Proper orthogonal decomposition analysis and modelling of the wake deviation behind a squareback Ahmed body

Bérengère Podvin, Stéphanie Pellerin, and Yann Fraigneau LIMSI, CNRS, Université Paris-Saclay, Orsay, France

Antoine Evrard IMSIA-ENSTA, 828 Bd des Maréchaux F-91762 Palaiseau, France

Olivier Cadot

School of Engineering, University of Liverpool, Liverpool L69 3GH, United Kingdom

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We investigate numerically the three-dimensional (3D) flow around a squareback Ahmed body at Reynolds number $Re = 10^4$. Proper orthogonal decomposition (POD) is applied to a symmetry-augmented database in order to describe and model the flow dynamics. Comparison with experiments at a higher Reynolds number in a plane section of the near wake at midheight shows that the simulation captures several features of the experimental flow, in particular the antisymmetric quasisteady deviation mode. 3D POD analysis allows us to classify the different physical processes in terms of mode contribution to the kinetic energy over the entire domain. It is found that the dominant fluctuating mode on the entire domain corresponds to the 3D quasisteady wake deviation, and that its amplitude is well estimated from 2D near-wake data. The next most energetic flow fluctuations consist of vortex shedding and bubble pumping mechanisms. It is found that the amplitude of the deviation is negatively correlated with the intensity of the vortex shedding in the spanwise direction and the suction drag coefficient. Finally, we find that despite the slow convergence of the decomposition, a POD-based low-dimensional model reproduces the dynamics of the wake deviation observed experimentally, as well as the main characteristics of the global modes identified in the simulation.

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I. INTRODUCTION

A surprisingly generic feature of the flow around symmetric bodies at high Reynolds numbers is the presence of permanent symmetry-breaking structures in the wake. These have been observed for the sphere [1], bullet [2], the flat plate [3], as well as academic models for ground vehicles such as the Ahmed body [4] and the Windsor body [5]. The origin and dynamics of these structures have not been entirely elucidated, although they have been the object of several experimental and numerical studies. They appear to be connected to the first steady bifurcation observed at much lower Reynolds number (see Refs. [6,7]) and will be referred to as deviation modes throughout the paper. At higher Reynolds numbers, rapid switches between different quasistable states can be observed, as described by Refs. [8] and [9].

In the remainder of the paper, we will focus on the Ahmed body. Grandemange *et al.* [4] have established that the appearance of bi- or multistable states for the flow around a squareback Ahmed body depends on the value of the ground clearance and the aspect ratio of the body base. Pasquetti and Peres [10] carried out the first numerical simulation of the Ahmed body that was able to reproduce the steady wake deviation. However, switches were not observed in Pasquetti and Peres's

simulation, since the typical time separating two switches is on the order of 1000 U/H where U is the incoming flow speed and H the body characteristic height (see Ref. [11]). In more recent work Dalla Longa *et al.* [12] were able to integrate over sufficiently long times to capture the switch in the wake deviation. However, they could only observe one switch, due to the still longer time scale separating two switches. They also applied proper orthogonal decomposition (POD) and dynamic mode decomposition (DMD) [13] to extract the dynamics of the flow. Interaction of the near-wake recirculation zone with the surrounding shear layers is expected to play a part in the occurrence of switches. They proposed that the switch is triggered by large hairpin vortices.

Grandemange *et al.* [4] characterized two periodic phenomena by carrying out a spectral analysis in the wake of the squareback Ahmed body. Two well-defined synchronized frequencies were identified in their experimental data corresponding to vortex shedding modes with large and small Strouhal numbers, respectively, associated with the width and the height of the body. Volpe *et al.* confirmed the presence of these two frequencies with Strouhal numbers of 0.13 and 0.19. In addition they identified a wake-pumping motion in the low-frequency range around a Strouhal number of 0.08. Pavia *et al.* [14] have applied POD to both the experimental pressure and velocity field of a Windsor body to identify structures corresponding to the bistable modes and the dynamics of the switch. They identified pumping motion in the wake with a bistable vortical structure in the streamwise direction and derived a phase-averaged model that describes the switch between the quasistable states. They confirmed the observation made by Evrard *et al.* [15] and Cadot *et al.* [3] that the symmetric state corresponds to a lower drag. The topology of the flow was confirmed by a recent study by the same group [16] based on volumetric POD analysis of the full experimental velocity field behind a Windsor body.

Drag reduction through control of the flow asymmetry has been the object of several passive and active control strategies [9,11,17–21]. An important question is to determine how the different structures present in the flow contribute to the drag. A low-dimensional description and modelling of the large-scale structures could be beneficial for understanding, predicting, and ultimately controlling the flow dynamics. Reduced-order models were developed by Refs. [8,9] in order to control the deviation of the wake in experiments. A general form is assumed *a priori* for these models based on limited information about the flow physics. In contrast, full information is available in numerical studies.

