

Modeling of the Wake Behind a Circular Cylinder Undergoing Rotational Oscillations

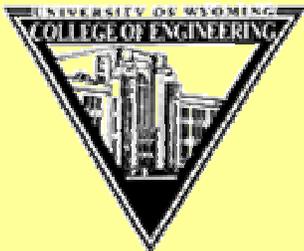
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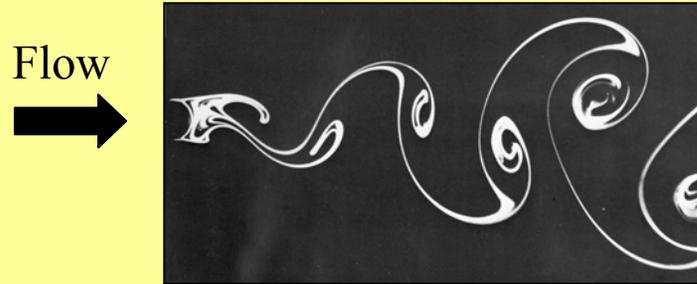
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USAFA Closed-loop Flow Control Program

Part I



- Demonstrate *closed-loop* control of a bluff-body wake
- Circular cylinder in a cross-flow at $Re_D=125$
 - Cylinder can rotate about its longitudinal axis: *control input*
- Using wake velocity measurements develop a reduced-order model of the wake when the cylinder oscillates
- Use this model to understand the response of the wake to the control input
- Develop a feedback control scheme that reduces *flow-induced vibration and drag*

Characteristics & Control of the Cylinder Wake

Characteristics

- A region of absolute instability in the near-wake gives rise to a self-sustained oscillation
 - Manifested as the Karman vortex street
 - Low-dimensional flow
- Oscillating wake flow feeds back unsteady forces to the cylinder
→ *Flow-induced vibrations*

Feedback Control

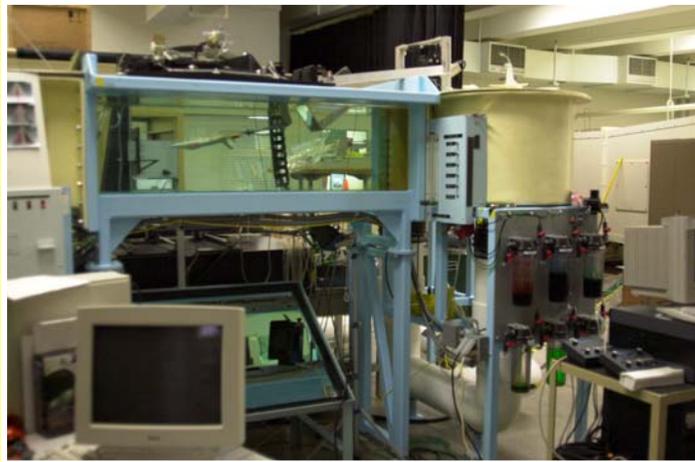
- Single sensor feedback can be effective at locally reducing wake oscillations → at sensor location
- 3-D effects from along the span re-establish shedding away from sensor (Roussopoulos & Monkewitz, 1996)
- Multi-sensor control works better but the control window closes as Reynolds number increases (Gillies, 1998)

Program

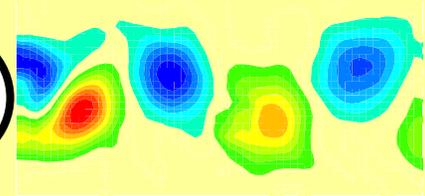
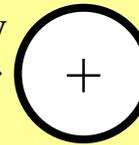
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Experimental Facility & Conditions

