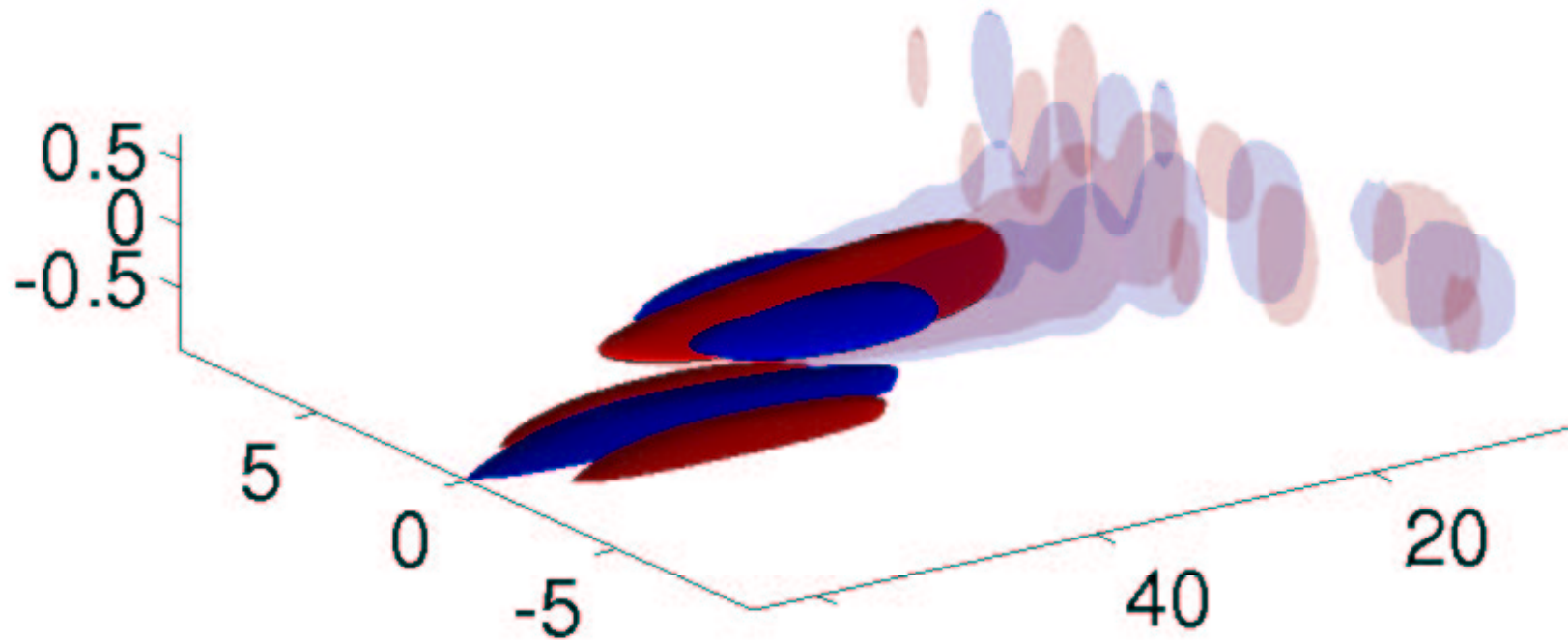




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# Coupling sensors and actuators for flow control



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## In this talk

Describe how the actuators uses the sensor information in optimal control of Channel flow.

## Framework

linear feedback control theory  
transition to turbulence

## Keywords

Reactive control (feedback)  
Transfer function



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## Why using reactive control?

### Act on the mean flow

And affect the stability of small perturbations

→ control effort of the order of magnitude of the mean flow

### Act on the fluctuations

And prevent them from growing and disrupting the mean flow

→ control effort on the order of magnitude of the fluctuations

**In a transitionnal case, the fluctuations are of much smaller amplitude than the mean flow.**



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# Information and action

## Sensors

Streamwise skin friction fluctuations

Spanwise skin friction fluctuations

Wall pressure fluctuations

## Actuators

Blowing and suction at the walls

**Only flow quantities at the wall are available.**



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# The control problem

Stochastic disturbances  $f$ ,  $g$ ,  $q_0$

(External sources, sensor noise, unknown initial condition)

Actuation and sensing  $u$ ,  $y$

$$\begin{cases} \dot{q} = Aq + B_1 f + B_2 u, & q(0) = q_0, \\ y = Cq + g, \end{cases}$$

Feedback control

$$u = \mathcal{G}(y)$$

Which is the optimal mapping  $\mathcal{G}$  ?



