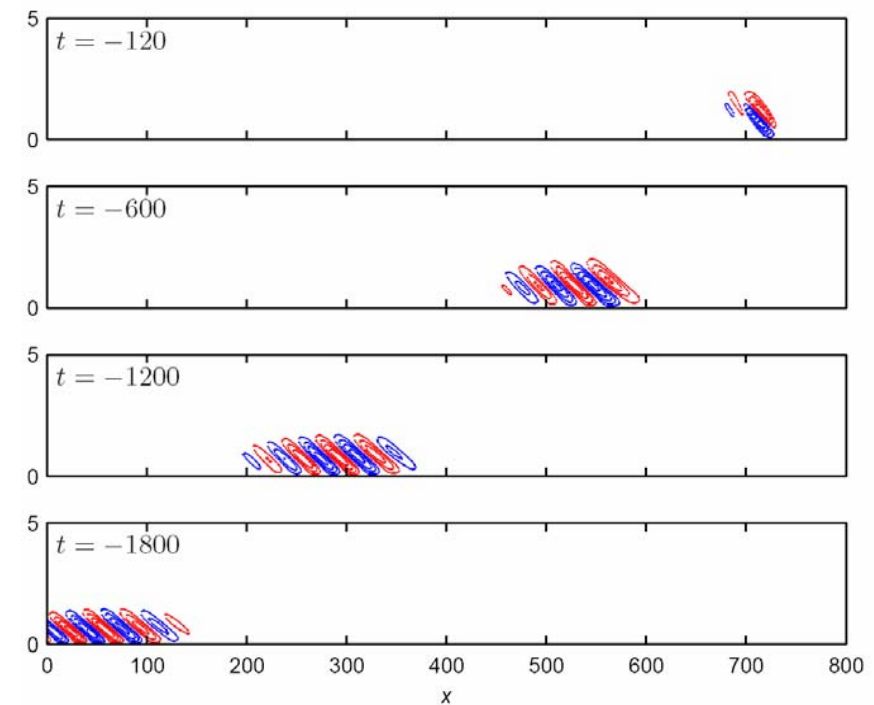
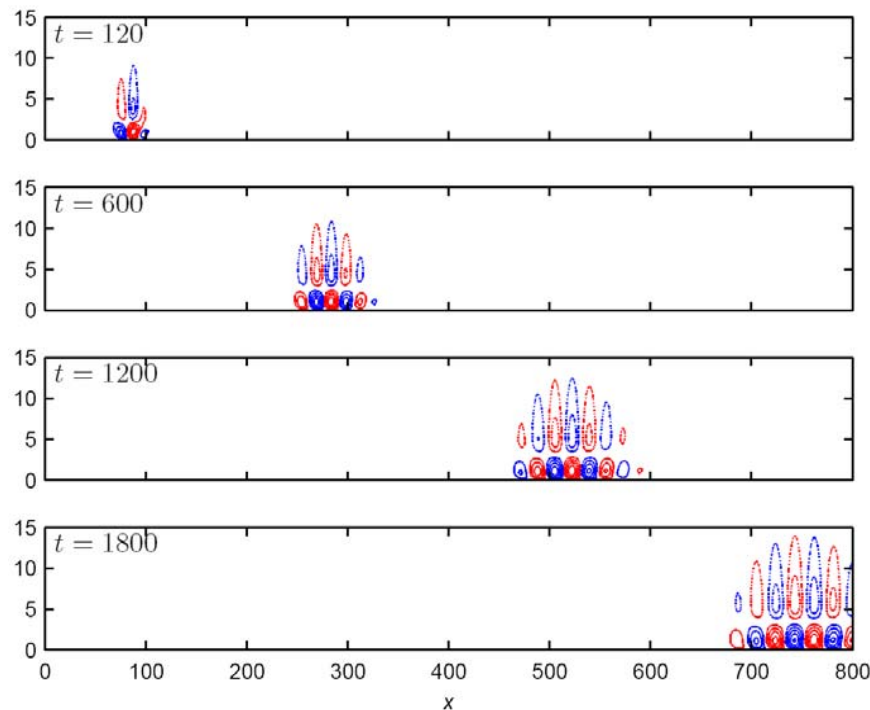
$$X = [u(t_1) \dots u(t_m)]$$

