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Performance of reduced-order models of fluid systems

Exemplified on Blasius flow and Ginzburg-Landau equation



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Motivation



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- Navier-Stokes equations too complex to manipulate
 - $n > 10^6$ degrees of freedom
- **Model reduction**: approximate the high-dimensional system with a low-dimensional system
 - $m \ll 10$ degrees of freedom
- Model reduction is problem dependent:
 - Each application has a suitable low-dimensional model that captures the essential feature of its dynamics.
- In **control design** one describes the relation between inputs-outputs in terms of transfer functions.
 - How can we find a low-dimensional model that describes the input-output behaviour?
 - Answer: Balanced truncation

What is relevant for capturing the input-output behaviour?



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- For control design we must analyze the mappings:
 - Perturbations \longrightarrow Objective function
 - Perturbations \longrightarrow Sensors
 - Actuators \longrightarrow Objective function
 - Actuators \longrightarrow Sensors
- Write linearized Navier-Stokes in state-space formulation:

$$\begin{aligned}\dot{q} &= Aq + Bu & q(0) &= q_0 \\ y &= Cq\end{aligned}$$

- The formal solution:

$$y = Ce^{At}q_0 + C \int_0^t e^{A(t-\tau)}u(x, \tau)d\tau$$

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Find the components of the state that are easily excited by input

What is relevant for capturing the input-output behaviour?



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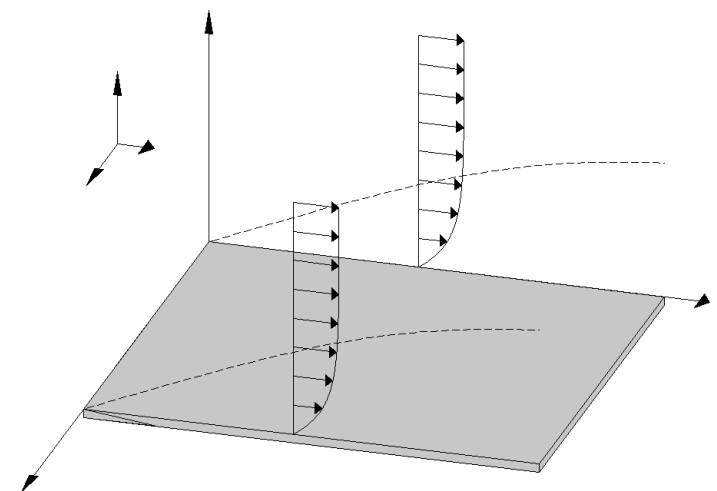
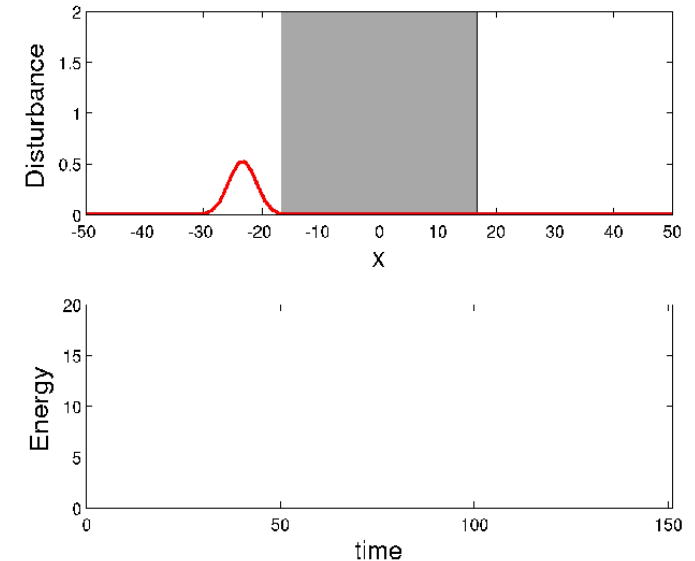
Find the components of state
with large influence on
output

Study cases – Choice of A



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- 1D-case: Linear Ginzburg-Landau equation:
 - Convective and absolute instability
 - Transient growth due to non-normality of A
- 2D-case: Flat-plate boundary Layer -Blasius:
 - $Re_{\delta}^* = 500$
 - $n = 2(512 \times 96)^{1/4} \approx 10000$
 - Spectral Code



States easily excited by input

For unit impulse input $u = \delta(t)$ the state is given by,

$$q(t) = e^{At} B$$

we can measure the "size" of the state with the $n \times n$ matrix

$$X = \int_0^{\infty} q(t)q(t)^H dt = \int_0^{\infty} e^{A\tau} B B^H e^{A^H \tau} d\tau$$

This matrix is called the **controllability Gramian**.

- Gives a measure of the controllability of the components of a state
- Let T diagonalize X :

$$T X T^T = \text{diag}\{\sigma_1, \dots, \sigma_n\}$$

- Eigenvalue σ_i measures how much the state $T_i q$ is excited by the input
- Cannot neglect states corresponding to small σ_i , these state may have a large influence on the output!!

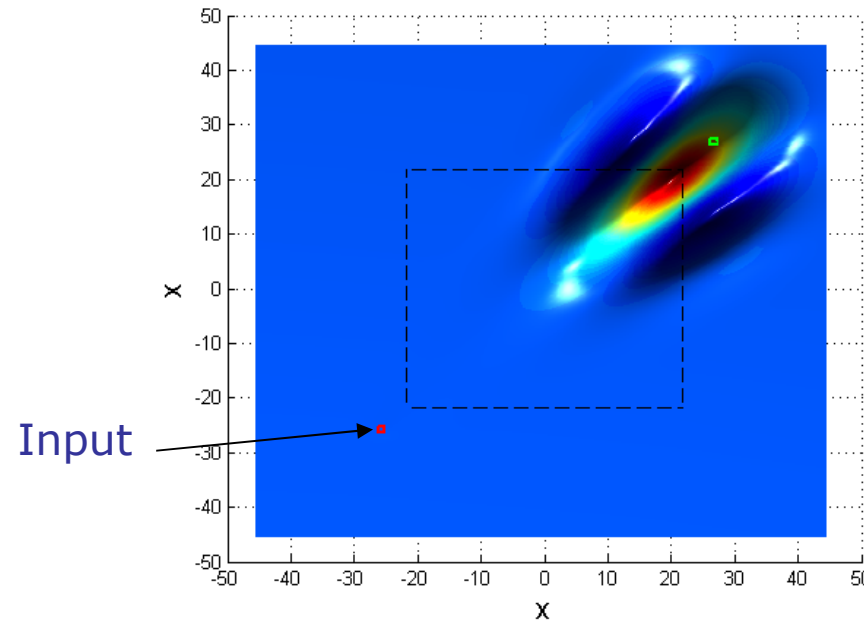


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Controllability of Ginzburg-Landau



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$$X = \underbrace{\int_0^{\infty} q(t)q(t)^H dt}_{\text{State Covariance matrix}} = \underbrace{\int_0^t e^{A\tau} B B^H e^{A^H \tau} d\tau}_{\text{Controllability Gramian}}$$