# Solution of the control problem

$$\text{Plant} \begin{cases} \dot{q} = Aq + B_1 f + B_2 u \\ y = Cq + g. \end{cases}$$

$$\text{Estimator} \begin{cases} \dot{\hat{q}} = A\hat{q} + B_2 u - v \\ \hat{y} = C\hat{q}. \end{cases}$$

$$\text{Feedback} \quad v = L\tilde{y} = L(y - \hat{y}), \quad u = K\hat{q}.$$

Decouple into an estimation problem and a full information problem. Solve two Riccati equations to get the optimal  $L$  and  $K$ .

Transfer function:

$$u(t) = \int_0^\infty \underbrace{-K e^{(A+BK+LC)\tau} L}_{G(\tau)} y(t - \tau) d\tau.$$



## Selected literature

- Hu H. H. & Bau, H.H. 1994 *Feedback control to delay or advance linear loss of stability in planar Poiseuille flow*  
Use of proportional controller :  $u(t) = Ky(t)$
- Joshi, S. S., Speyer, J. L. & Kim, J. 1995 *Modeling and control of two dimensional Poiseuille flow*  
Introduction of the optimal feedback control method (LQG, or  $H_2$ ).
- Högberg, M. & Bewley, T. 2002 *Spatially localised convolution kernels for decentralised control and estimation of plane channel flow*  
Decomposition of the control into state estimation and full information control.  
Spatial localisation of the feedback law.



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# LQG (or $H_2$ ) feedback control

LQG for

**Linear**

Use of a linear model for the dynamics

→ Use of linearised Navier–Stokes equations

**Quadratic**

A quadratic objective function

→ minimise the energy of flow fluctuations

**Gaussian**

Gaussian disturbances to the flow

→ Use a covariance model for the disturbances

**Fondamental achievement of control theory**





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## Physical assumptions

- Dense array of sensors and actuators  
We know all the wall information
- Periodic domain in the two homogeneous direction  
For Fourier transform and temporal study
- Low amplitude for the fluctuations  
to use the linearised Navier–Stokes equations



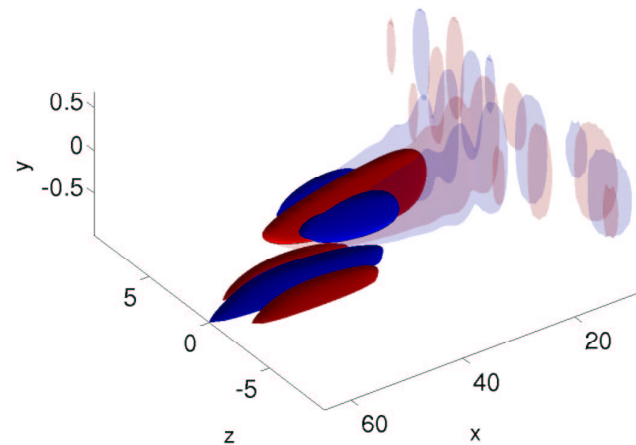
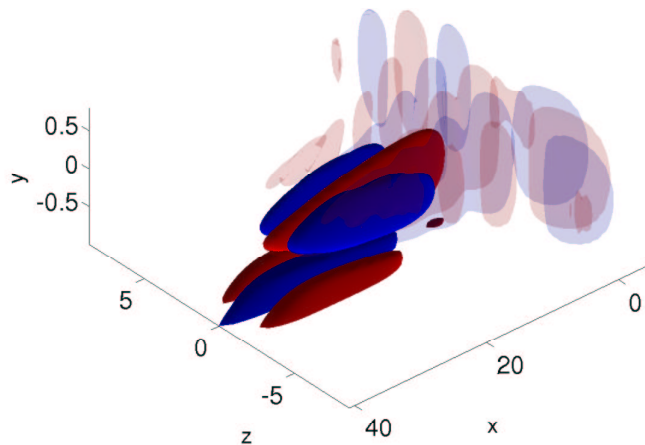
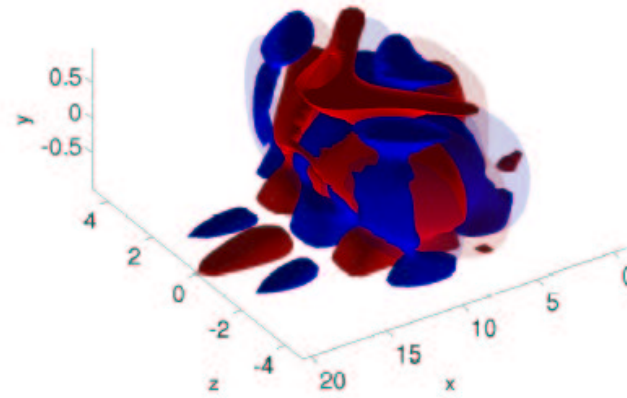
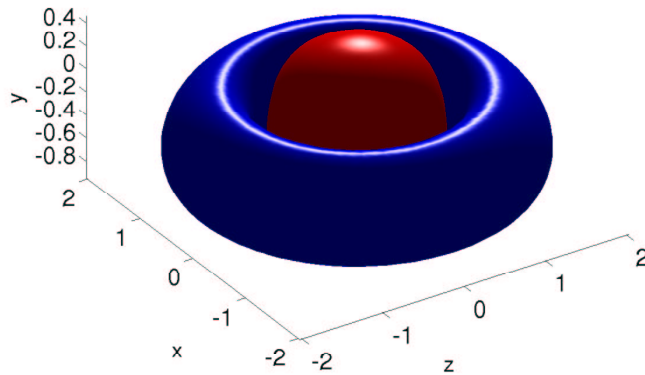
## Why a numerical study?