Water channel



Flow
→

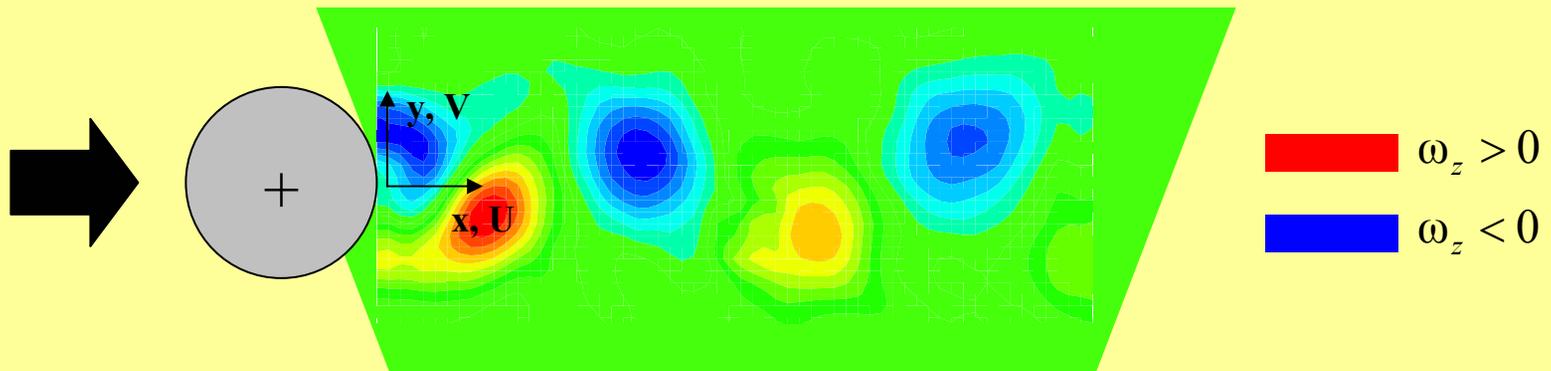


- Vortex shedding frequency is a fcn of Re_D
$$\left. \begin{array}{l} D = 2.38 \text{ mm} \\ U_\infty = 6.0 \text{ cm/s} \end{array} \right\} Re_D = 125 \Rightarrow S_f \equiv \frac{f D}{U_\infty} \approx 0.18$$
- Forcing Strouhal numbers, $S_f = 0.189, 0.248$
- Rotational forcing: $\Omega_1 = \frac{V_{\theta, \max}}{U_\infty} = \frac{2\pi r \theta_{\max} f_o}{U_\infty}$
 - Seventeen forcing conditions,
 $0.0 \leq \Omega_1 \leq 1.5$

PIV Measurements

- Velocity field measurements were made with Particle Image Velocimetry (U, V)
 - Seed water w/ small particles (10 μ m diameter)
 - Illuminate particles & take two, short-exposure photographs separated in time by a small, known Δt
 - Particle displacements Δx & Δy divided by Δt give velocities (U,V)
 - Data images were acquired at 15 Hz, asynchronously & phase-averaged w/ f_o

- Contour plots show cross-span vorticity, $\omega_z = \left(\frac{\partial V}{\partial x} - \frac{\partial U}{\partial y} \right)$



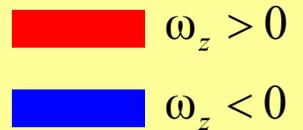
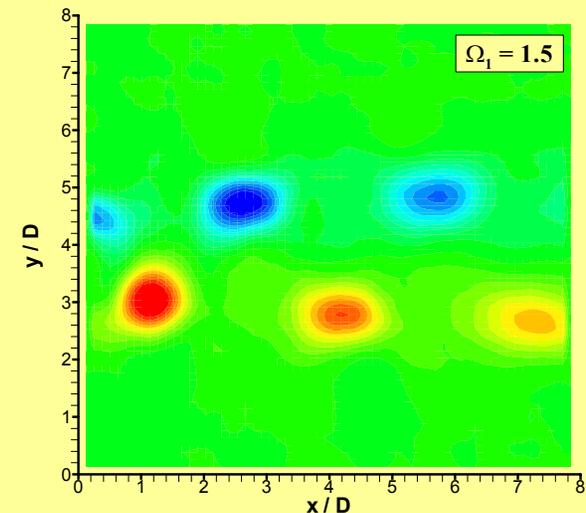
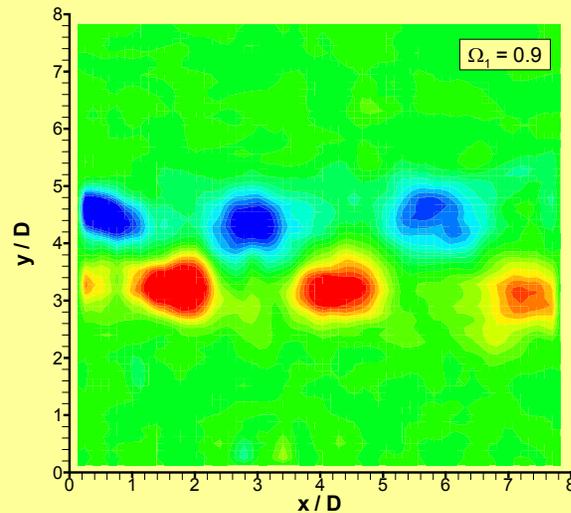
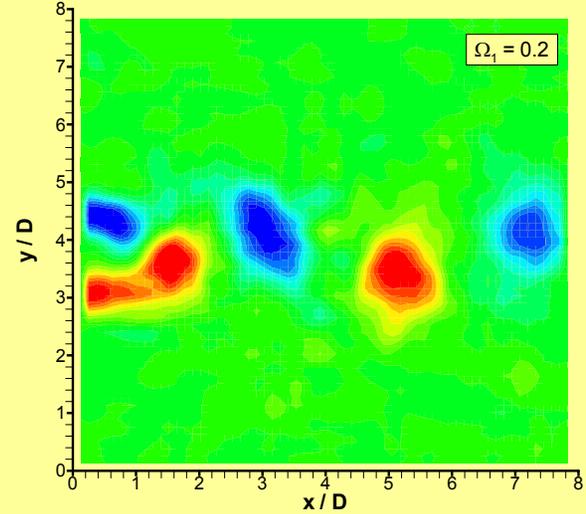
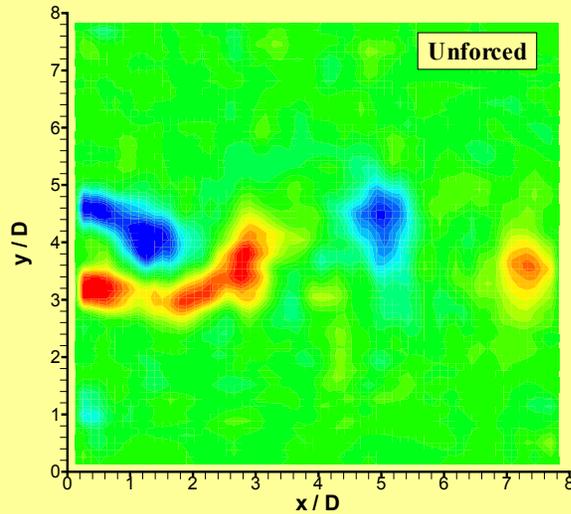
(View from below)

