

# Feedback Control of Turbulent Wall Flows

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# Outline

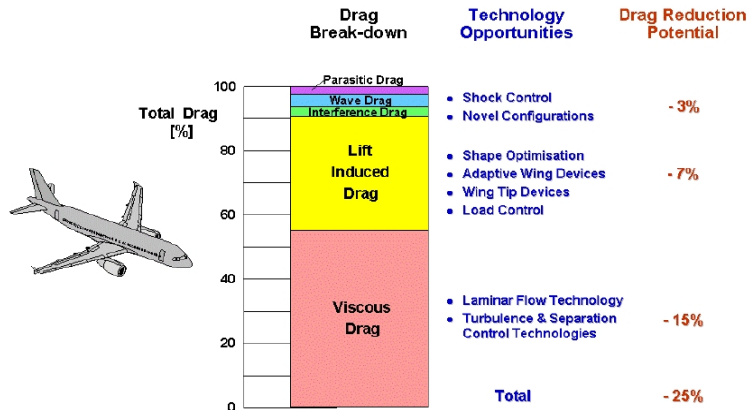
Introduction

Standard approach

Wiener-Hopf approach

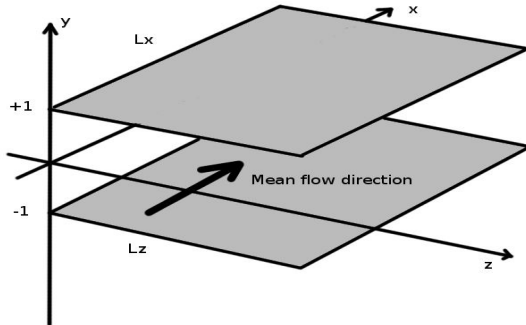
Conclusions

# Drag reduction



# A model problem: turbulent channel flow

Incompressible flow between two plane, parallel, infinite walls



- ▶ Flow is **spatially invariant** with  $x$  and  $z$
- ▶ In DNS, typically,  $\approx 10^8$  d.o.f.s

## Linear(ized) model

- ▶ Governing Navier–Stokes equations written in  $v-\eta$  formulation
- ▶ Linearization about a reference profile (**direct** approach)
- ▶ Fourier expansion in homogeneous directions: wavenumbers decouple

Recent evidence exists (e.g. Kim & Lim, PoF 2000; Kim, PoF 2003 ) that linear models are good for control design.

# State-space form

Linear time-invariant (LTI) system:

$$\dot{x} = Ax + Bu + r$$

$$y = Cx + d$$

- ▶ **Actuation**: distributed wall blowing/suction at the walls
- ▶ Sensing: distributed skin friction or pressure measurements
- ▶ Nonlinearity and modeling errors recovered as a state **noise** (typically **white**)

# Model-based optimal compensator

**Goal:** design of a feedback compensator minimizing

$$J = E\{x^H Q x + u^H R u\}$$

- ▶ Separation theorem: optimal controller and state estimator can be designed separately
- ▶ Requires the solution of two matrix Riccati equations having the same dimension of  $A$

# Open issues

- ▶ A state-space realization is required from measured models
- ▶ Accounting for the full space-time structure of the state noise is impractical
- ▶ Riccati-based design is prohibitively expensive for high-dimensional systems



# A measured linear model

## Looking for the average impulse response to wall forcing

- ▶ Wall forcing with a **small** space-time white Gaussian noise on the wall-normal velocity at the wall  $v_w$
- ▶ In the **linear** setting, the perturbed flow reads:

$$v_{tot}(x, y, z, t) = \bar{v}(x, y, z, t) + v(x, y, z, t)$$

$$\eta_{tot}(x, y, z, t) = \bar{\eta}(x, y, z, t) + \eta(x, y, z, t)$$

# A measured linear model

Computing the cross-correlation between the state and the wall forcing:

$$\begin{aligned}
 E\{v_{tot}(x' + x, y, z' + z, t' + t)v_w^*(x', z', t')\} = & \dots \\
 & \dots \underbrace{E\{\bar{v}(x' + x, y, z' + z, t' + t)v_w^*(x', z', t')\}}_{=0} + \dots \\
 & \dots + E\{v(x' + x, y, z' + z, t' + t)v_w^*(x', z', t')\}
 \end{aligned}$$