- Sensing and actuation  
Dense arrays of actuators and sensors are difficult to implement in experiment.
- Computational time  
With the actual formulation we need to run an on-line simulation of the flow.
- Understanding  
There is still many issues to be addressed on disturbance modeling, choice of control objective, and feedback formulation for flow applications.

# Test case

Axisymmetric localised initial condition

Wall normal velocity for original flow and controlled flow, Time 0, 10, 70, 90.

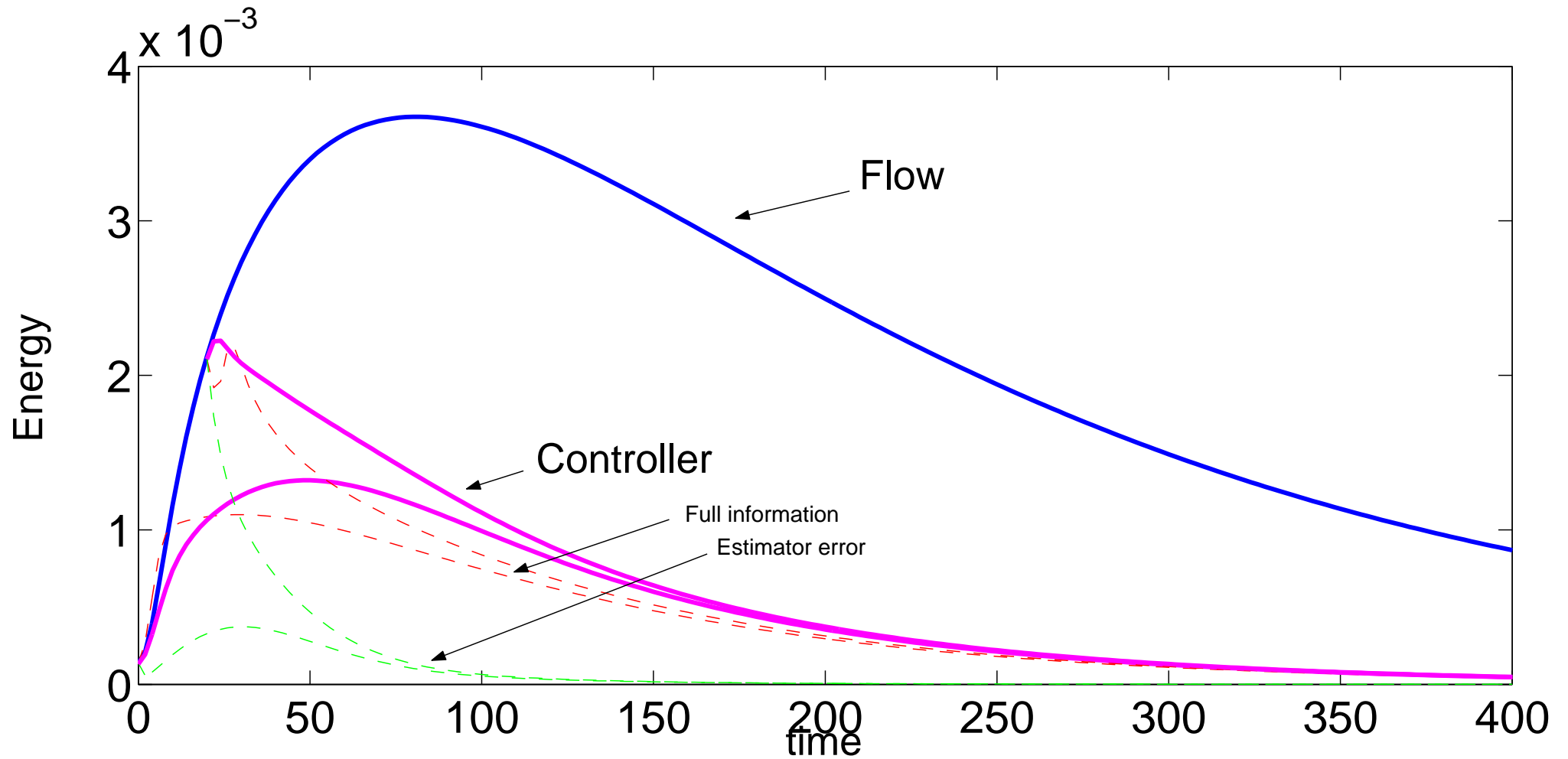




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# Performance of the control

Turn on the controller at time 0 and time 20





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## Transfer function formulation

The linear mapping can be written in the transfer function formalism:

$$u(0, 0, t) = \mathcal{G}(y) = \int_x \int_z \int_0^\infty G(x, z, \tau) y(x, z, t - \tau) d\tau$$

Convolution of the measurement history over the wall.

$\tau$  is the time lag



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## Potential instability of the TF

The closed loop is stable by construction

$$\begin{cases} \dot{q} = Aq + B_1 f + B_2 u, & q(0) = q_0, \\ y = Cq + g, \end{cases}$$