Program

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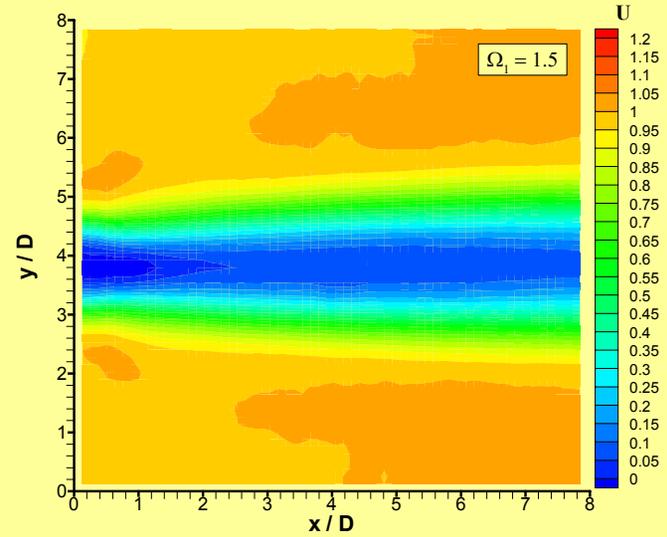
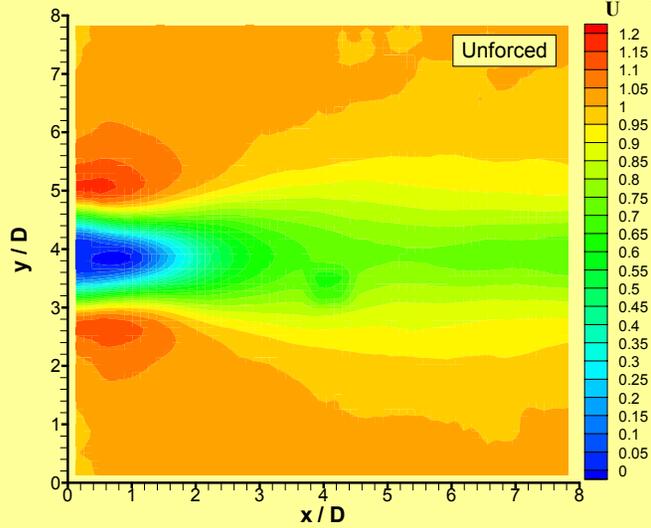
Instantaneous Vorticity Contours

$$S_f = 0.189$$

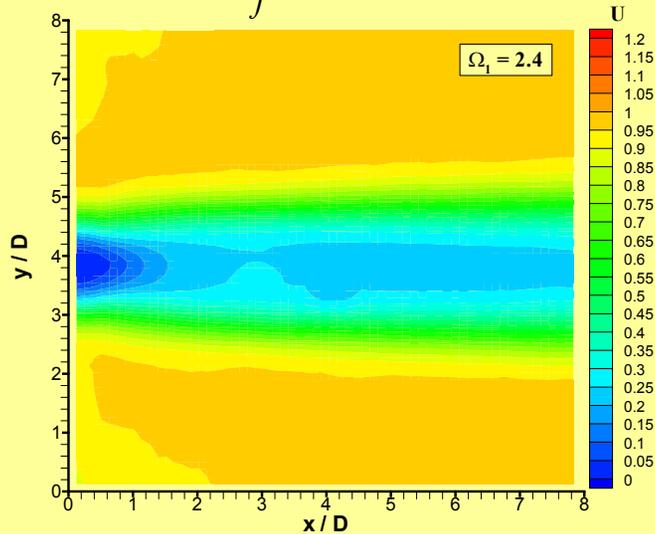


Mean Wake

$$S_f = 0.189$$



$$S_f = 0.248$$



Contours of streamwise velocity component

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Modeling Low-dimensional Flows

(Lumley, 1967, 1981)

- Following the work of Gillies (1998) and Graham et al. (1999), develop a physics-based model of cylinder wake
- Use a Galerkin method to obtain the model

$$u_i(x_j, t) = \sum_{k=1}^{\infty} a_k(t) \psi_i^k(x_j)$$

- Obtain a small set of characteristic velocity fields that possess the majority of the turbulent kinetic energy in the flow \Rightarrow Proper Orthogonal Decomposition
- Project the Navier-Stokes equations onto this set to obtain a low-dimensional model of the flow
- Use this model to recreate the basic features of the wake
 - Reconstruct the large-scale vortices in the wake
 - Gain insight to the flow dynamics
 - If the model is robust, we might attempt flow control

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Proper Orthogonal Decomposition

(Holmes, Berkooz & Lumley, 1996)

- Use POD to obtain modes for Galerkin projection
- POD is an *optimal* decomposition in that it will “capture the most energy in the fewest modes”
 - Maximize the average projection of u onto Ψ

$$\max \frac{\langle |(u, \psi)^2| \rangle}{\|\psi\|^2}$$

- From calculus of variations:

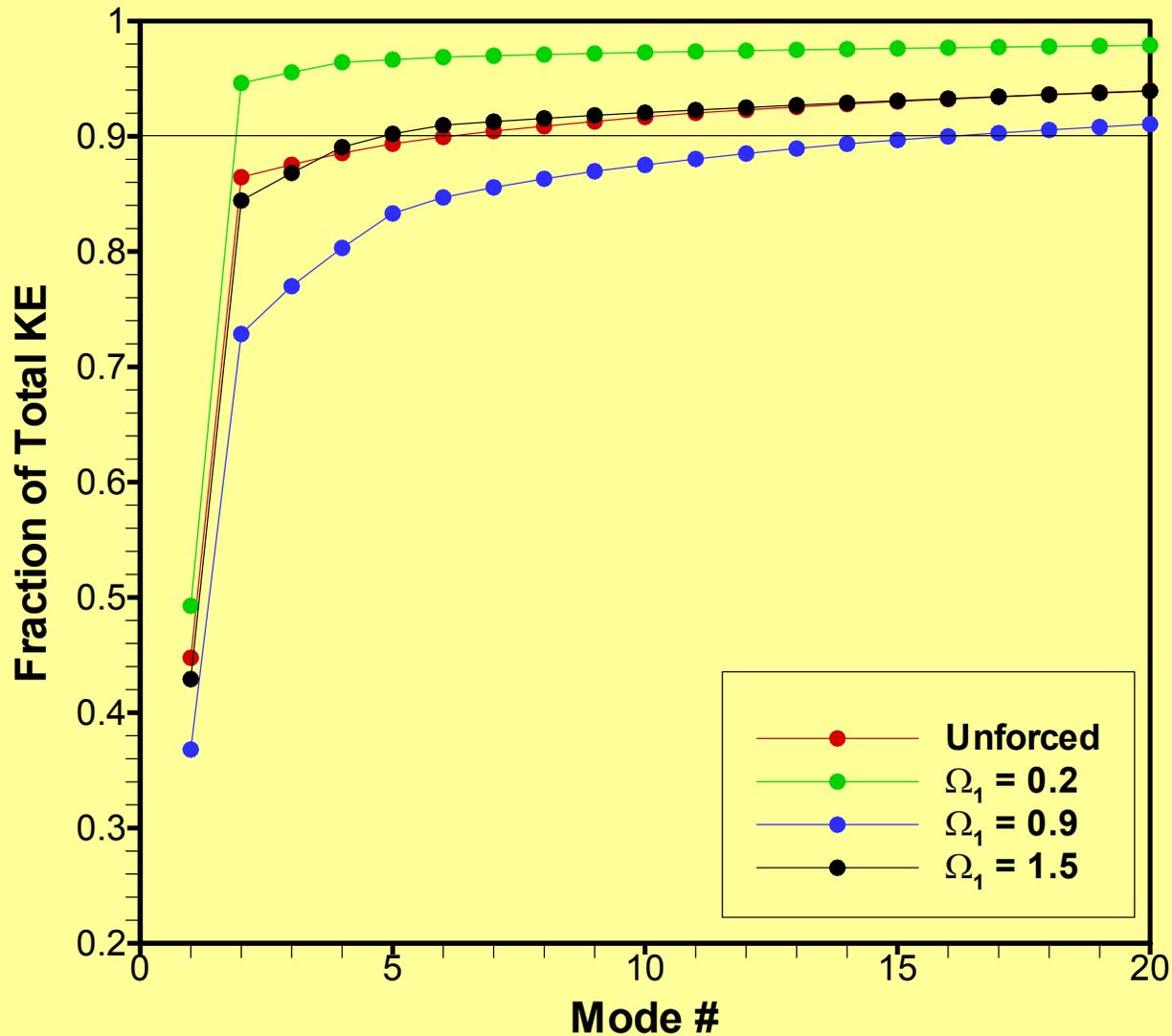
$$\int K(x, x') \psi(x') dx' = \lambda \psi(x)$$

where $K(x, x')$ is the two-pt velocity autocorrelation tensor, $\Psi(x)$ are the empirical eigenfunctions and λ are the empirical eigenvalues