Adjoint simulation:

$$Y = [p(t_m) \dots p(t_1)]$$



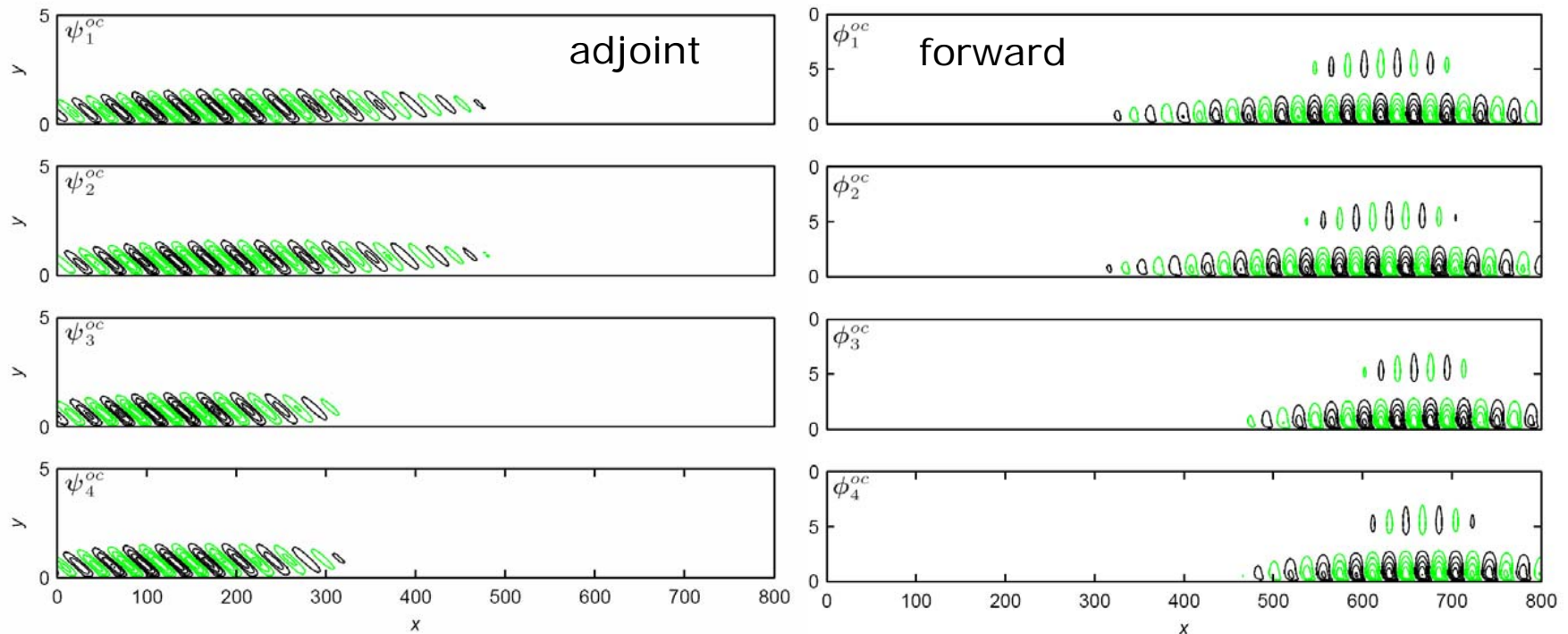
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Balanced modes for Blasius flow



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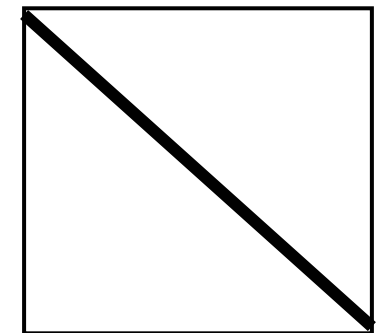
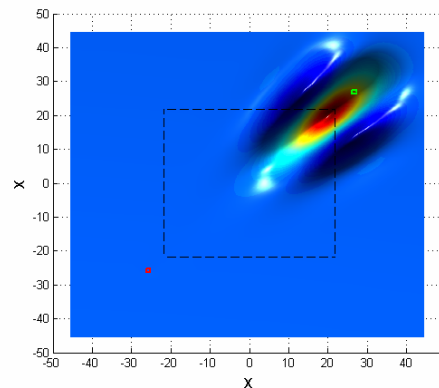
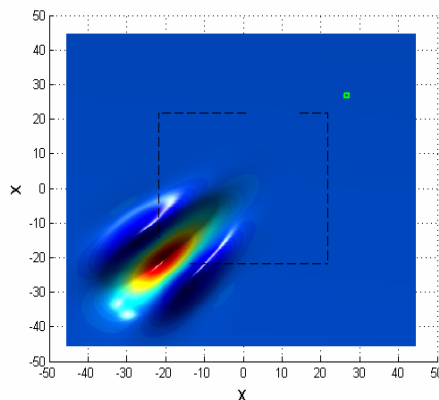
Properties of balanced modes