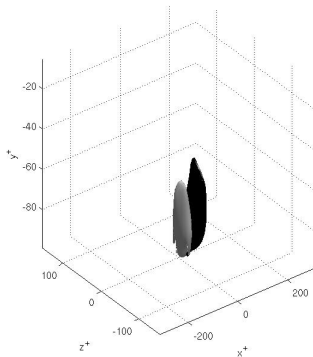
# The average impulse response

Leveraging a result in linear system theory, the state-forcing cross-correlation **defines** the **impulse response** of a LTI system:

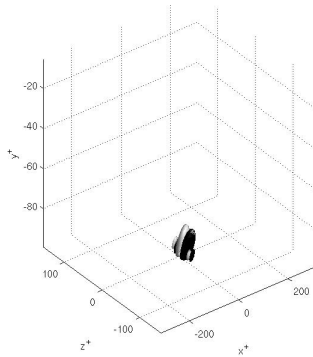
$$H_v(x, y, z, t) = E\{v(x' + x, y, z' + z, t' + t)v_w^*(x', z', t')\}$$
$$H_\eta(x, y, z, t) = E\{\eta(x' + x, y, z' + z, t' + t)v_w^*(x', z', t')\}.$$

This function represents the **average** response of a turbulent channel flow when impulsive wall forcing on  $v$  is applied.

# The average impulse response

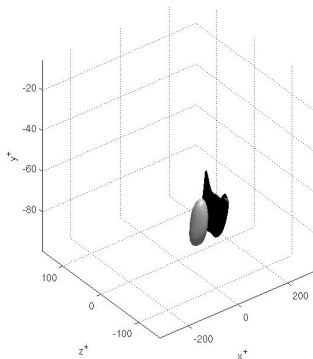


$$H_V(x, y, z, t^+ = 5)$$

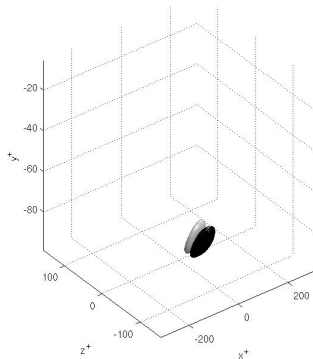


$$H_\eta(x, y, z, t^+ = 5)$$

# The average impulse response

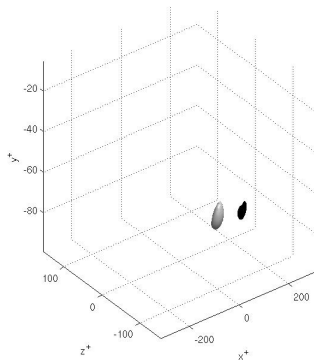


$$H_V(x, y, z, t^+ = 15)$$

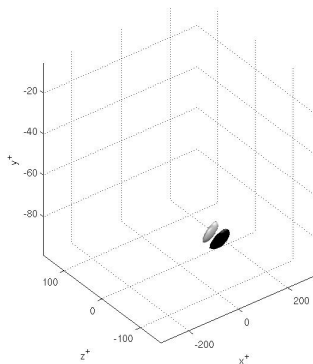


$$H_\eta(x, y, z, t^+ = 15)$$

# The average impulse response

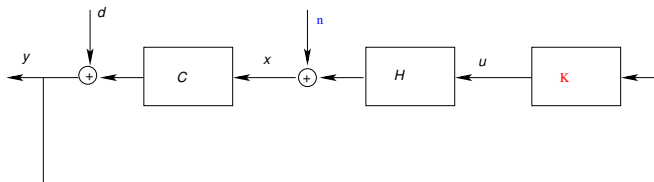


$$H_V(x, y, z, t^+ = 25)$$



$$H_\eta(x, y, z, t^+ = 25)$$

# Feedback control loop



- ▶  $n$  represents turbulent fluctuations in the **uncontrolled** flow
- ▶ The aim is to design a  $K$  such that the expectation