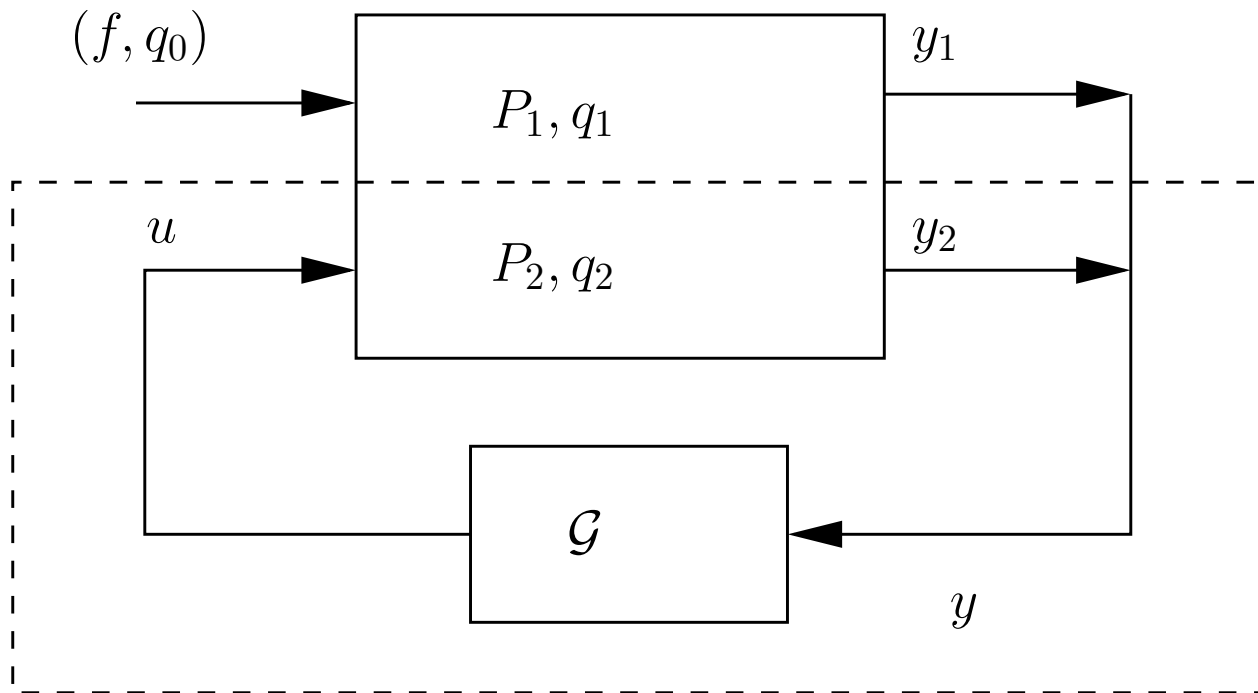
$$u = \mathcal{G}(y)$$

But  $\mathcal{G}$  is not guaranteed to be stable.

The interconnection of unstable systems can be stable.

# How to continue?

## Redefine of the input



Because the input  $y$  should be dependent on the output

$u$