State Covariance matrix

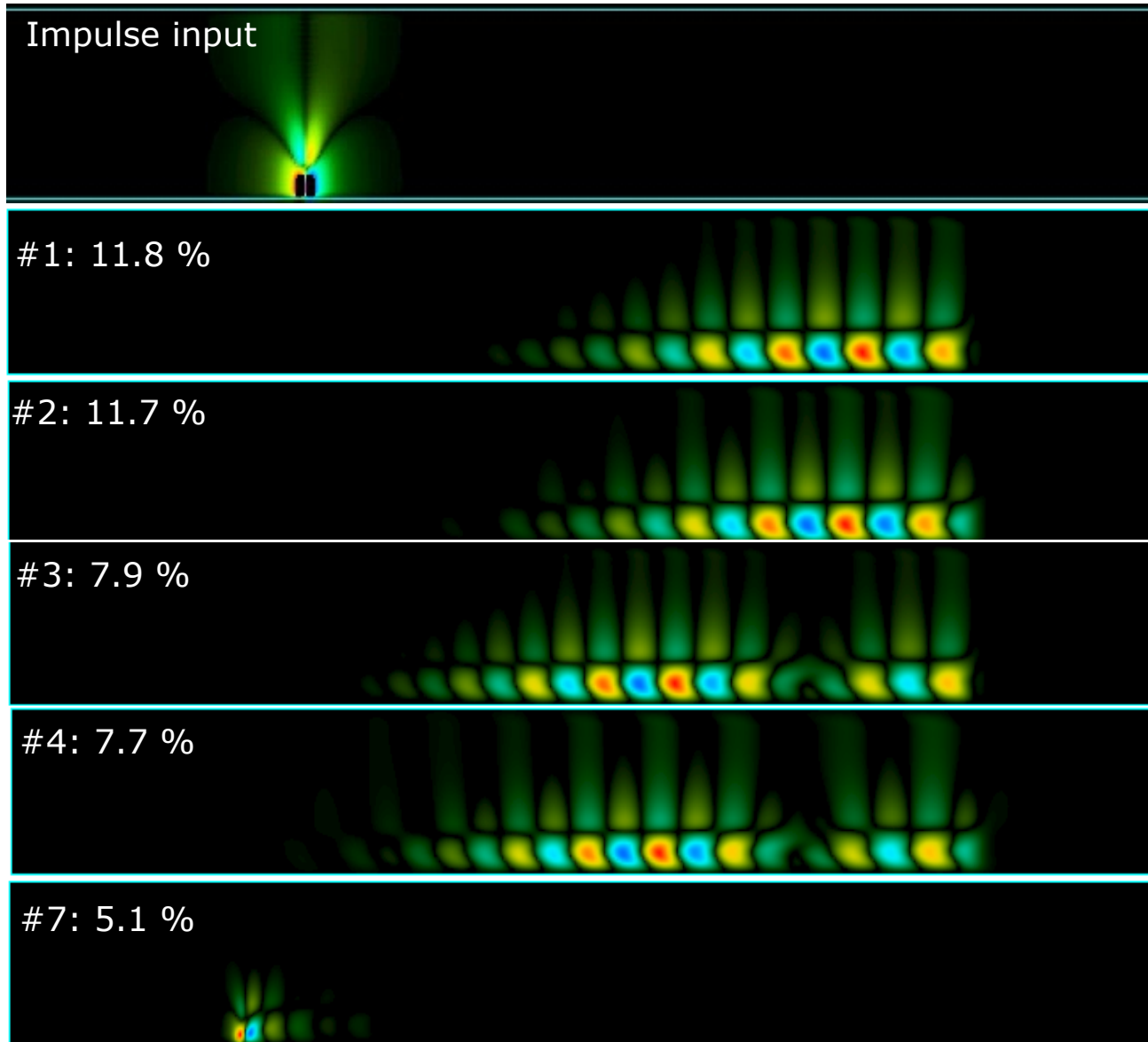
Controllability Gramian

- T are POD modes!
- Snapshot method

Controllable modes of Blasius flow



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400 snapshots

States with large influence on output

How much does a state contribute to the output?

Consider an initial state q_0 , the state is given by

$$y(t) = Ce^{At}q_0$$

We can measure the size of the output by,

$$\|y\|^2 = q_0^H \underbrace{\int_0^\infty e^{A^H t} C^H C e^{At} dt}_Y q_0 = q_0^H Y q_0$$

where Y , is called the **observability Gramian**.

- Measures how much energy a component of q_0 is transferred to output
- Let T^{-1} diagonalize Y :

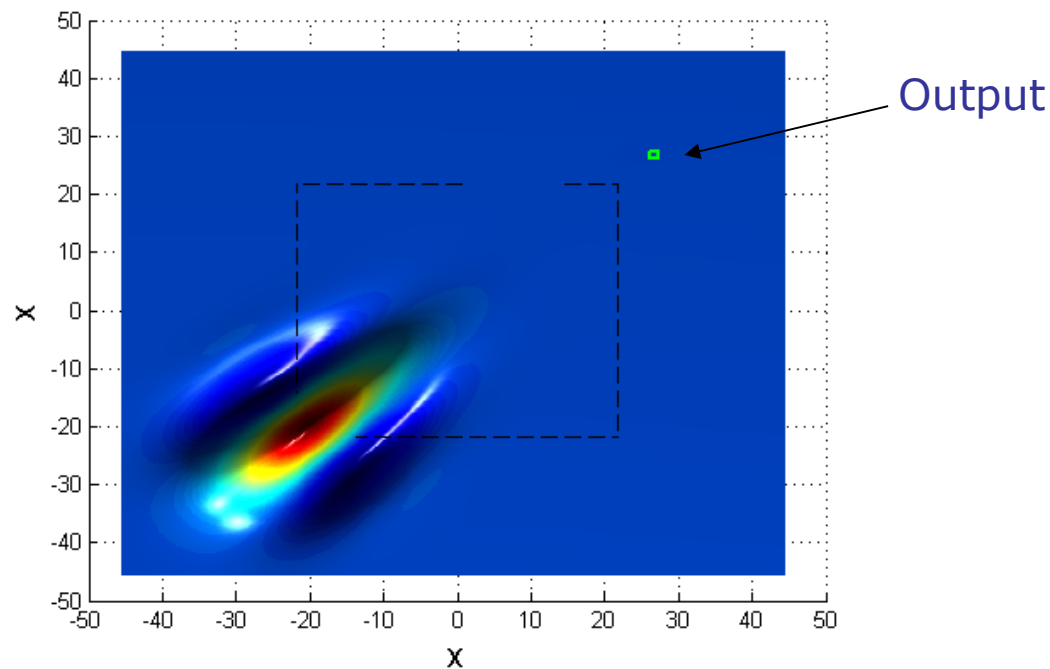
$$T^{-T} Y T^{-1} = \text{diag}\{\sigma_1, \dots, \sigma_n\}$$

- If σ_i is zero, the output cannot sense $T^{-1}_i q_0$.