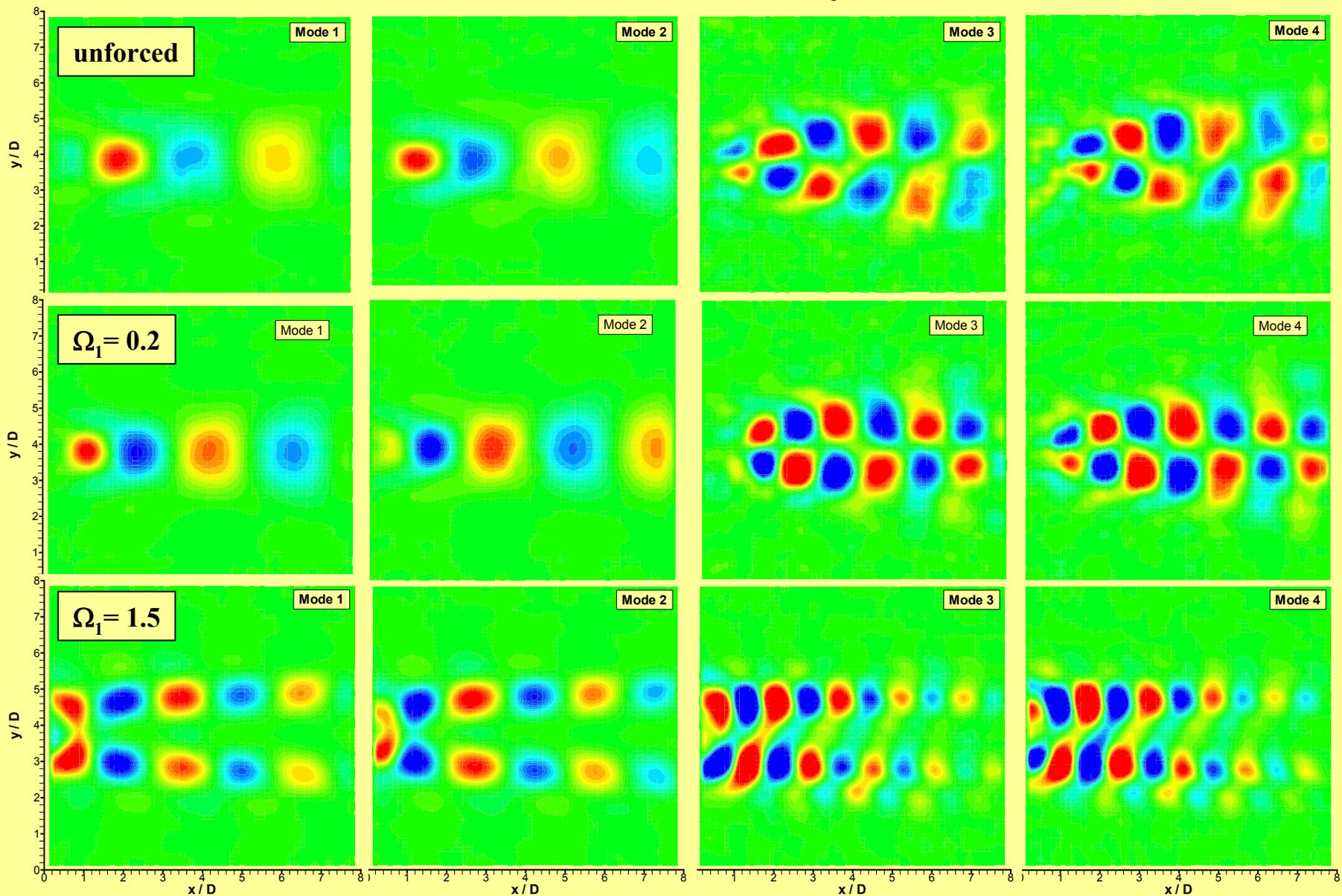
- Use “Method of Snapshots”, Sirovich (1987)

Modal Energy Distribution

$$S_f = 0.189$$



POD Modes – Vorticity Contours



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Galerkin Projection

- Modes identify the space in which to construct a model of the flow
- Project the Navier-Stokes equations onto this space to obtain our flow-field model

$$\left(\psi_i, \underbrace{\frac{\partial u_i}{\partial t} + \frac{\partial}{\partial x_j} \left(U_j u_i + u_j U_i + u_i u_j - \overline{u_i u_j} \right)}_{u_i(x_l, t) = \sum_{k=1}^N a_k(t) \psi_i^k(x_l)} \right) = \frac{1}{\text{Re}_D} \frac{\partial^2 u_i}{\partial x_j^2} \Bigg) \left. \vphantom{\frac{\partial u_i}{\partial t}} \right\} \frac{da_k}{dt} = F(a)$$

Navier-Stokes Eqns for fluctuation

where F is a non-linear function

- Order N of the model depends on how many modes we require to faithfully represent the large-scale turbulence

Low-Dimensional Model

- Model for a turbulent flow:

$$\frac{da_k}{dt} = -B^{kn} a_n - C^{knm} \left(a_n a_m - \overline{a_n^2} \right) + \frac{1}{\text{Re}_D} D^{kn} a_n$$

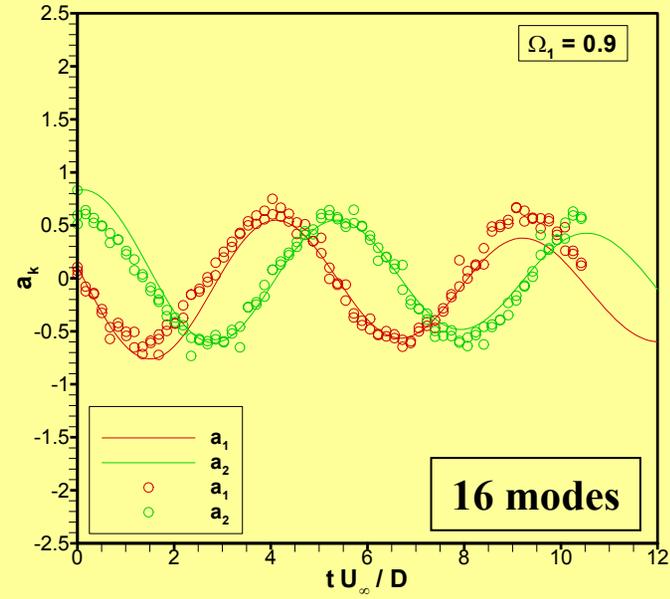
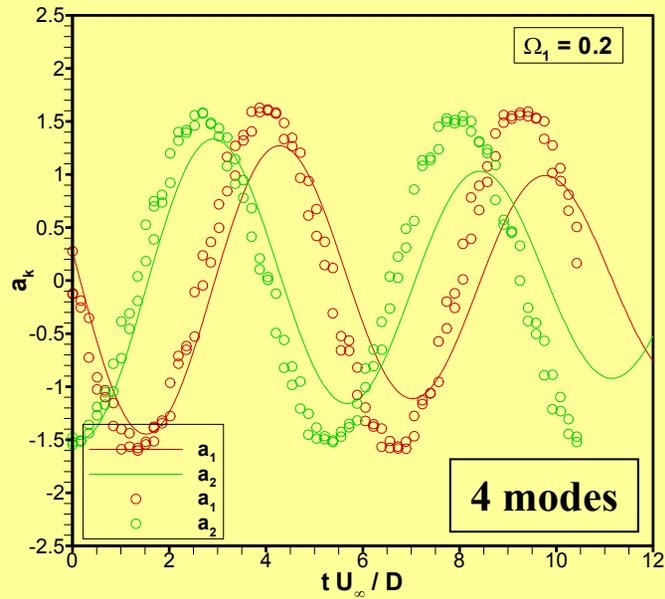
where $k = 1, 2, \dots, N$