$$J = E\{x^H Q x + u^H R u\}$$

is minimized in the closed loop system

# Compensator design

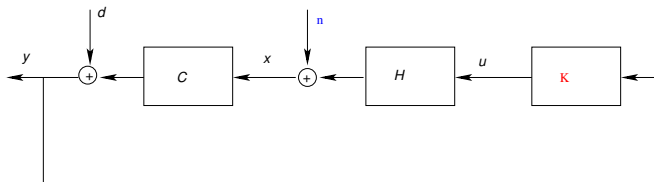
- ▶ It would be nice to avoid a state-space realization of  $H$
- ▶ It would be very nice to devise a procedure reducing the complexity of the compensator design

However...

- ▶ In this problem, LTI systems with wide-sense stationary stochastic forcing are considered
- ▶ A **frequency domain** approach is feasible



## Frequency domain approach



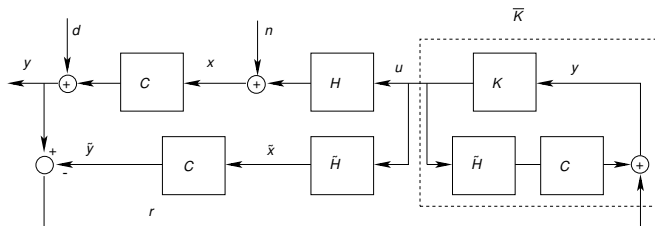
Rewriting the objective functional in frequency:

$$J = \int_{-\infty}^{+\infty} \text{Tr}[Q\phi_{xx}(f)] + \text{Tr}[R\phi_{uu}(f)] df.$$

Substituting,  $J$  is not quadratic in  $K$ .

## Internal Model Control structure

Introducing a model  $\tilde{H}$  of the plant  $H$ :



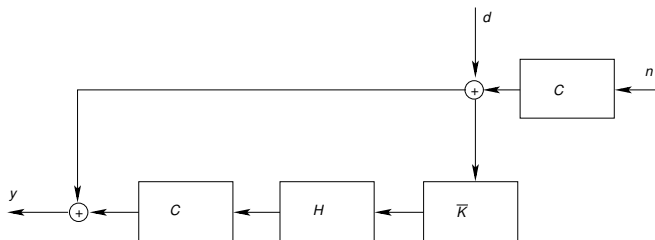
and  $\bar{K}$  may be written as:

$$\bar{K} = (I - K\tilde{H})^{-1}K.$$

This change of variable is known as Youla's parametrization (Morari & Zafiriou, 1998).

## Equivalent open loop system

It is assumed that  $\tilde{H} = H$  (modeling errors included in  $n$  only).



Substituting,  $J$  is now quadratic in  $\bar{K}$ !

# Optimal compensator in frequency domain

$$\begin{aligned}
 J = \int_{-\infty}^{+\infty} \text{Tr} \{ & Q\phi_{nn} + QH\bar{K}C\phi_{nn} + Q\phi_{nn}C^H\bar{K}^H H^H + \dots \\
 & \dots + QH\bar{K}C\phi_{nn}C^H\bar{K}^H H^H + QH\bar{K}\phi_{dd}\bar{K}^H H^H \} + \dots \\
 & \dots + \text{Tr} \{ R\bar{K}C\phi_{nn}C^H\bar{K}^H + R\bar{K}\phi_{dd}\bar{K}^H \} df.
 \end{aligned}$$

- ▶ Minimization leads to the best possible LTI compensator for the problem at hand
- ▶ However, such compensator is **noncausal**

## Causality enforcement

Introducing an appropriate Lagrange multiplier to enforce causality

$$\begin{aligned}
 J = \int_{-\infty}^{+\infty} \text{Tr} \{ & Q\phi_{nn} + QH\bar{K}_+ C\phi_{nn} + Q\phi_{nn}C^H\bar{K}_+^H H^H \dots \\
 & \dots + QH\bar{K}_+ C\phi_{nn}C^H\bar{K}_+^H H^H + QH\bar{K}_+ \phi_{dd}\bar{K}_+^H H^H \} + \dots \\
 & \dots + \text{Tr} \{ R\bar{K}_+ C\phi_{nn}C^H\bar{K}_+^H + R\bar{K}_+ \phi_{dd}\bar{K}_+^H \} + \text{Tr}[\Lambda_- \bar{K}_+^H] df.
 \end{aligned}$$

Plus and minus subscripts denote frequency response functions of causal and anticausal response functions, respectively.