- Largest outputs possible to excite with chosen forcing
- Balanced modes diagonalize observability Gramian
- Adjoint balanced modes diagonalize controllability Gramian
- Ginzburg-Landau example revisited



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$$\hat{Q} = \hat{P} = \text{diag}\{\sigma_1, \dots, \sigma_n\}$$



Model reduction

- Project dynamics on balanced modes using their biorthogonal adjoints
- Reduced representation of input-output relation, useful in control design



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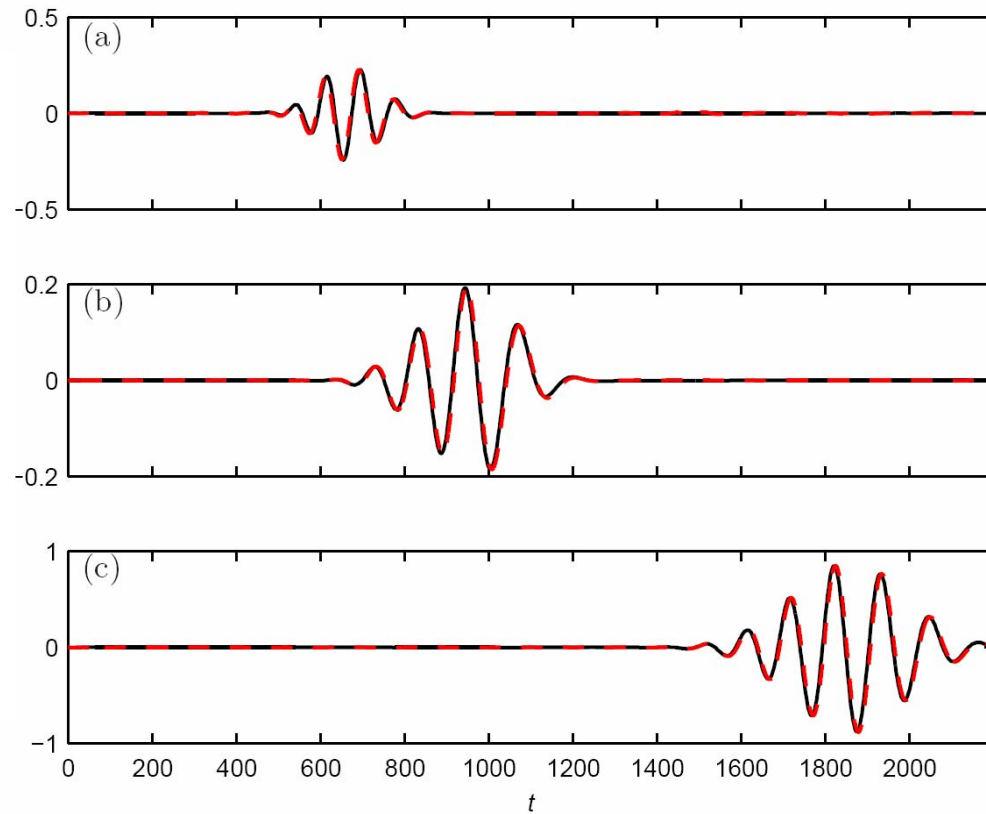
$$\frac{d\kappa}{dt} = A\kappa + Bf \quad u \approx \sum_{j=1}^r \kappa_j(t) u_j$$
$$y = C\kappa$$