$$B^{kn} = \left(\psi_i^k, U_j \frac{\partial \psi_i^n}{\partial x_j} + \psi_j^n \frac{\partial U_i}{\partial x_j} \right) \quad \longrightarrow \quad \text{Velocity-strain rate interactions}$$
$$C^{knm} = \left(\psi_i^k, \psi_j^n \frac{\partial \psi_i^m}{\partial x_j} \right) \quad \longrightarrow \quad \text{Fluctuating velocity interactions}$$
$$D^{kn} = \left(\psi_i^k, \frac{\partial^2 \psi_i^k}{\partial x_j^2} \right) \quad \longrightarrow \quad \text{Viscous term}$$

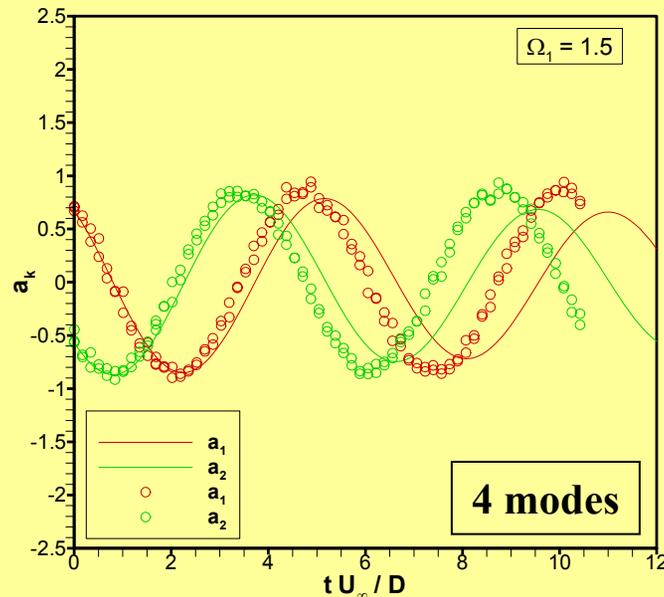
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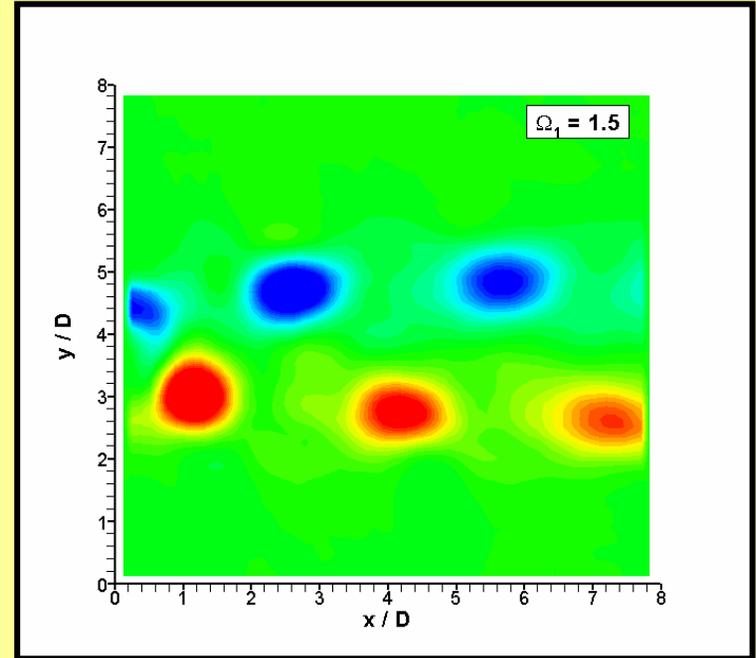
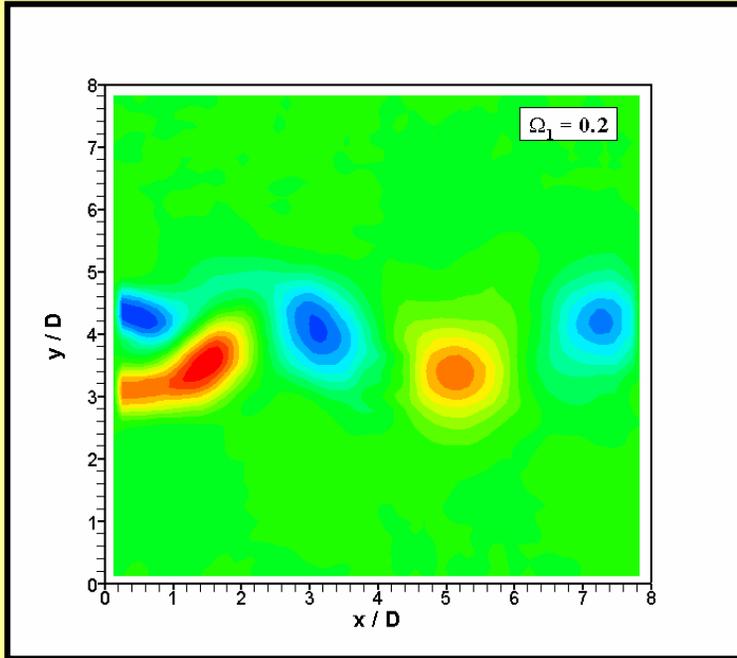
Evaluation of Model Predictions