The objectives of the paper are threefold. The first is to compare classical 2D POD obtained with experiment to the 3D POD available with the simulation. The second goal is to use 3D POD to provide a hierarchy of the global wake dynamics that contribute the most to the aerodynamic drag. These global dynamics thoroughly identified in the literature are the pumping motion, the two vortex shedding modes, and the deviation mode. The third goal of the paper is to derive, in a manner as objective as possible, a model able to reproduce the dynamics of the large scales. This is achieved through Galerkin projection of the Navier-Stokes equations onto a basis of POD eigenfunctions.

The numerical configuration considered here consists of the flow around a squareback Ahmed body at a Reynolds number of $Re = 10^4$. The spanwise to vertical aspect ratio of the body is 1.18, so that bistability corresponds to two deviation modes, which are symmetric through reflection of the vertical midplane. Due to the long time scales separating switches, the change in the wake deviation could not be captured by the simulation. However, reflection symmetry was enforced artificially to build the POD modes, so that we provide a description of the structures of the flow as a superposition of reflection-symmetric and reflection-antisymmetric POD modes.

The paper is organized as follows. Section II presents the 3D numerical configuration, while Sec. III describes the specifics of proper orthogonal decomposition. Section IV presents a 2D comparison of the numerical simulation and an experimental configuration corresponding to the same geometry but a higher Reynolds number. Section V presents a 3D POD analysis of the structures in the full configuration. A POD-based low-dimensional model is constructed in Sec. VI and compared with experimental and numerical results. A conclusion is given in Sec. VII.



FIG. 1. Numerical configuration. The streamwise, spanwise, and wall-normal positions are referred to as x, y, z. The length, width, and height of the body are respectively L, W, and H. The origin of the axes (indicated in red) is taken at the top and foremost position of the Ahmed body in the vertical symmetry plane.

II. NUMERICAL CONFIGURATION

Figure 1 presents the numerical configuration. The dimensions of the squareback Ahmed body are the same as in the experimental configuration of Evrard *et al.* [15], i.e., L = 1.124m, H = 0.297m, W = 0.35m. We note that there is no support in the numerical simulation and the ground clearance (distance from the body to the lower boundary of the domain) in the experiment is C = 0.13 H while it is 0.3 H in the simulation. The Reynolds number based on the incoming velocity U, body height, H and fluid viscosity is 10^4 . The foremost and upper part body defines the reference position (x = 0, z = 0). The plane y = 0 corresponds to the midplane of the body. In what follows we will focus on the flow in the midheight plane of the body. The domain extent in the streamwise (x), spanwise (y), and vertical (z) directions is $[-1, 10]H \times [-2, 2]H \times [-1.3, 1.2]H$.

The code used is SUNFLUIDH, an in-house code, which is based on a second-order finite volume approach, which has been described for instance in Ref. [22]. The temporal discretization is based on a second-order backward Euler scheme. Diffusion terms are treated implicitly and convective terms are solved with an Adams-Bashforth scheme. The Poisson equation for the computation of pressure field is solved iteratively. We use $(512 \times 256 \times 256)$ grid points in, respectively, the longitudinal direction x, the spanwise direction y, and the vertical direction z. The Cartesian grid is refined close to surfaces. Periodic boundary conditions are used in the spanwise direction. No-slip velocity boundary conditions on the Ahmed body and on the ground are implemented on the mesh cells corresponding to the physical boundaries. The simulation was initialized from a uniform condition. About 100 time units based on the upstream velocity U and body height H were necessary for the flow to develop and statistical convergence to be reached. We note that all times will be expressed in those units in the remainder of the paper. All lengths will be made nondimensional with the body height H.

III. PROPER ORTHOGONAL DECOMPOSITION

The main tool of analysis used in this paper is proper orthogonal decomposition (POD) [23]. On a domain Ω , the field <u>u</u> is written as a superposition of spatial modes ϕ_{μ}

$$\underline{u}(\underline{x},t) = \sum_{n} \tilde{a}_{n}(t) \underline{\phi}_{n}(\underline{x}), \tag{1}$$

where the spatial modes are orthogonal (and can be made orthonormal), i.e.,

$$\int \underline{\phi}_n(\underline{x}) \cdot \underline{\phi}_m(\underline{x}) d\underline{x} = \delta_{nm}$$

Here we have used the standard inner product definition, although other choices can be made [24]. The modes can be ordered by their magnitude. We have $\lambda_1 \ge \lambda_2 \ge \ldots \ge \lambda_n = \langle \tilde{a}_n \tilde{a}_n \rangle$, where $\langle . \rangle$ represents a time average. The amplitudes \tilde{a}_n can be obtained by projection of the velocity field onto the spatial modes:

$$\tilde{a}_n(t) = \int_{\Omega} \underline{u}(\underline{x}, t) \underline{\phi}_n(\underline{x}) d\underline{x}.$$
(2)

In the remainder of the paper we will consider normalized amplitudes $a_n = \tilde{a}_n / \sqrt{\lambda_n}$.