$$A_{ij} = (\psi_i, A\phi_i)$$
$$B_{1j} = (\psi_j, B_1)$$
$$C_{1j} = C_1\phi_j$$

Impulse response $f = \delta(0)$



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Disturbance → Sensor

Actuator → Objective

Disturbance → Objective

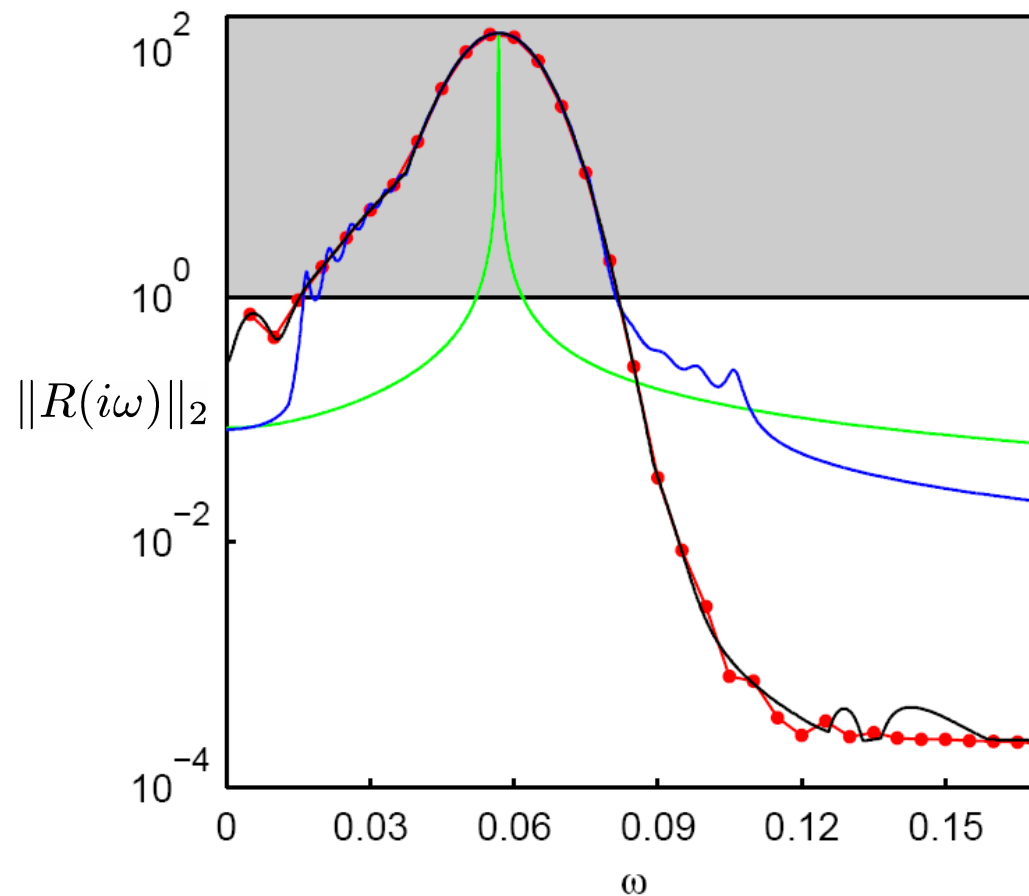
DNS: $n=10^5$ $y = Ce^{A\tau}Bf$
 ROM: $m=50$ $y \approx Ce^{A\tau}Bf,$