- Symbols are projections of experimental data
- Lines are model predictions

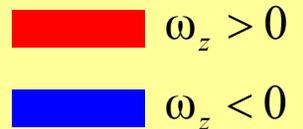


Reconstruction of the Wake



$$U_i(x_l, t) = \langle U_i(x_l) \rangle + u_i(x_l, t)$$

$$u_i(x_l, t) = a_1(t)\psi_i^1(x_l) + a_2(t)\psi_i^2(x_l) + a_3(t)\psi_i^3(x_l) + a_4(t)\psi_i^4(x_l)$$



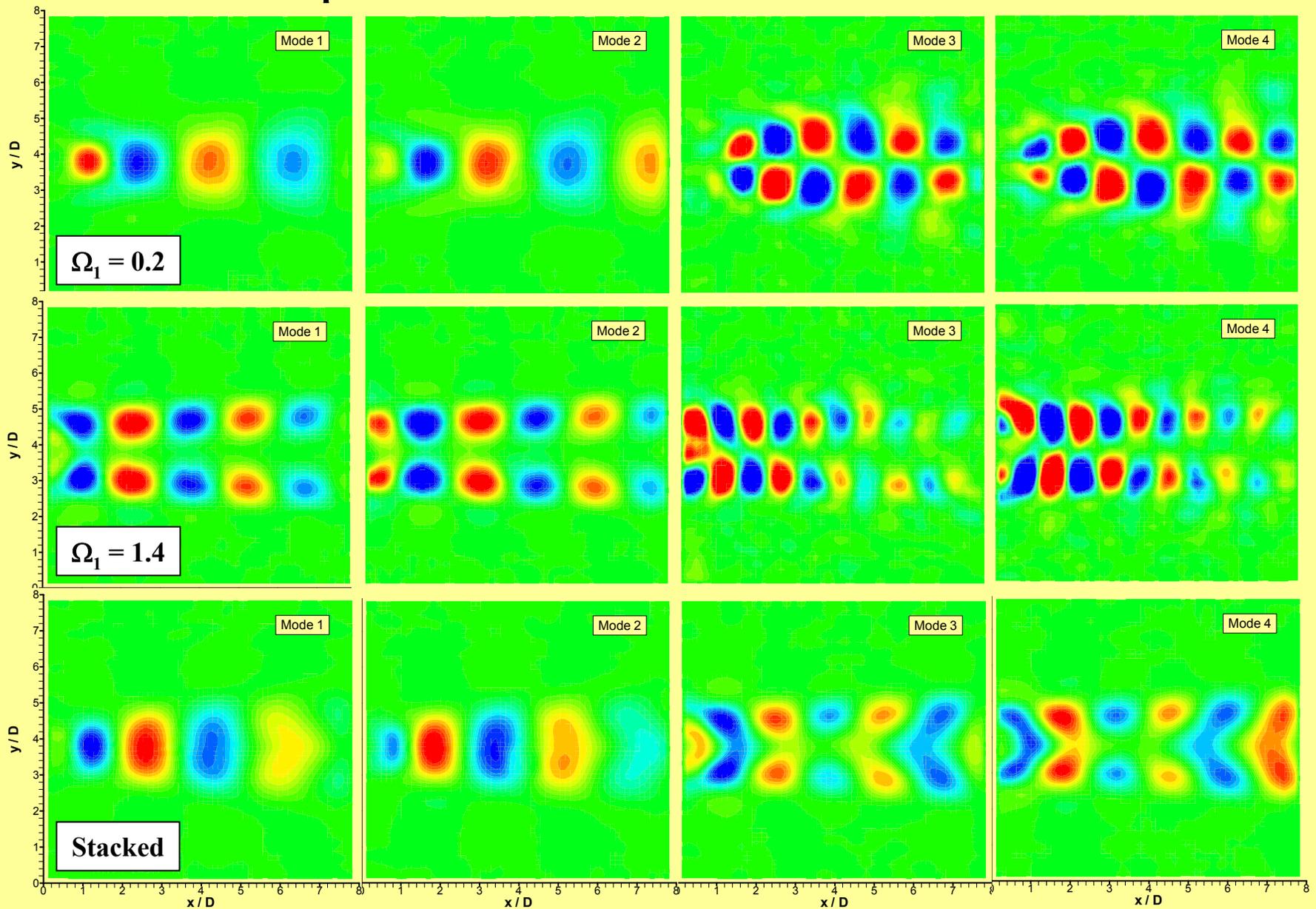
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Introduction to Data Stacking