In all that follows POD is implemented following the method of snapshots [25], which is based on computing the autocorrelation between the different samples of the field obtained at times t_i , i = 1, ..., N:

$$C_{ij} = \frac{1}{N} \int_{\Omega} \underline{u}(\underline{x}, t_i) . \underline{u}(\underline{x}', t_j) d\underline{x},$$

and extracting the eigenvalues λ_n and temporal eigenvectors $A_{in} = a_n(t_i)$ such that

$$CA = \lambda A.$$
 (3)

The spatial modes can then be reconstructed using the following rule in tensor notation

$$\underline{\phi}_{n}(\underline{x}) = u(\underline{x}, t_{i})A_{\text{in}} \tag{4}$$

and renormalizing $\underline{\phi}_n(\underline{x}) = \frac{1}{N_n} \underline{\tilde{\phi}}_n(\underline{x})$ where

$$N_n^2 = \int \underline{\tilde{\phi}}_n(\underline{x}) \cdot \underline{\tilde{\phi}}_n(\underline{x}) d\underline{x}.$$
 (5)

Different spatial domains, as well as different quantities will be considered in the decomposition. In Sec. IV, we will limit the analysis to 2D velocity fields in the near-wake region in order to match the data of Evrard *et al.* [15]. In Sec. V, we will apply POD to the full 3D velocity field in the full numerical domain.

A key ingredient of the procedure is the definition of the data set. Cross sections of the timeaveraged streamwise velocity are represented in Fig. 2 for two planes: one is the midheight plane of the base z = -0.5 (the origin is located at the rooftop) and the other is the vertical midspan plane y = 0. We observe a steady deviation (or asymmetry) of the wake in the horizontal plane, with stronger fluctuations for y > 0 in agreement with previous observations [10]. This deviation is breaking the reflection symmetry with respect to the vertical midplane. However, the equations are symmetric; for each flow realization, the flow obtained by symmetry with respect to the vertical midplane is also a possible solution. One has

$$u_i(x, y, z, t) \rightarrow u_i(x, -y, z, t)\eta_i,$$

where $\eta_i = (-1)^{i+1}$. Following recommendations in Ref. [26], an enlarged data set enforcing the statistical symmetry could then be created as follows: for each snapshot of the original data set, a symmetrized snapshot corresponding to the image of the snapshot by the reflection symmetry was



FIG. 2. Streamwise velocity contours of the mean flow; left) horizontal plane at body midheight z = -0.5; right) vertical midplane at y = 0.

created. The size of the new data set was therefore twice that of the original one. By construction (see the consequences of symmetry for POD eigenfunctions in Ref. [26]), POD modes are thus either symmetric or antisymmetric with respect to the vertical midplane y = 0, so that the amplitude of a POD mode for a given snapshot of the original data set is either identical or opposite to that on the symmetrized snapshot. This allows us to break down flow patterns into symmetric and antisymmetric components.

We emphasize that individual POD modes are different from coherent structures. A coherent structure typically corresponds to local patterns identified in a realization. POD modes are defined over the full domain chosen for decomposition, and every single realization contains a combination of modes. A coherent structure will therefore correspond to a combination of a few POD modes, which may be restricted to a portion of a spatial domain. As a consequence of the data enlargement procedure we have adopted, the mean flow of the simulation does not correspond to one single mode but to the sum of the first two modes representing, respectively, the symmetric and the antisymmetric part of the mean flow.

We note that the data does not contain information about the switching process, however, since they are of small duration, we expect that the most energetic POD modes obtained with the symmetrized data set are not very different from those that would be found in a long simulation, which would include switches between quasisteady states. This was found to be the case for a turbulent Rayleigh-Bénard convection flow displaying intermittent reversals [27].

IV. 2D POD: COMPARISON WITH THE EXPERIMENT

In this section we apply POD analysis in the near-wake to the same variables defined over the same domain in both the experiment and the numerical simulation, i.e., the full (i.e., including the mean) streamwise and the spanwise velocity components u and v defined over the domain $L \le x \le L + 1.2H$, $-0.6 \le y \le 0.6H$. Details are indicated in Table I. The main differences are

(i) the Reynolds numbers considered, as $Re = 4 \ 10^5$ in Evrard's experiment and $Re = 10^4$ in our numerical simulation;

(ii) the time separation between two snapshots, which is 25H/U in the experiment versus 0.5H/U in the simulation. The separation time is 50 times larger in the experiment than in the DNS, and the total length window is about 80 times larger in the experiment than in the simulation.

