



55th Annual Meeting

American Physical Society

Division of Fluid Dynamics

November 24th - 26th, 2002
Dallas, Texas



Feedback Control of a Cylinder Wake Low-Dimensional Model

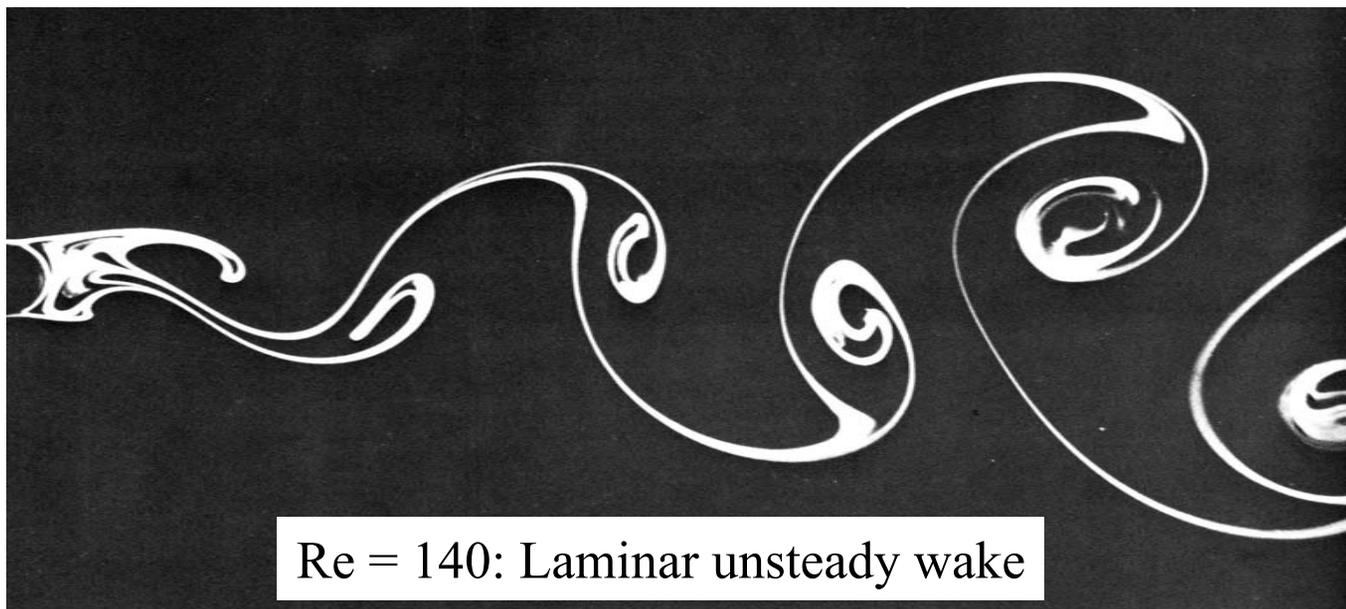
Kelly Cohen, Stefan Siegel, Thomas McLaughlin
Department of Aeronautics, US Air Force Academy, Colorado