Frequency response

$$f = e^{st}$$

$$y = \underbrace{\int_0^\infty C e^{A(t-\tau)} B d\tau}_{R(s)} e^{st}$$

From all inputs to all outputs



- DNS: $n=10^5$
- ROM: $m=80$
- $m=50$
- $m=2$



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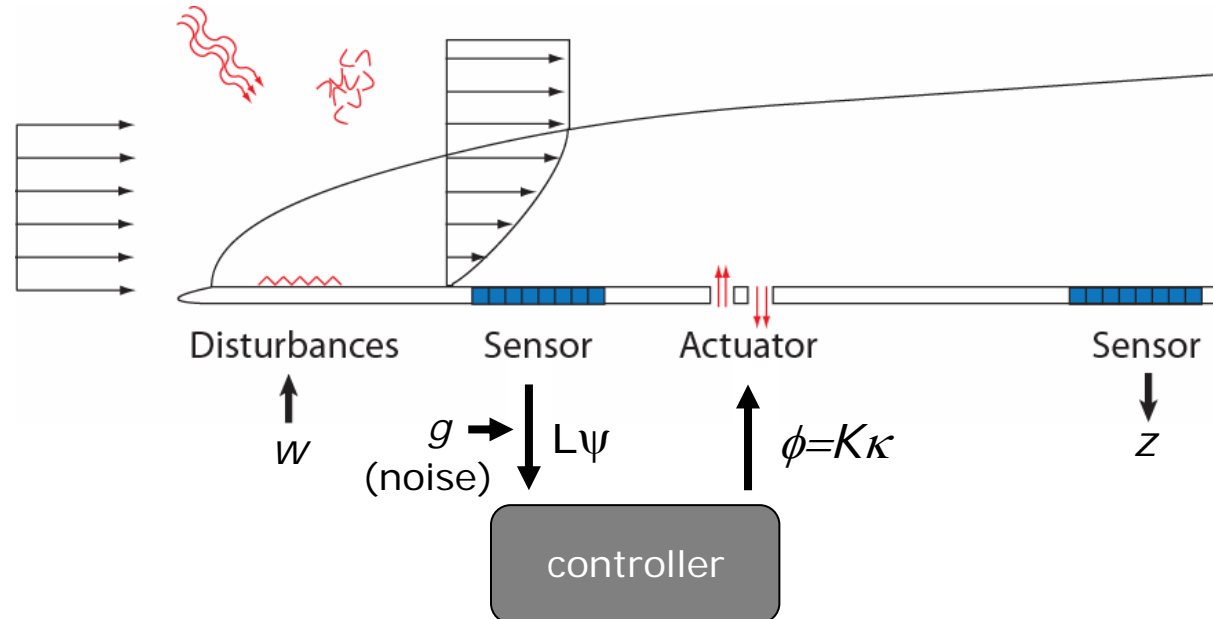
Feedback control



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- LQG control design using reduced order model
- Blasius flow example

Optimal Feedback Control – LQG



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Find an optimal control signal $\phi(t)$ based on the measurements $\psi(t)$ such that in the presence of external disturbances $w(t)$ and measurement noise $g(t)$ the output $z(t)$ is minimized.

→ Solution: LQG/H2

LQG controller formulation with DNS

- Apply in Navier-Stokes simulation

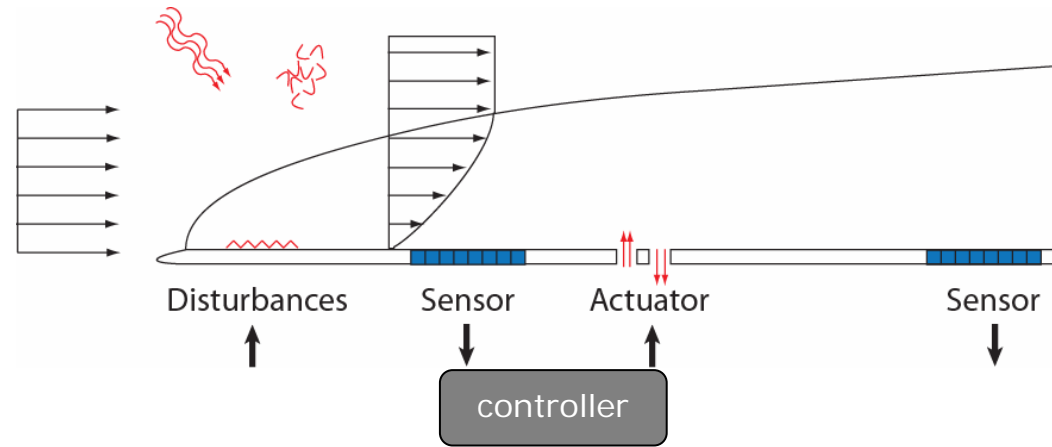


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$$\begin{aligned}\frac{du}{dt} &= NS(u) + B_1 w + B_2 \phi \\ \psi &= C_1 u\end{aligned}$$

$$\begin{aligned}\frac{d\kappa_e}{dt} &= (A + B_2 K + L C_2) \kappa_e - L \psi \\ \phi &= K \kappa_e\end{aligned}$$

Performance of controlled system



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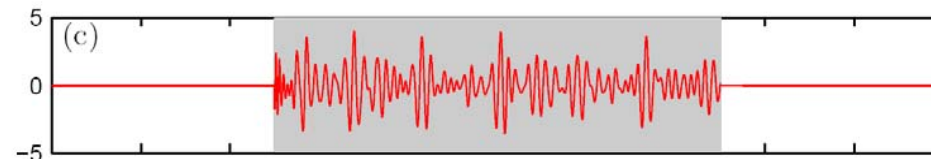
Noise