## Wiener-Hopf problem

Minimization leads to the following **Wiener-Hopf** problem:

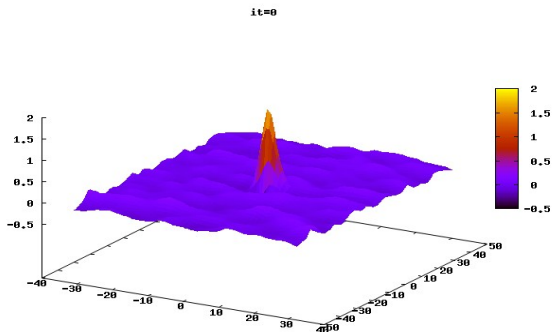
$$(H^H QH + R)\bar{K}_+(C\phi_{nn}C^H + \phi_{dd}) + \Lambda_- = -H^H Q\phi_{nn}C^H$$

- ▶ Solution to this problem yields directly the compensator's frequency response, without invoking the separation theorem
- ▶ Directly accounts for the input-output relation (no model order reduction needed)
- ▶ Noise spectral densities appear in functional form  $C\phi_{nn}$ : full space-time structure of the noise easily accounted for
- ▶ **Scalar** equation for the single-input/single-output case

# Compensator design and testing procedure

- ▶ Response function and noise spectral densities are measured via DNS and Fourier transformed in  $x$  and  $z$
- ▶ Wiener-Hopf problem is solved wavenumber-wise
- ▶ Compensators are tested in a full nonlinear DNS

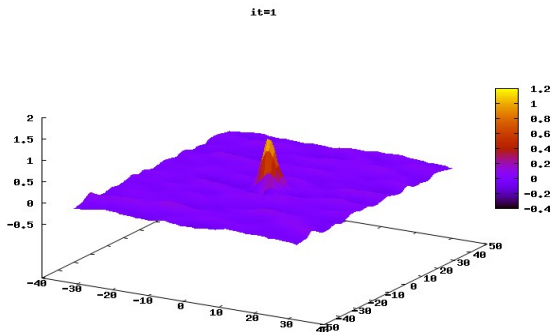
# Compensator kernel in physical space



$$u(x, z, t) = \int K(x - x', z - z', t - t') y(x', z', t') dx' dz' dt'$$

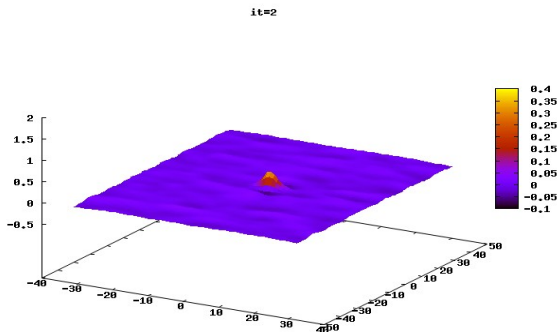


# Compensator kernel in physical space



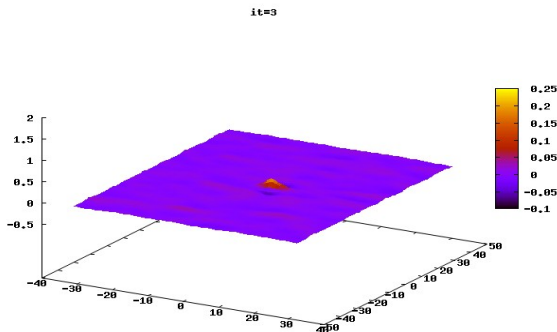
$$u(x, z, t) = \int K(x - x', z - z', t - t') y(x', z', t') dx' dz' dt'$$