The control affects the measurement.

Split  $y$  into  $y_1$  and  $y_2$

$$y = y_1 + y_2$$

$$q = q_1 + q_2$$

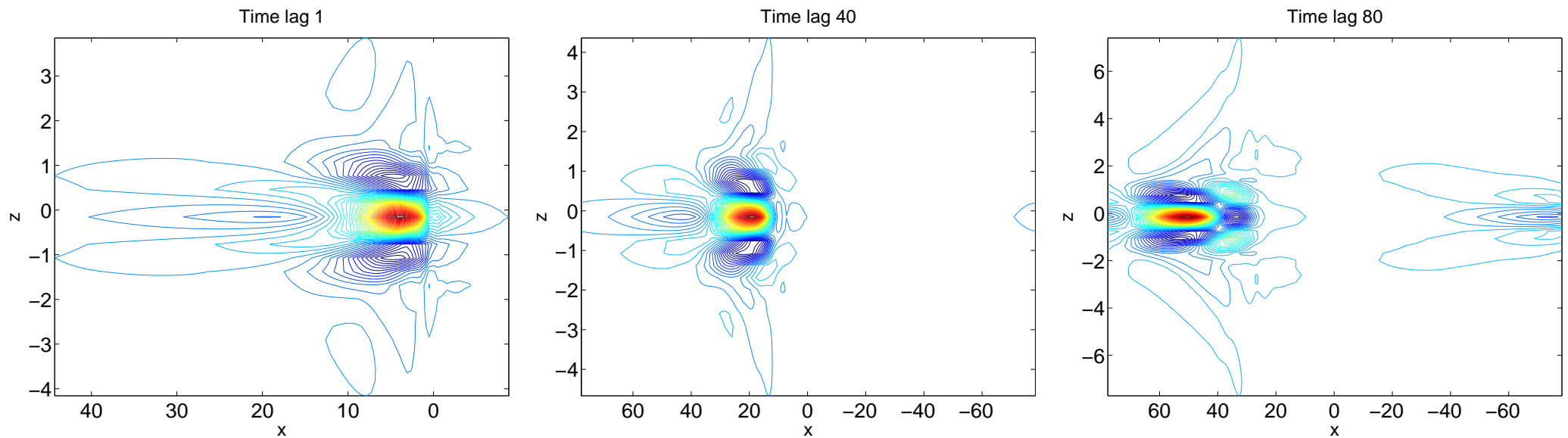
$u = \mathcal{G}^*(y_1)$  is the optimal control.

Now we continue with  $\mathcal{G}^*$

# The TF in the channel

For selected time lags  $\tau=1, 40, 80$ .