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Observability of Ginzburg-Landau



$$\dot{q}^+ = A^H q^+ + C^H v$$

$$q^+ = e^{A^H t} C^H$$

$$Y = \underbrace{\int_0^\infty e^{A^H t} C^H C e^{A t} dt}_{\text{Observability Gramain}}$$

Observability
Gramain

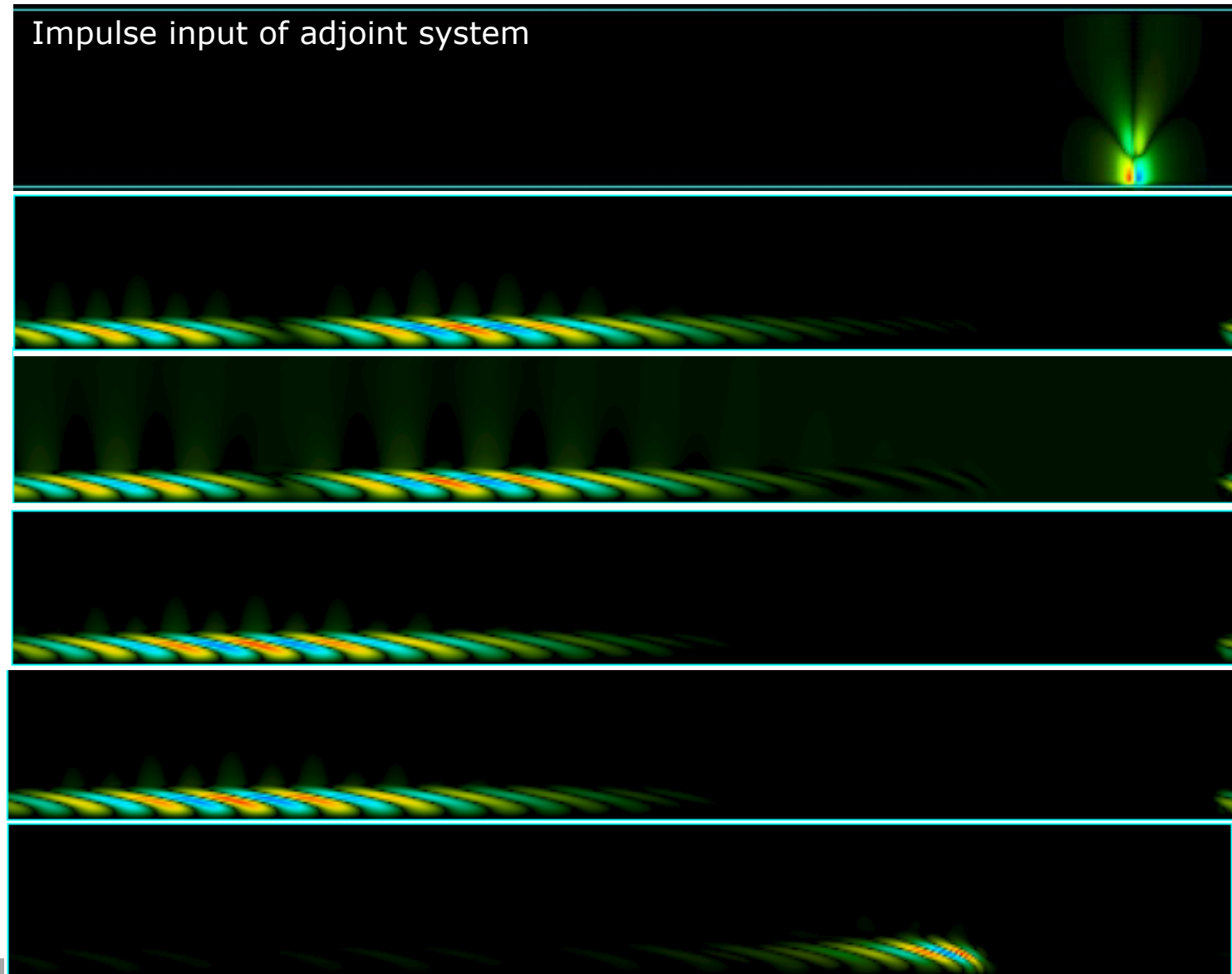
Covariance matrix of adjoint state

- T^{-1} are adjoint POD modes!
- Snapshot method

Observable modes of Blasius flow

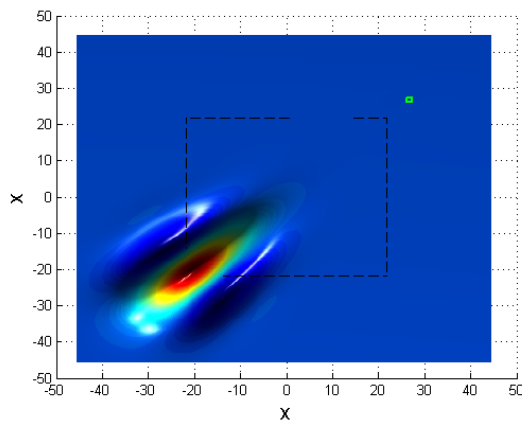


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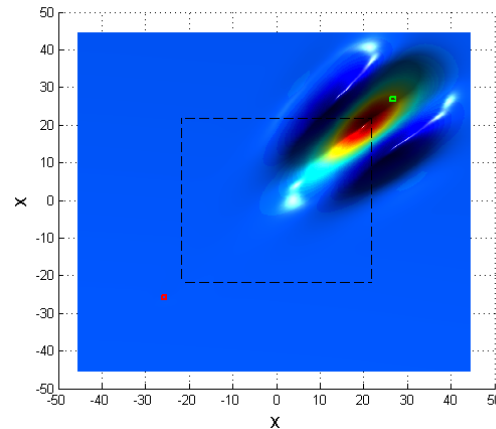


Balanced Realization

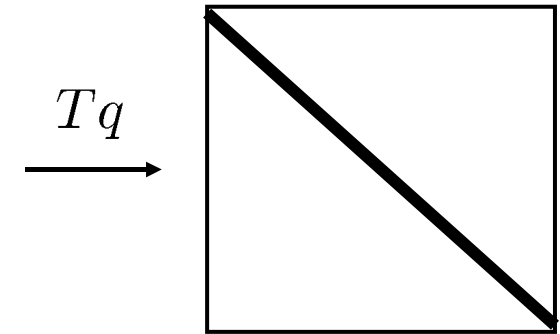
Unobservable and uncontrollable states do no influence the input-output behaviour of a system