# Compensator kernel in physical space



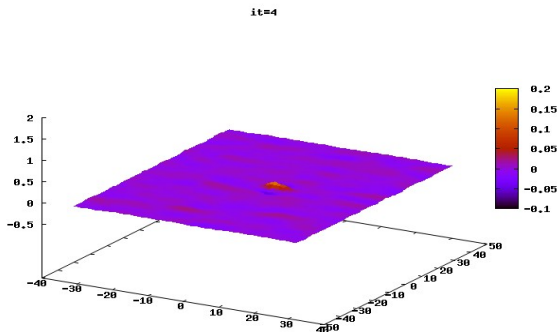
$$u(x, z, t) = \int K(x - x', z - z', t - t') y(x', z', t') dx' dz' dt'$$

# Compensator kernel in physical space



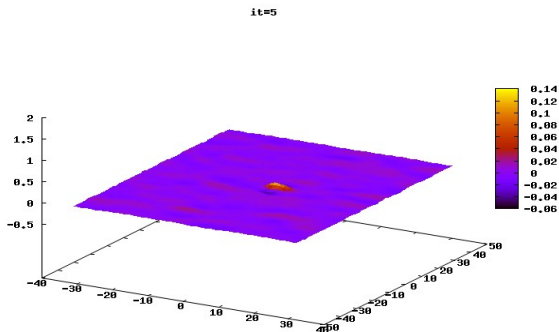
$$u(x, z, t) = \int K(x - x', z - z', t - t') y(x', z', t') dx' dz' dt'$$

# Compensator kernel in physical space



$$u(x, z, t) = \int K(x - x', z - z', t - t') y(x', z', t') dx' dz' dt'$$

# Compensator kernel in physical space



$$u(x, z, t) = \int K(x - x', z - z', t - t') y(x', z', t') dx' dz' dt'$$

# Performance assessment

## Parametric study addressing:

- ▶ Objective functional (Q matrix based on energy and dissipation)
- ▶ Actuation/sensing technique
- ▶ *Re* effects

More than 300 DNS ( $\approx$  40 years of CPU time) run at the supercomputing system located at the University of Salerno.

## Best performance results

	Dissipation			Energy		
$Re_\tau$	$\tau_x$	$\tau_z$	$p$	$\tau_x$	$\tau_z$	$p$
100	2%	0%	0%	0%	0%	0%
180	8%	6%	0%	0%	0%	0%

- ▶ Energy norm is ineffective
- ▶ Dissipation norm is effective
- ▶ Pressure measurement alone is not useful
- ▶ “Inverse”  $Re$ -effect when using dissipation norm

Overall best performance with  $\approx 7.7\%$  of net power saved.

## “Inverse” *Re*-effect

The performance of dissipation-based compensators improves with *Re*. Why?

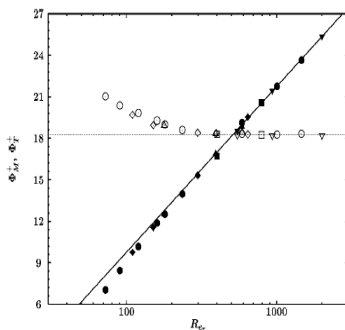
$$\frac{d\langle U \rangle}{dy} \Big|_w = -\frac{1}{U_B} \left\langle \underbrace{\sum_{(\alpha, \beta) \neq (0,0)} D(\alpha, \beta)}_{D_{turb}} + \underbrace{\frac{1}{2} \int_{-1}^1 \left( \frac{\partial \hat{U}}{\partial y} \right)_{(0,0)} \left( \frac{\partial \hat{U}}{\partial y} \right)_{(0,0)}^* dy}_{D_{mean}} \right\rangle$$

- ▶  $D_{turb}$  is affected **directly** by zero net mass flux blowing/suction
- ▶  $D_{mean}$  is affected **indirectly** via nonlinear interactions between fluctuations and the mean flow



## “Inverse” $Re$ -effect

But the relative contribution of the  $D_{turb}$  to the total dissipation **increases** with  $Re$ !