Eric Gillies

Department of Aerospace Engineering, University of Glasgow, Scotland



INTRODUCTION

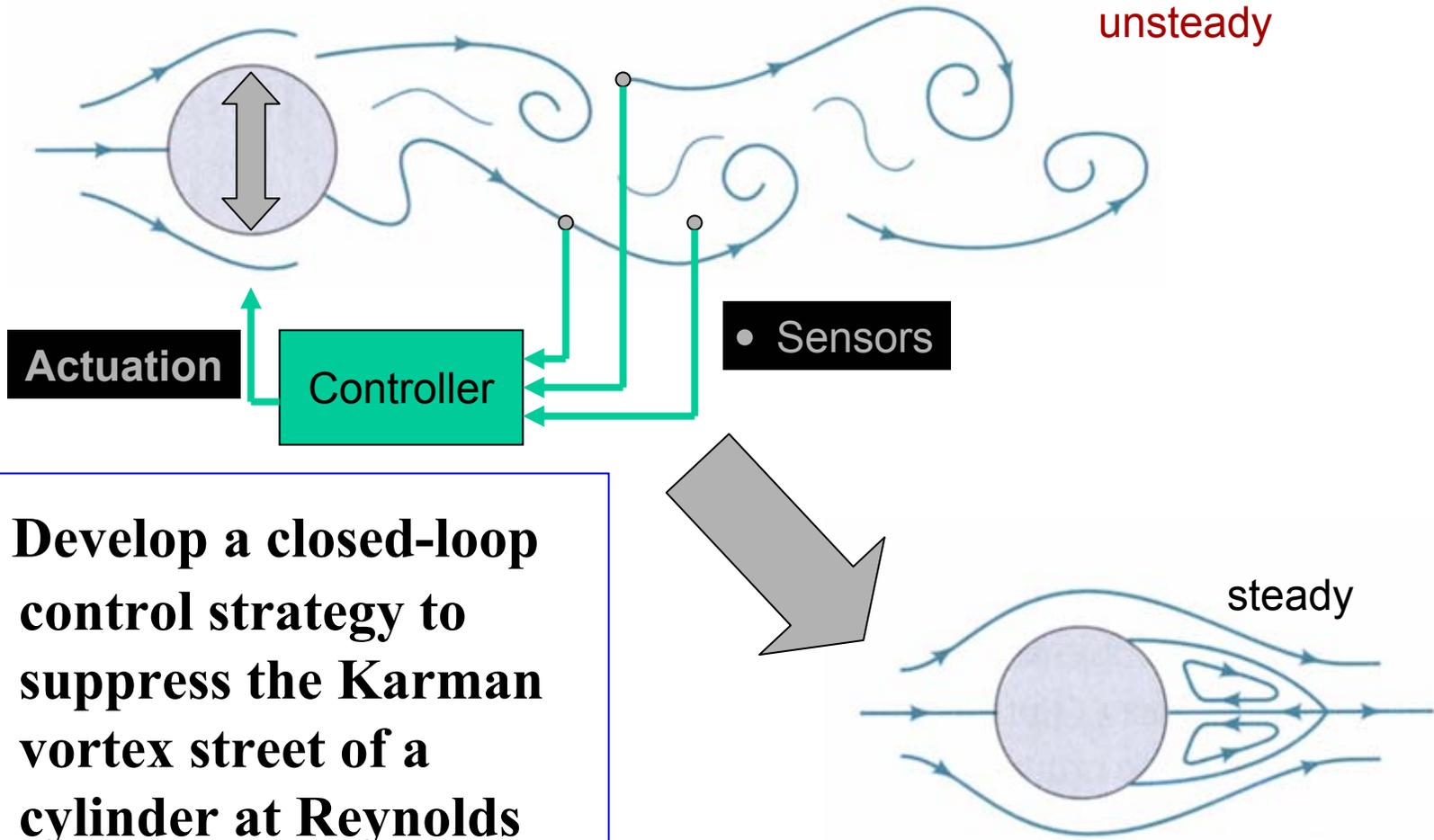


Re = 140: Laminar unsteady wake

- In a two-dimensional cylinder wake, self-excited oscillations in the form of periodic shedding of vortices are observed above a critical Reynolds number of around 50.
- These flow-induced non-linear oscillations lead to some undesirable effects associated with unsteady pressures such as fluid-structure interactions.
- The only way of suppressing the self-excited flow oscillations is by the incorporation of active closed-loop flow control.