Observable states



Controllable states



Balanced system

Choose T and T^{-1} so that the two Gramians become equal and diagonal:

$$T X T^T = T^{-T} Y T^{-1} = \text{diag}\{\sigma_1, \dots, \sigma_m\}$$

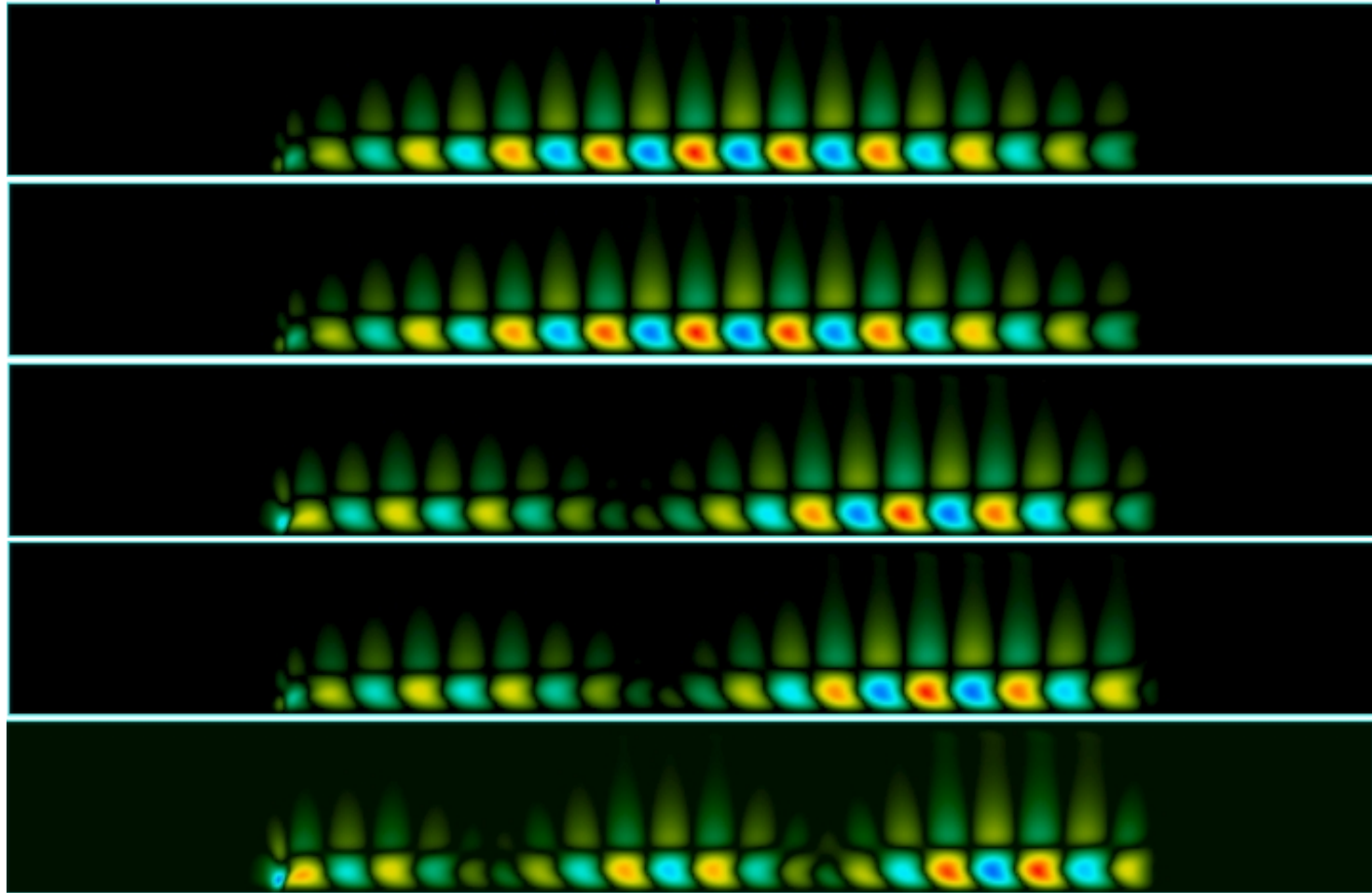
The columns of T contain the **Balanced modes**.

Balanced modes of Blasius flow

u-component



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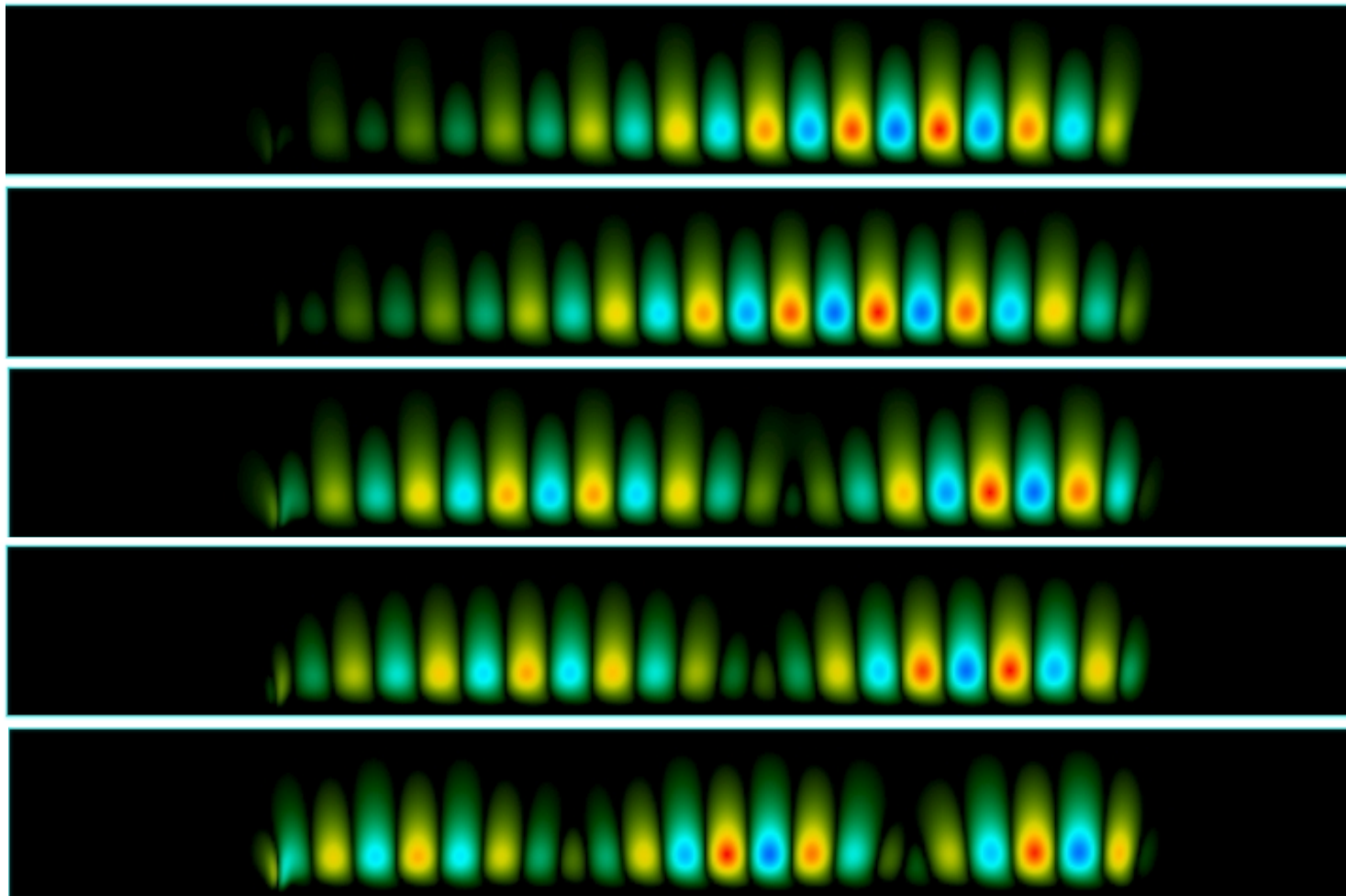