$Re_\tau$	$D_{turb}$	$D_{mean}$
100	26.8%	73.2%
180	39.5%	60.5%

Laadhari, PoF 2007

# Critical discussion

## The use of linear estimators

- ▶ Present compensators incorporate Wiener filters (instead of Kalman filters) accounting for the full space-time structure of the state noise
- ▶ They are the best possible LTI filters for this problem
- ▶ However, their estimation capability is similar to that of Kalman filters
- ▶ This suggests that the use of **linear** filter is the issue

Substantial improvement may be obtained by using nonlinear filters, providing accurate estimates of the state far away from the wall.

# Critical discussion

## Selecting of appropriate cost functions

- ▶ Present compensators are the best possible LTI compensators for the problem at hand
- ▶ Their performance is, however, rather poor if compared to that of optimal compensators designed with analogous, state-space techniques

In the linear setting, the sole remaining degree of freedom is the cost function to be minimized.

# Conclusions

- ▶ A novel cost-effective compensator design formulation has been proposed
- ▶ A measured linear model of the turbulent channel flow has been employed
- ▶ The approach accounts for the full time-space structure of the state noise

# Possible developments

## Control of wall turbulence

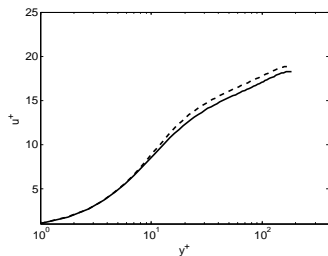
- ▶ Exploiting the compensator design methodology to optimize the cost function, with approximate models for the system dynamics and state noise
- ▶ Design based on experimentally measured shear-fluctuations cross-correlations
- ▶ Resorting to a state-space to optimize nonlinear (possibly reduced-order) estimators to be used in conjunction with standard optimal controllers

# Possible developments

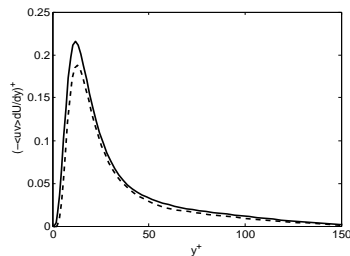
## Design methodology

- ▶ Robust formulation in the IMC framework
- ▶ Multiple-input/multiple-output design
- ▶ Use of the present technique in a nonlinear optimization

# Statistics of the controlled flow

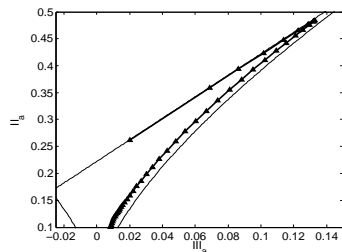


Mean velocity profile in the law-of-the-wall form

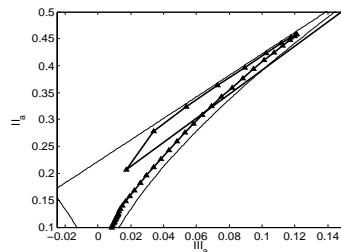


Production of turbulent kinetic energy

# Anisotropy pattern



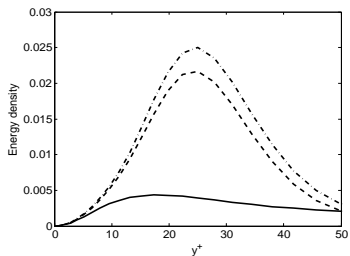
Uncontrolled case



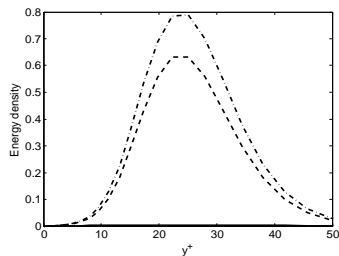
Controlled case



# Wiener filter performance



Streamwise skin friction



Pressure