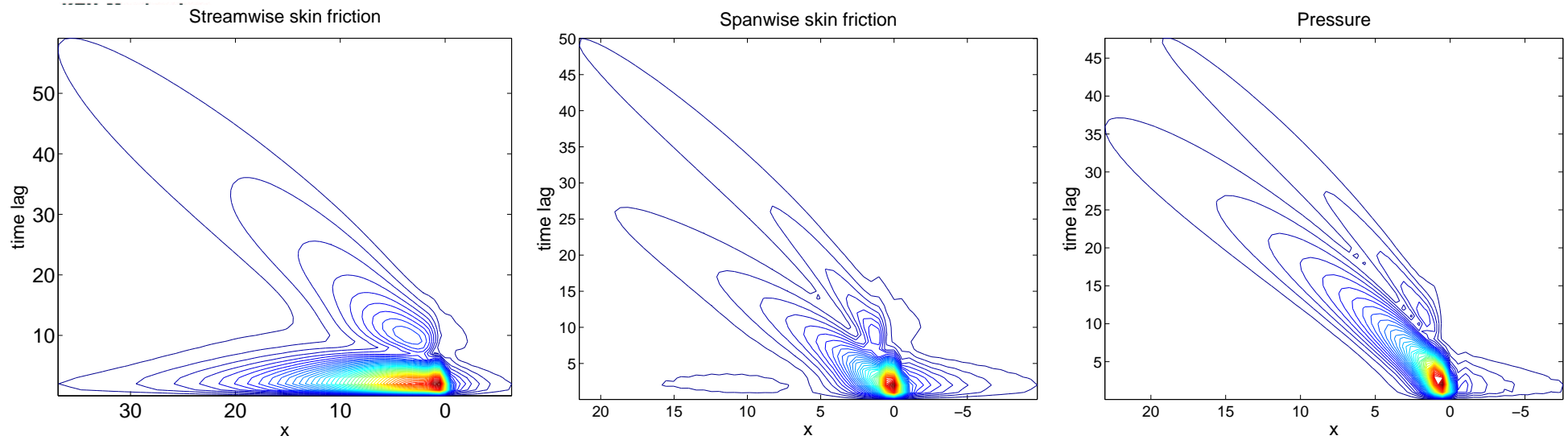
Streamwise skin friction measurement.



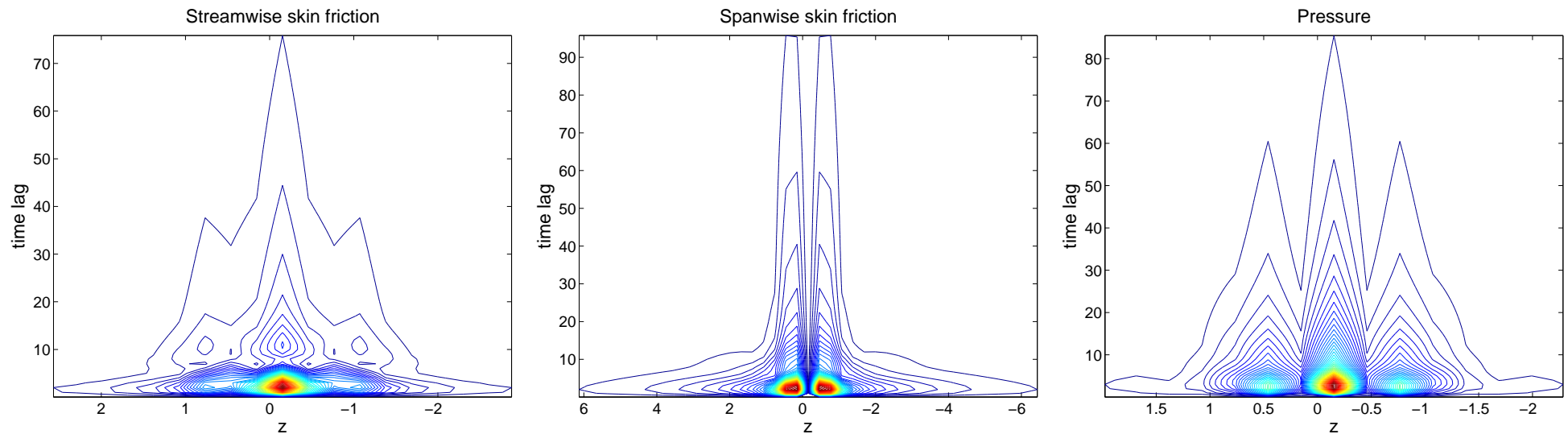


# Convected information

Integrated in streamwise direction



Integrated in spanwise direction





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## Conclusions

- The transfer function is a natural formulation for control with spatially distributed sensing and actuation
- The transfer function is potentially unstable, even though it stabilises the flow
- This instability is due to the coupling between the input  $y$  and the output  $u$