Balanced modes of Blasius flow

v-component



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Balanced truncation – model reduction



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1. Compute balanced modes from snapshots

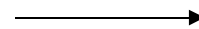
2. Change coordinates:

$$\hat{q} = Tq$$

3. Truncate the least controllable and observable states

$$\begin{aligned} \dot{q} &= Aq + Bu \\ y &= Cq \end{aligned}$$

Huge



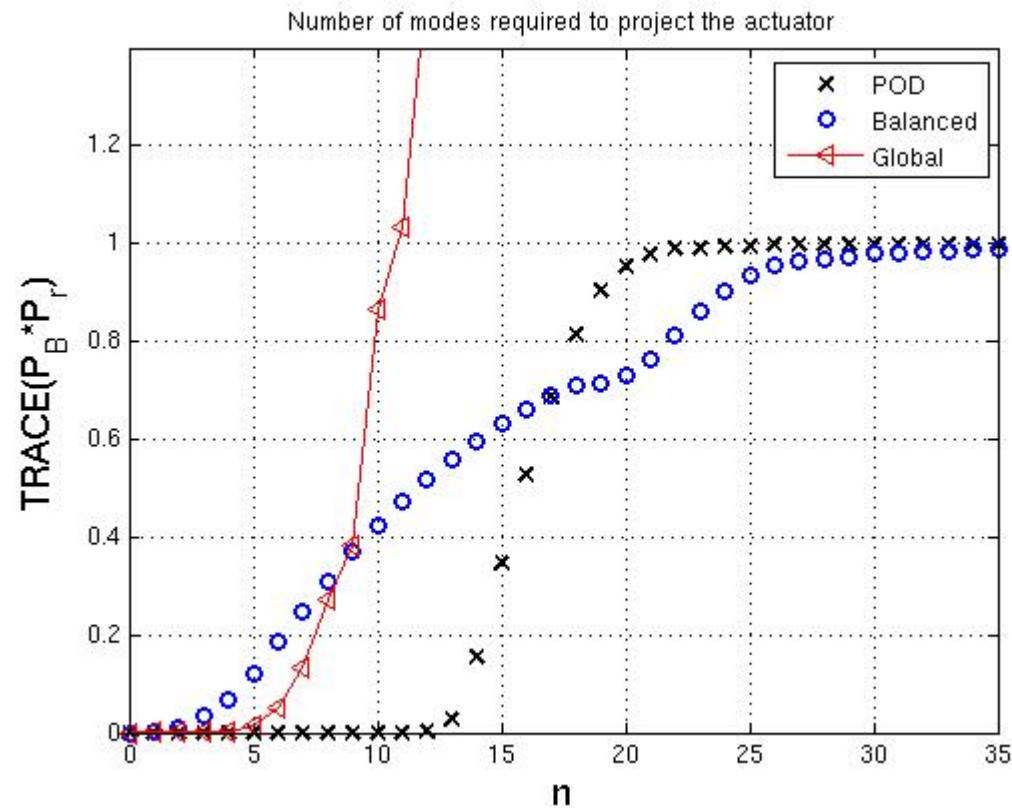
$$\begin{aligned} \dot{\hat{q}} &= TAT^{-1}\hat{q} + TBu \\ y &= CT^{-1}q \end{aligned}$$

Small

Projection of actuator



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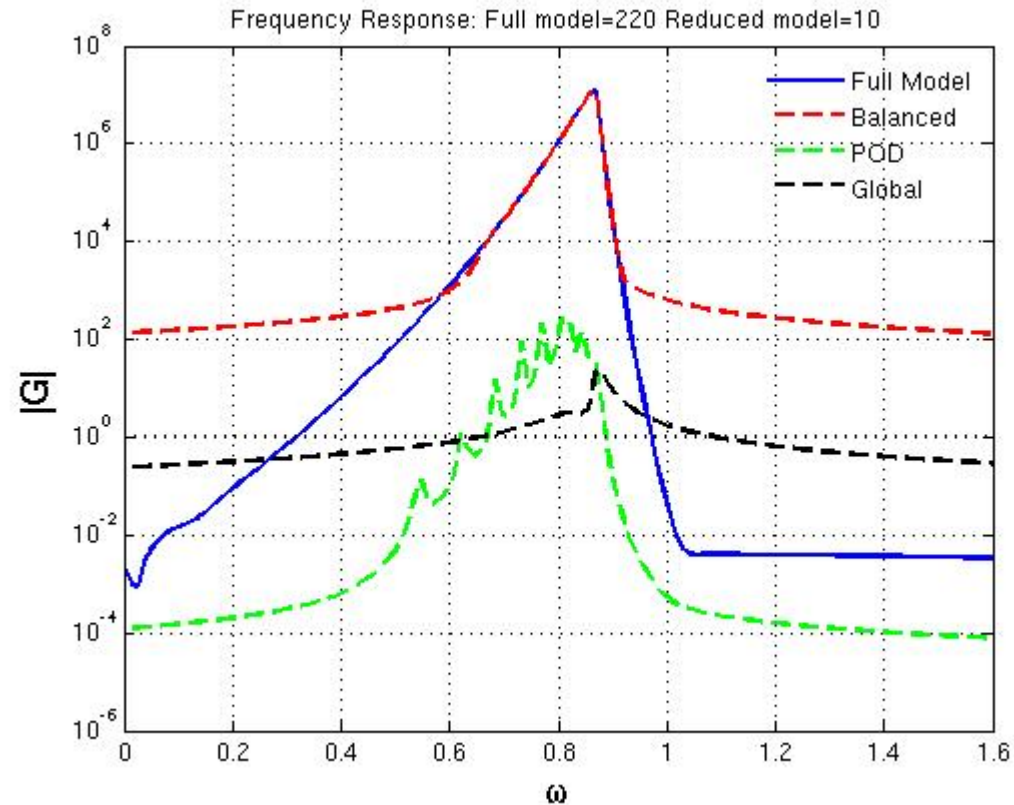


- Number of modes required to project an actuator
- Actuator placed upstream, close to Branch I
- Reduced-order model of order 13, based POD captures nothing!