RESEARCH OBJECTIVE



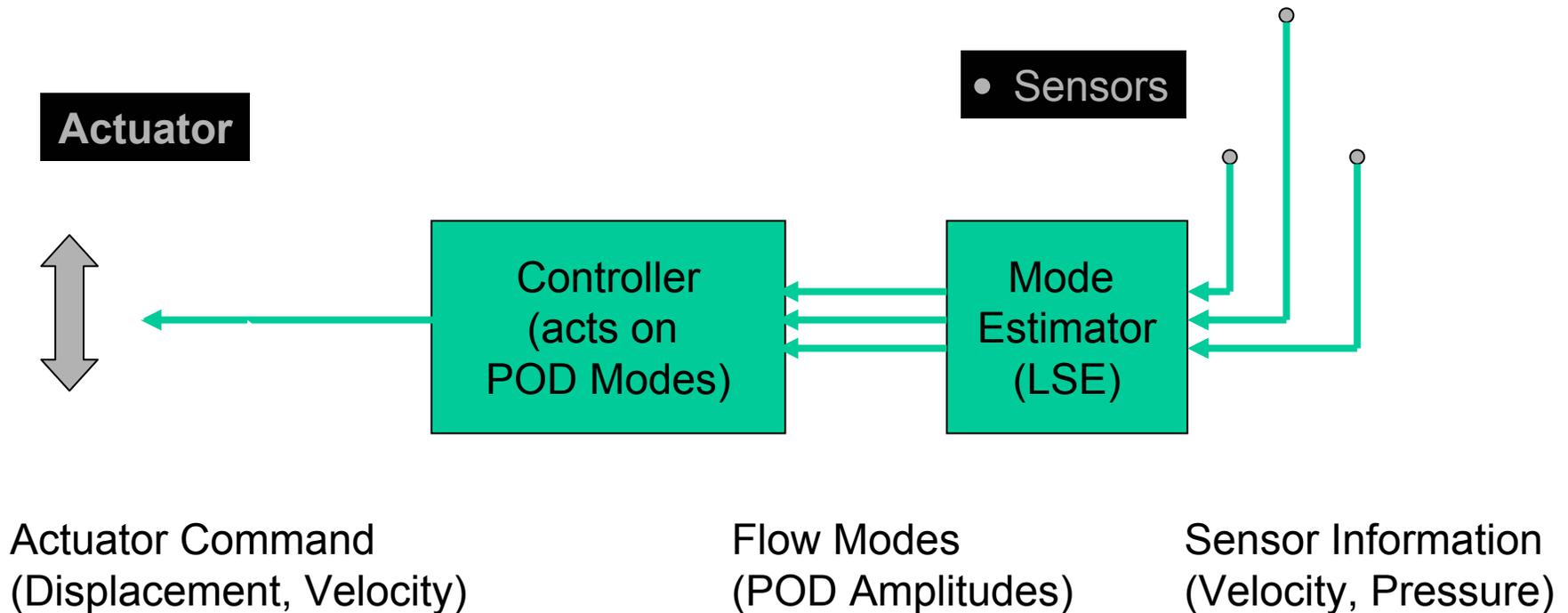
Develop a closed-loop control strategy to suppress the Karman vortex street of a cylinder at Reynolds numbers of 100-120.



CONTROL APPROACH



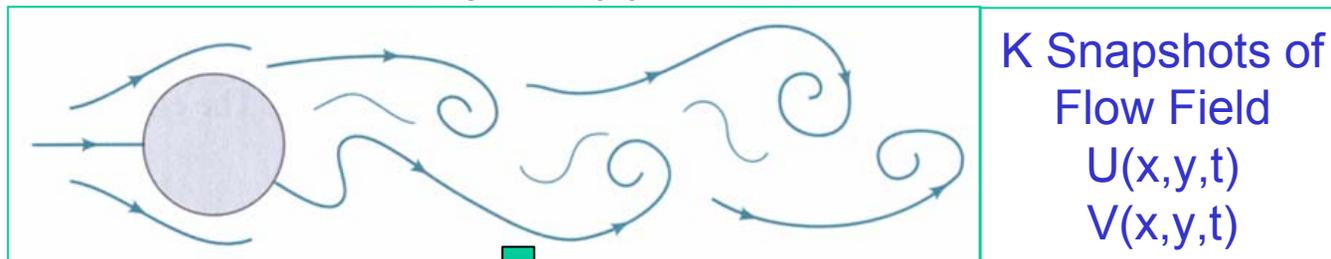
- **Low-dimensional model identification of the closed-loop system based on proper orthogonal decomposition (POD).**