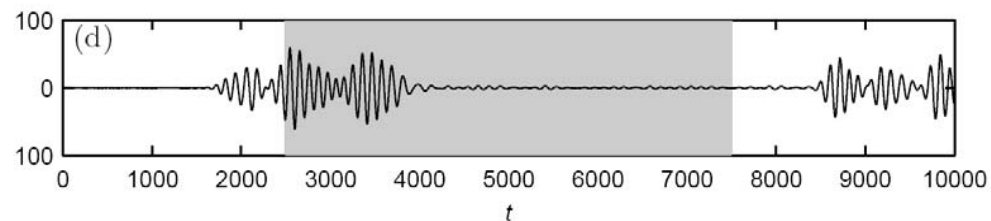
Sensor



Actuator



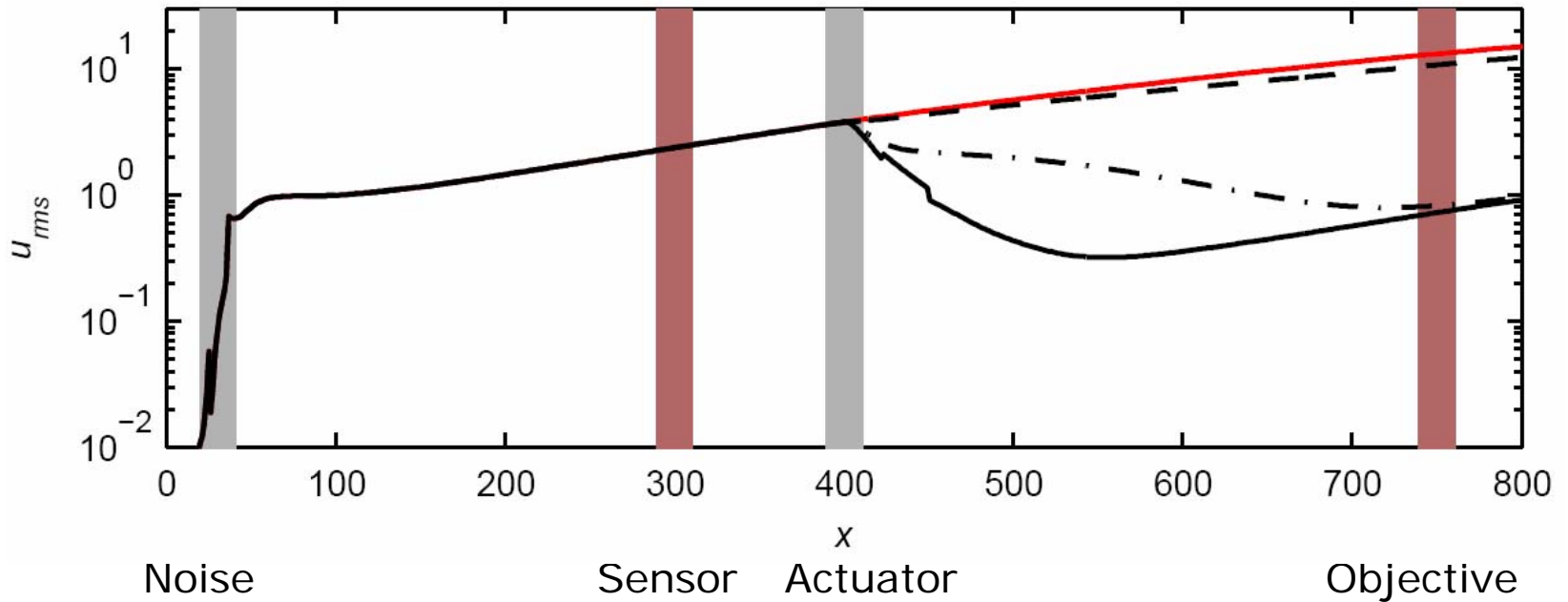
Objective



Performance of controlled system



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- Control off
- Cheap Control
- Intermediate control
- Expensive Control

Conclusions



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- Complex stability/control problems solved using Krylov/Arnoldi methods based on snapshots of forward and adjoint Navier-Stokes solutions
- Optimal disturbance evolution brought out Orr-mechanism and propagating TS-wave packet automatically
- Balanced modes give low order models preserving input-output relationship between sensors and actuators
- Feedback control of Blasius flow
Reduced order models with balanced modes used in LQG control
Controller based on small number of modes works well in DNS
- Outlook: incorporate realistic sensors and actuators in 3D problem and test controllers experimentally

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