White noise -> Sensor



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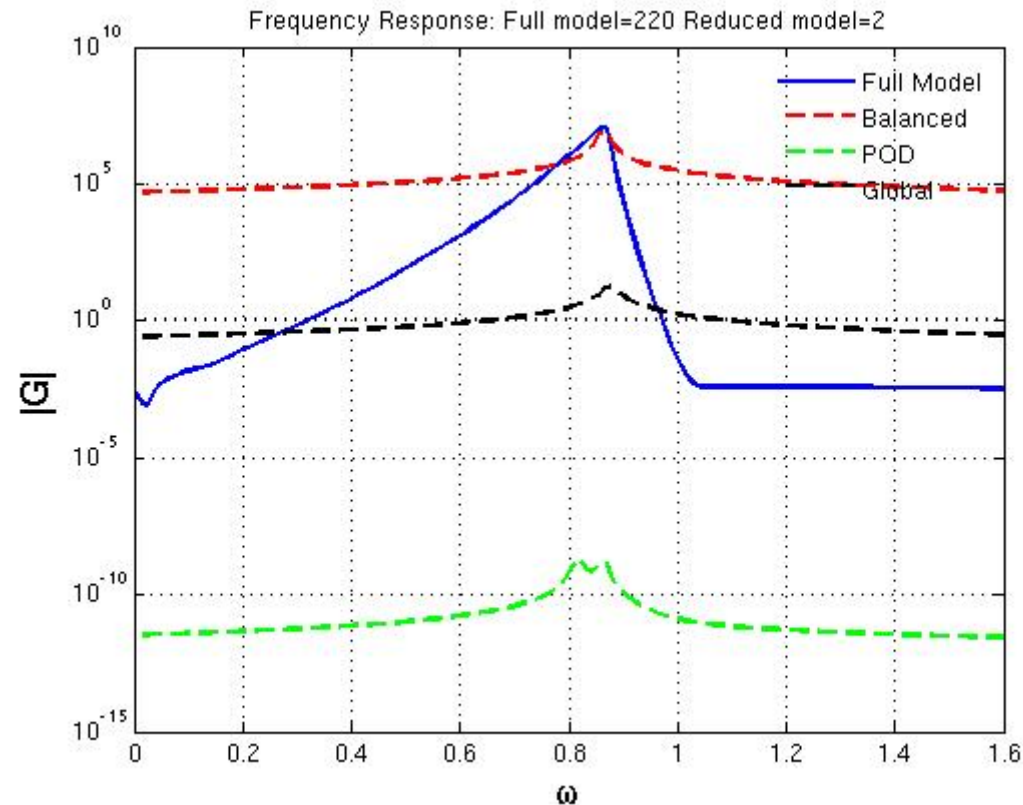


- B close to branch I and C close to branch II
- Full-order: $n=220$
- Reduced-order: $m=30$
- Balanced modes perform best

White noise -> Sensor



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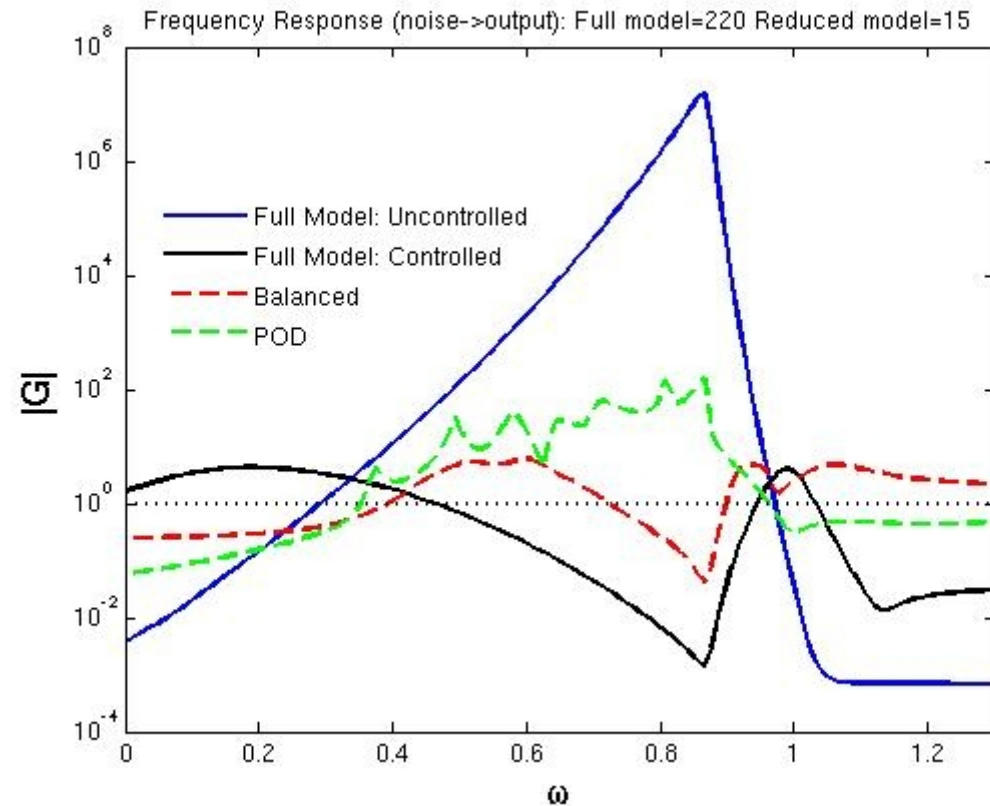


- B close to branch I and C close to branch II
- Full-order: $n=220$
- Reduced-order: $m=2$
- Balanced modes perform best again

LQG -Feedback control (1)



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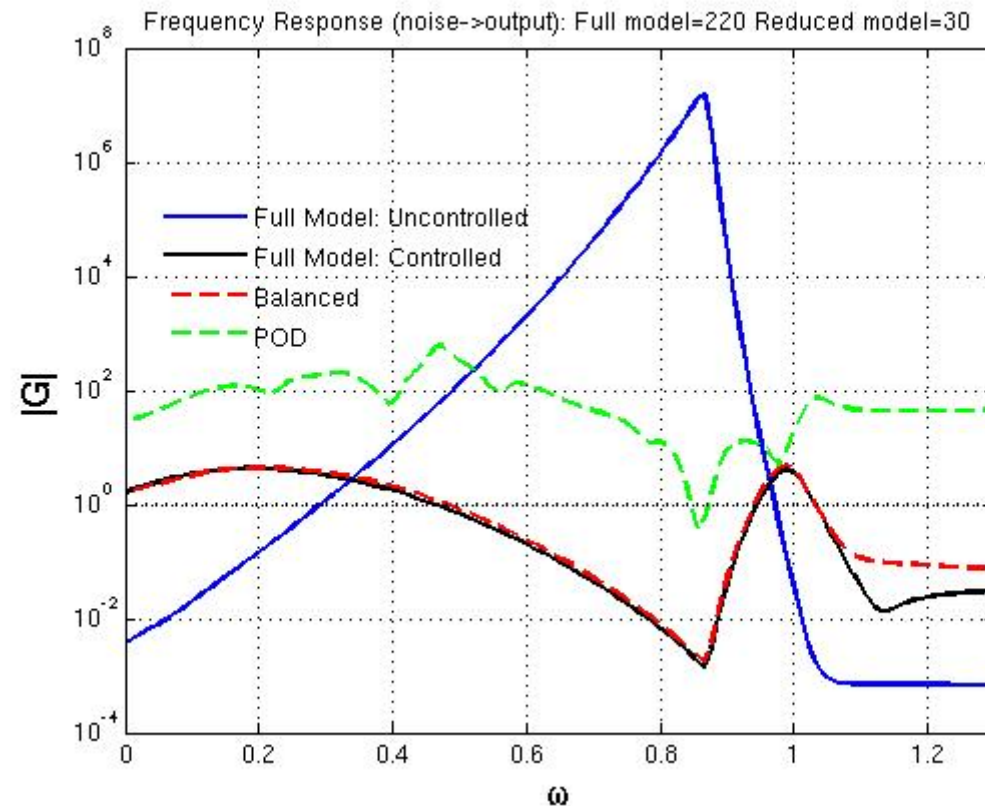