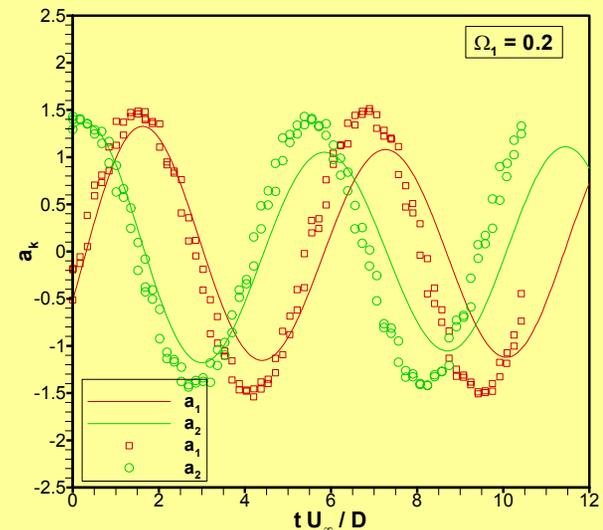
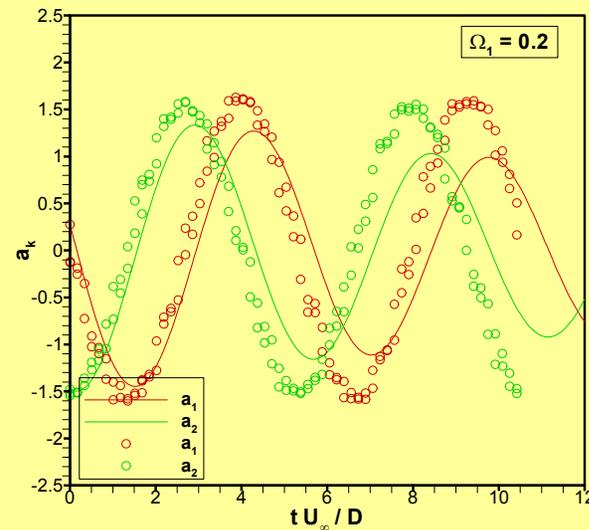
- POD analysis for $\Omega_1 = 0.2$ and 1.5 reveal distinct differences in the mode sets
 - A set of modes for one forcing condition may not well-represent the flow dynamics for a different forcing condition
 - How can we obtain a set of modes that spans a range of forcing conditions? *Use a stack of data sets from different Ω_1*
- A much larger set of data is used in the determination of the POD modes
 - Improves the velocity correlation statistics
 - Yields a basis that presumably is applicable to a wider variety of flow conditions
- Use data sets for $\Omega_1 = 0.2, 0.4, 0.6, 0.8, 1.0, 1.2$ and 1.4 to obtain the POD modes, then model one of the flows

Comparison of Stacked POD Modes

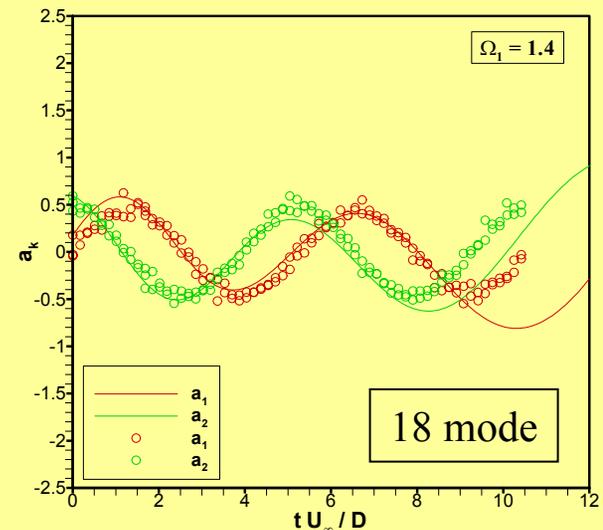
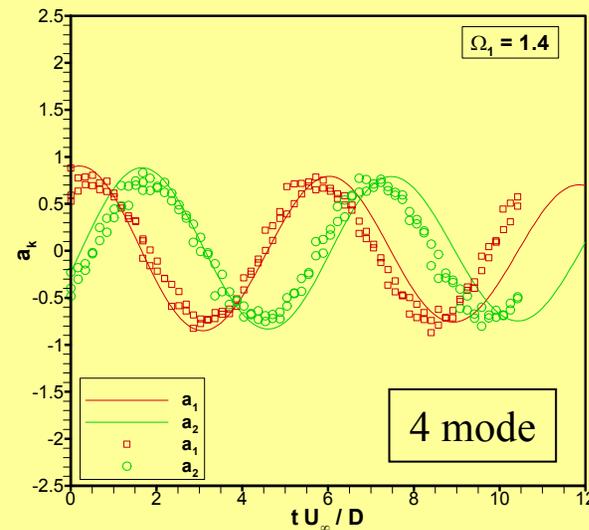


Comparison of Low-D Model Predictions

4 mode models



4 & 18 mode models

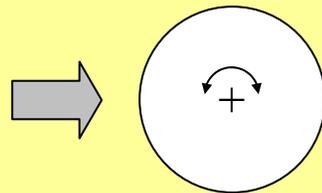


Single forcing modes

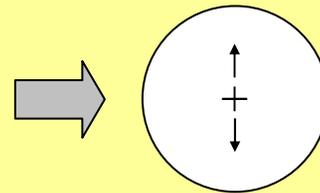
Stacked modes

Continuing Work

- Model development
 - Implement a low-dimensional model with control fn
 - Incorporate 3-D effects into the model
 - Better understand the wake response to oscillations at the cylinder
- Continuing experimental work: rotational/transverse oscillations



UW experiments



USAFA experiments

- Integrate CFD results into the analysis & modeling
- Demonstrate closed-loop control of the wake

Closed-loop Flow Control Program