LOW DIMENSIONAL MODELING

Re = 100



4 Mode POD Model based on DNS CFD simulations

K Temporal Mode Amplitudes
 $A_1(t)$
 $A_2(t)$
...
 $A_K(t)$

The control input was added to the POD model as an external source that emulates the translation actuation of the cylinder

N Spatial Modes
 $M_1-U(x,y)$
 $M_1-V(x,y)$
 $M_2-U(x,y)$
....
 $M_K-V(x,y)$



LOW-DIMENSIONAL POD MODEL

- Based on the temporal coefficients of the POD model, the following system of partial differential equations was developed using a least squares fit:

$$\dot{A}_k = g_k(A_1, A_2, \dots, A_M) + b_k f_a \quad (\text{for } k = 1, 2, \dots, M)$$

- The function g_k is chosen to be cubic:

$$g_k = c_0^k + c_{1i}^k A_i + c_{2ij}^k A_i A_j + c_{3ijl}^k A_i A_j A_l \quad (\text{for } i, j, k, l = 1, 2, \dots, M)$$

A_k - time-dependent coefficient of the k th mode

M - total number of modes in the low-dimensional model

g_k - nonlinear function of the time-dependent coefficients

f_a - control input to the cylinder

b_k - represents the coefficients associated with the control input



CONDITIONS FOR CONTROLLABILITY

- For nonlinear system $\dot{A}_K = g_K(A_1, A_2, \dots, A_M) + b_K f_a$
- The simplest approach to study controllability is to consider its linearization

$$\dot{A}_K = J \cdot A_j + b_K \cdot f_a \quad \text{Let } B = [b_1, b_2, \dots, b_M]$$

- Definition: *The pair (J, B) is state controllable if and only if there exists a control f_a that will transfer any initial state $A_K(t = 0)$ to the desired equilibrium point in finite time.*
- We will demand that the following algebraic condition for controllability:

$$\text{rank}(B : JB : J^2 B : \dots : J^{n-1} B) = n$$



CONTROL APPROACH

- It is desirable to develop estimator and controller strategies that are simple yet effective.
- **Let** the control law be based on the estimate of **only one mode** as follows:

$$\mathbf{f}_a = -\mathbf{K}_p \cdot A_1^{\text{est}}$$

- where \mathbf{K}_p is the proportional gain of the P controller and A_1^{est} is the estimate of the time-dependent coefficient of Mode 1, A_1 extracted from sensors placed in the wake.
- An accurate linear stochastic estimator for the first four POD modes of a circular cylinder wake was developed based on five sensors in the flow field.
- Based on this experience, in this effort estimation errors are neglected and it is assumed that $A_1 \sim A_1^{\text{est}}$.
- **Questions raised:**
 - **Is stability of all POD states assured for such a controller?**
 - **How are the gains of the controller determined?**
 - **Is such an approach effective?**



LINEARIZATION OF THE POD MODEL



- The function g_k is expanded locally as a Taylor series about the desired equilibrium point: $A_k = 0$
- Inserting the proposed proportional control law into the linearization of A_k yields:

$$\dot{A}_k = J_C \cdot A_j$$

- J_C is the “closed-loop” Jacobian and a linear stability analysis based on J_C will provide an insight into the behavior of the closed-loop system.

$$J_C = \begin{bmatrix} \frac{\partial g_1(0)}{\partial A_1} - b_1 \cdot K_P & \dots & \frac{\partial g_1(0)}{\partial A_M} \\ \dots & \dots & \dots \\ \frac{\partial g_M(0)}{\partial A_1} - b_M \cdot K_P & \dots & \frac{\partial g_M(0)}{\partial A_M} \end{bmatrix}$$