- Inputs & Outputs:
 - Disturbance: White noise
 - Actuator: Gaussian Volume forcing
 - Sensor: Gaussian function
- Full-order: $n=220$, Reduced-order: $m=15$

LQG -Feedback control (2)



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- Inputs & Outputs:
 - Disturbance: White noise
 - Actuator: Gaussian Volume forcing
 - Sensor: Gaussian function
- Full-order: $n=220$, Reduced-order: $m=30$

Conclusions



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- Balanced modes perform better than POD modes approximating the large-scale model for the examples studied here.
- Balanced truncation is extendable to large-scale systems, using the snapshot method. Cost similar to computing POD modes.
- These mode capture the relation between input signals and output signals, and are hence suitable for control design.

Extra slides

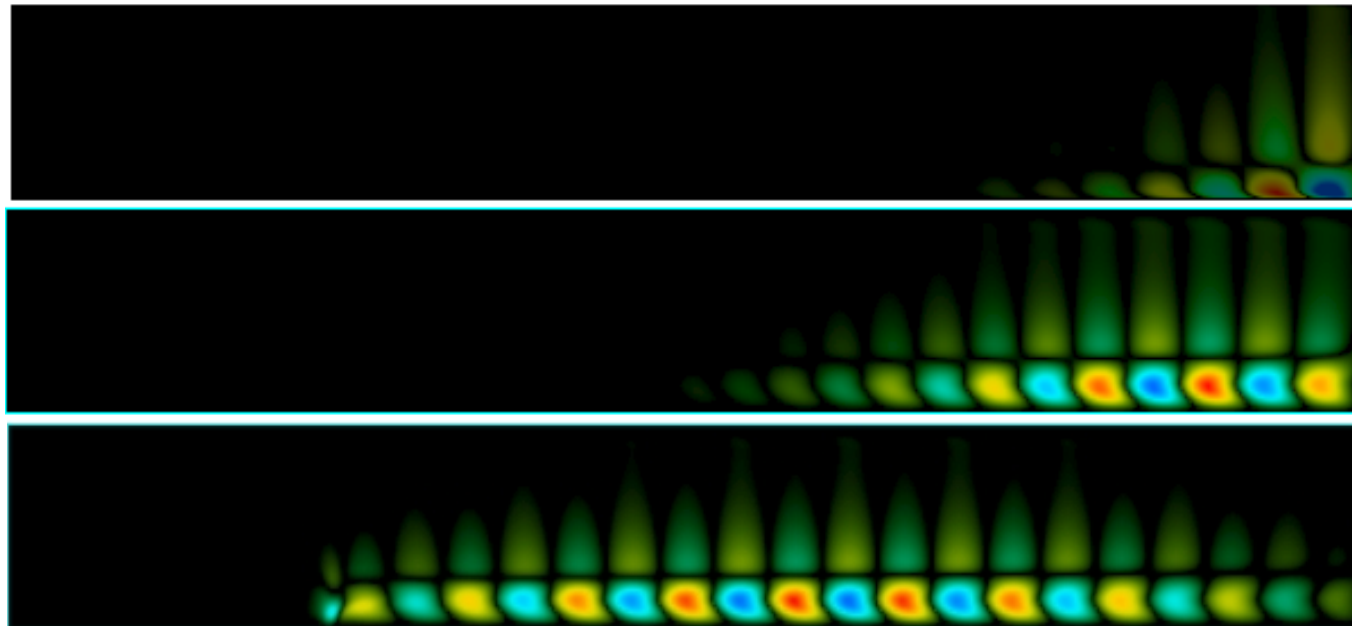


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Blasius Modes



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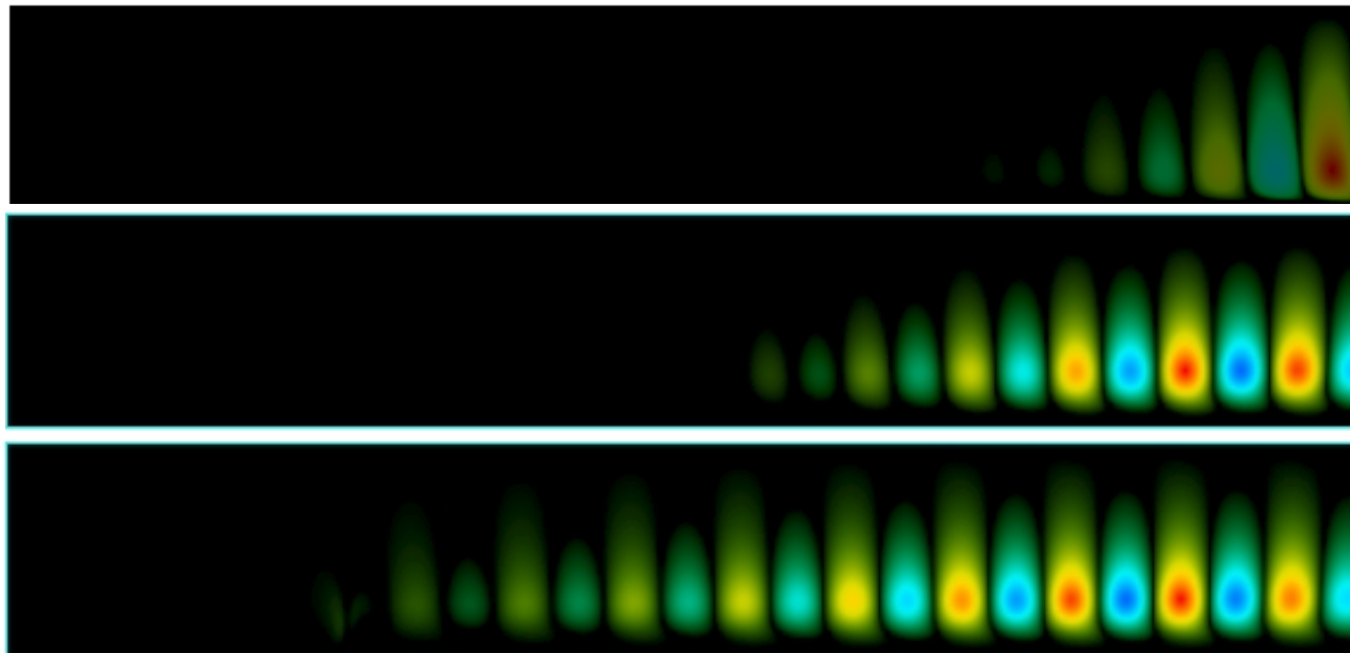
Matrix-free methods:

- Arnoldi method
- Snapshot method
- Snapshot method of direct and adjoint equations

Blasius Modes



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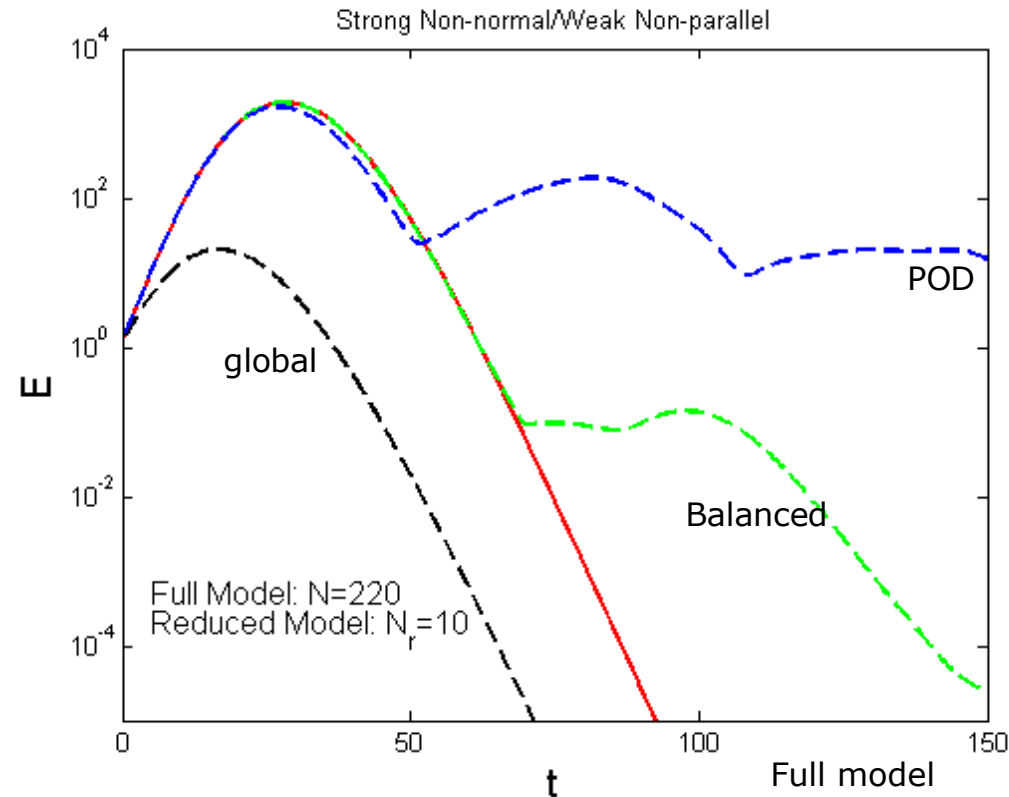
Matrix-free methods:

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Optimal Growth



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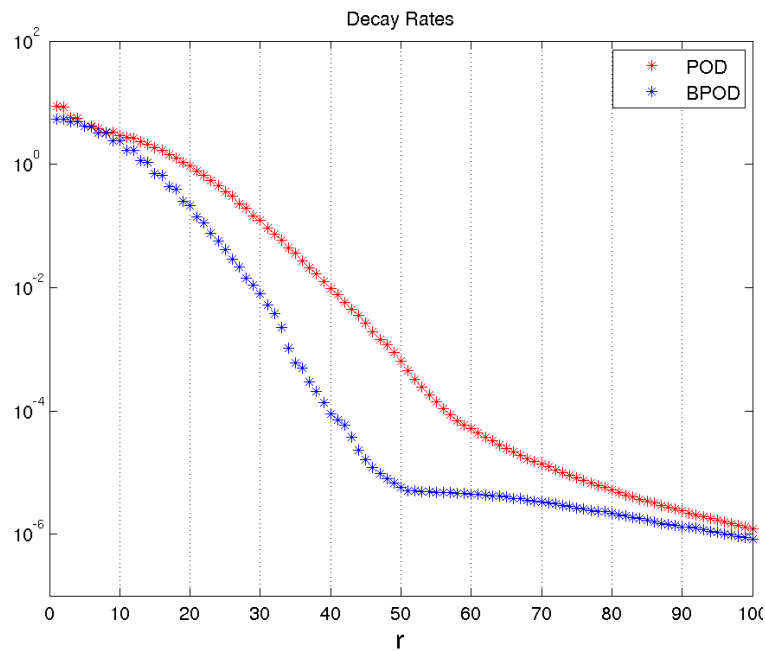


- B and C are at branch I and II respectively.
- Full-order: $n=220$.
- Reduced-order: $m=10$
- Balanced modes performs best

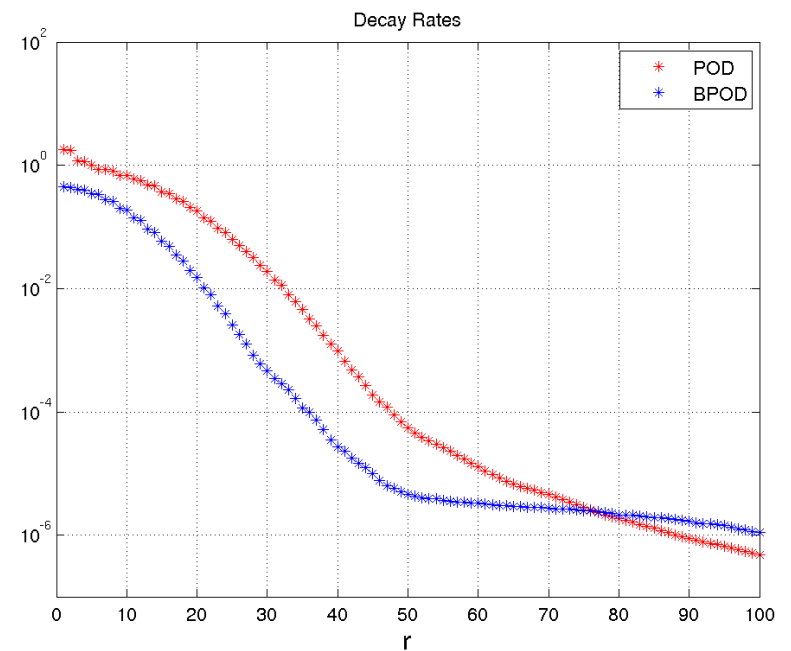
Decay rates for Blasius



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u-component



v-component

Examples of model reduction for fluid systems



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- Control and optimal growth of globally unstable cavity-like flow using 2D **global modes**. Åkervik et al. (Jfm, 2007)
- Projection of actuators and sensors, frequency response and optimal growth using **balanced modes** on the 3D channel flow and 2D aerofoil. Rowley et al 2005-2007.
- **POD modes** (with shift mode) to describe flow around circular buildings. Noack et al (jfm, 2003).

Comparison of modes



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- Spatial support. of modes,

$$\frac{\text{abs}(V)}{\max(V)} < 0.99$$