STABILIZATION OF THE POD MODEL



- The conditions for asymptotic stability for linearized models about their equilibrium point, as follows: *if the Jacobian, J_C , has n eigenvalues, each of which has a strictly negative real part, then the equilibrium point is asymptotically stable.*
- The Hartman-Grobman theorem states that the local phase portrait near a hyperbolic (all the eigenvalues of the linearization lie off the imaginary axis) fixed point is “topologically equivalent” to the phase portrait of the linearization.
- A linearized system that is hyperbolic is equivalent in terms of stability and bifurcations, chaos and attractors, equilibria and limit cycles to the nonlinear POD model.
- From a practical point of view, it is the aim of the control design to find an appropriate gain, K_p , which will render *all* the eigenvalues of J_C to have a *negative real part*. In addition, the eigenvalues need to lie off the imaginary axis by an adequate margin so that the system is hyperbolic.



CONTROLLABILITY TEST

Based on 4 Mode POD Model

Controllability condition: $\text{rank}(B:JB:J^2B:\dots:J^{n-1}B) = n$

where $n = 4$ (number of modes in the prototype wake model)

The matrices J and B are extracted from the above model

$$J = \begin{bmatrix} 0.0065 & 0.1447 & 0.0518 & -0.0258 \\ -0.1293 & 0.0065 & 0.0060 & 0.0364 \\ -0.0008 & -0.0000 & -0.0347 & -0.3156 \\ 0.0003 & 0.0007 & 0.2433 & -0.0292 \end{bmatrix} \quad B = \begin{bmatrix} -9.998e-03 \\ -9.866e-03 \\ -6.0577e-03 \\ -6.632e-03 \end{bmatrix}$$

We observe that $\text{rank}(B:JB:J^2B:\dots:J^{n-1}B) = 4$

Therefore, the pair (J, B) is state controllable



CLOSED-LOOP SIMULATIONS



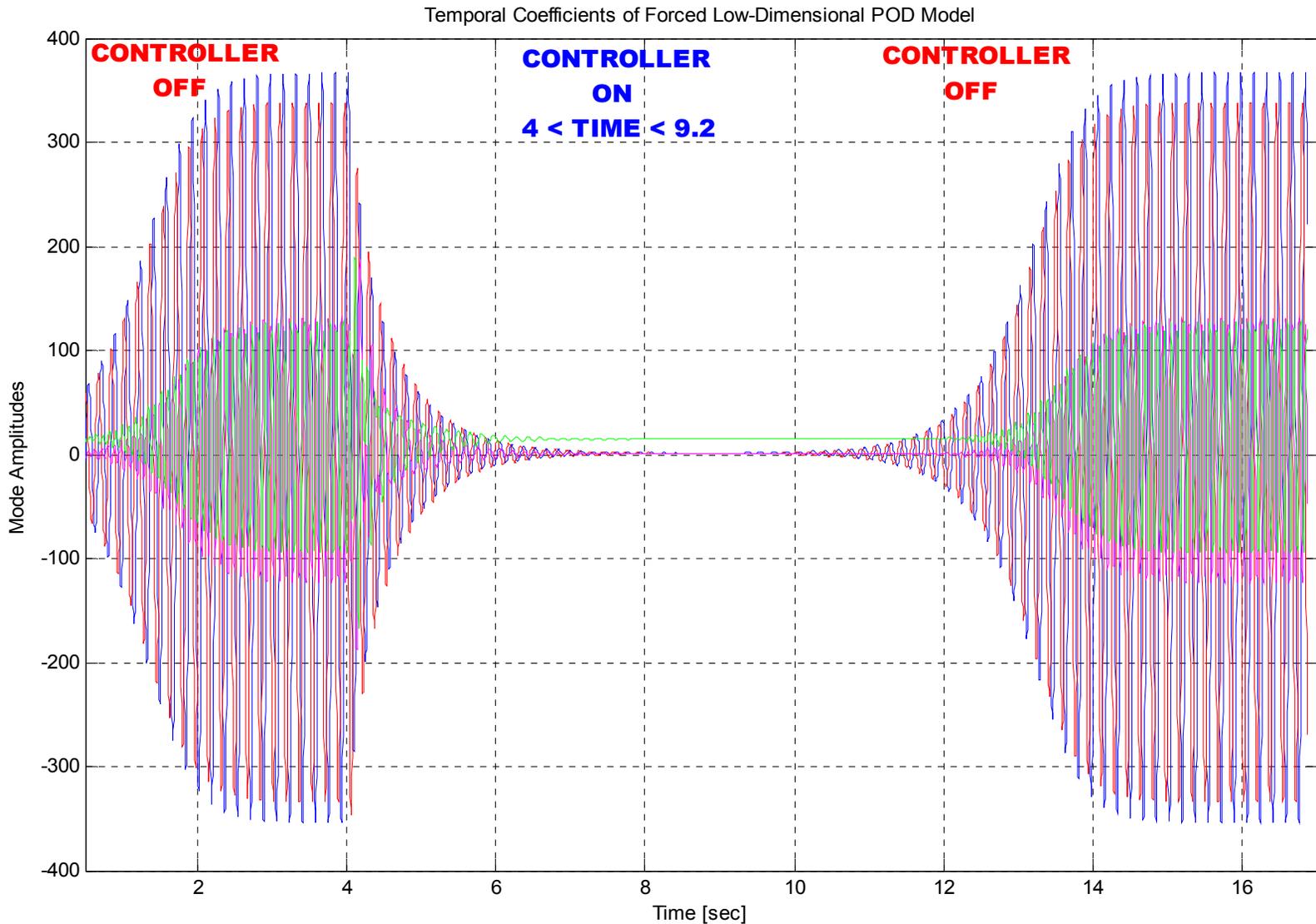
- MATLAB simulations were run to find an appropriate gain, K_p , that will render *all* the eigenvalues of the closed-loop Jacobian, J_C , to have a *negative real part*.

K_p	Eigenvalues 1 & 2	Eigenvalues 3 & 4
-1.2	$0.0115 \pm 0.1300i$	$-0.0310 \pm 0.2768i$
-0.8	$0.0098 \pm 0.1323i$	$-0.0313 \pm 0.2769i$
-0.4	$0.0081 \pm 0.1345i$	$-0.0316 \pm 0.2770i$
0.0	$0.0065 \pm 0.1367i$	$-0.0320 \pm 0.2772i$
0.4	$0.0048 \pm 0.1388i$	$-0.0324 \pm 0.2773i$
0.8	$0.0032 \pm 0.1409i$	$-0.0327 \pm 0.2774i$
1.2	$0.0016 \pm 0.1429i$	$-0.0331 \pm 0.2775i$
1.6	$-0.0000 \pm 0.1449i$	$-0.0335 \pm 0.2776i$
2.0	$-0.0016 \pm 0.1468i$	$-0.0339 \pm 0.2777i$
3.0	$-0.0055 \pm 0.1514i$	$-0.0350 \pm 0.2779i$

Therefore, for $K_p > 1.6$, the closed loop system is stable

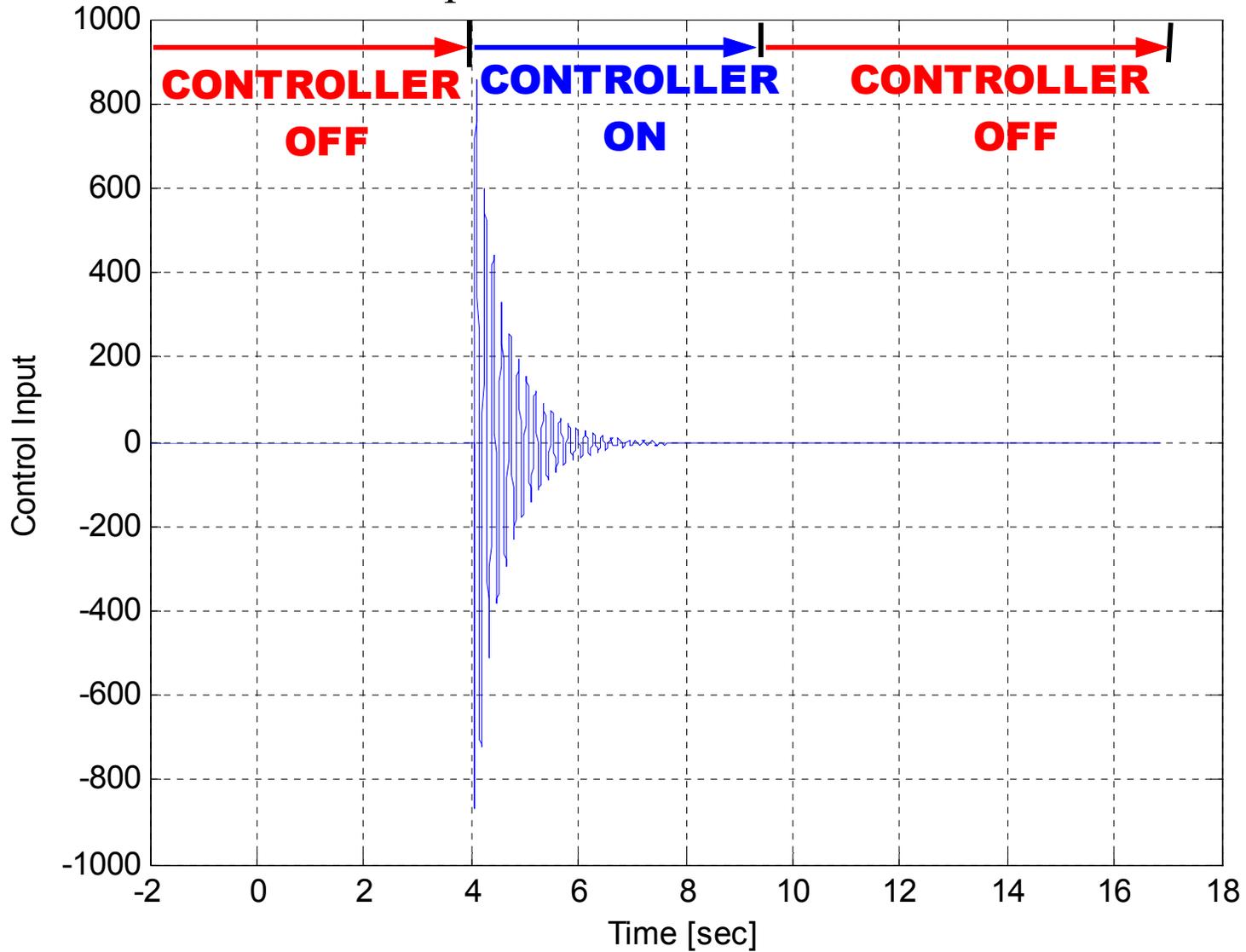


CLOSED-LOOP SIMULATIONS





Control Input Based on Estimate of Mode 1





CONCLUSIONS



- For the low-dimensional POD model of a cylinder wake instability, conditions for controllability and stabilization were developed.
- A simple approach, based on the proportional feedback of the estimate of the first POD mode, has been introduced for the control of the temporal growth of the forced prototype modes.
- The above control approach was applied to a POD model developed using data obtained from a Direct Navier Stokes CFD simulation of a cylinder wake at $Re = 100$.
- The control approach, simulated in MATLAB using the 4 mode cylinder wake model, shows desired closed-loop behavior.
- While the controller uses only the estimated amplitude of the first mode, all four modes are stabilized. This suggests that the higher order modes are caused by a secondary instability. Thus they are suppressed once the primary stability is controlled.
- The control approach stabilizes the wake fully. Once stabilized, the required actuation amplitude drops to $1/1000^{\text{th}}$ of the initial value. This means that very little energy is required to maintain a stable wake.