TABLE I. Comparison between model and experiment. All times are made nondimensional with the upstream velocity and body height. Δt is the separation between two snapshots. N is the number of original snapshots (original or unsymmetrized database).

Туре	Re	Δt	Ν	$L_{\rm window} = N\Delta t$
Simulation	$10^4 \\ 4 \times 10^5$	0.5	280	140
Experiment		25.25	400	10000



FIG. 3. Near-wake 2D POD spectrum in the simulation and experiment.

Results for the POD spectrum are shown in Fig. 3 and show a good agreement between the experiment and the simulation. The second mode represents about 35% of the total fluctuating horizontal kinetic energy (the sum of the squares of the streamwise and the spanwise components) in the wake $\sum_{p\geq 2} \lambda_p$ and the first eight modes capture more than 50%. Figure 4 compares the first four POD modes for the experiment and the simulation. The first mode represents a cross section of a toroidal-like structure constituting the recirculation bubble. The second mode corresponds to a single vortical structure located in the recirculation bubble close to the rear of the body and sweeping fluid from one side of the wake to the other. It represents the wake deviation. Overall a good agreement is observed between all the modes found in the experiment and those found in the simulation.

We can also see that mode 3 is symmetric and mode 4 is antisymmetric. Mode 3 consists of a longitudinal motion at the extremity of the recirculation bubble, which converges towards (or diverges away from, depending on the sign of the amplitude) the vertical midplane. This motion could be associated with wake pumping, i.e., successive enlargement and shrinking of the



FIG. 4. 2D wake POD modes 1 to 4; top row: simulation; bottom row: experiment.



FIG. 5. Amplitudes of 2D POD modes 1 to 4 (from top to bottom); (left) simulation; (right) experiment.

recirculation region. Mode 4 consists of three vortical structures: one larger structure extending across the wake and located close to the rear of the body, and two smaller ones, both rotating in the opposite direction, on each side of the recirculating bubble. Its action is therefore to distort the recirculation bubble in the spanwise direction.

Figure 5 shows for both the experiment and the simulation the corresponding amplitudes (normalized by the square root of the eigenvalue) of the modes represented in Fig. 4. The amplitudes of the first two modes in the simulation (Fig. 5, left) display small oscillations around a constant positive value. In the experiment (Fig. 5, right), the amplitude of the first mode oscillates near a constant positive value, while the amplitude of the second mode changes sign several times, which may correspond to a switch in the wake asymmetry, as shown by Ref. [28].

Since the PIV results are not resolved in time and the simulation total time is relatively short, it is difficult to compare directly time evolutions. However histograms of the amplitudes can be compared in Fig. 6. We can see that there is a good agreement between the experiment and the simulation, which suggests that the flow dynamics are similar despite the disparity in Reynolds number.

The next four modes in the simulation, shown in Fig. 7, also present similarities with the modes in the experiment. Modes 5 and 8 are antisymmetric and consist of vortical motions respectively dominant on the inner part and the outer part of the recirculation. Modes 6 and 7 are symmetric and consist of two counterrotating vortical structures, extending over the full recirculation length and located on each side of the wake. Given the differences in Reynolds number and time resolution, the agreement is remarkable and suggests that the most energetic structures have common features over a wide range of Reynolds numbers.



FIG. 6. Histogram comparison of 2D POD modes 1 to 4 (from left to right); top row: simulation; bottom row: experiment.

Due to the low temporal resolution of the PIV measurements, the numerical and the experimental frequency domains are not comparable in Fourier space. Thus, spectral analysis of the experimental data is not shown. The spectral content of the temporal coefficients in the numerical simulation is presented in Fig. 8. All modes are characterized by low frequencies. The two red and black lines, respectively, correspond to the reference frequencies 0.08 and 0.2. Modes 3 and 7, which are both symmetric, are characterized by a dominant frequency around 0.08, while mode 4, 5, and 6, which are antisymmetric, are characterized by a frequency of 0.2. This frequency, which is associated with vortex shedding, can be identified most clearly in modes 5 and 6. Mode 8 is characterized by a mixture of frequencies.

V. 3D POD

To investigate the flow structure and dynamics in more detail, POD is applied to the full 3D velocity field over the entire computational domain with the same symmetrization procedure as described earlier. The eigenvalue spectrum is shown in Fig. 9. The first 3D mode, which coincides with the mean flow (as can be checked from examination of the corresponding amplitude a_1 , which



FIG. 7. 2D wake POD modes 5 to 8; top row: simulation; bottom row: experiment.



FIG. 8. 2D POD amplitude power spectral density of the amplitudes in the numerical simulation $|\hat{a}_n|^2$ for modes 3–8; the red and black dashed lines, respectively, correspond to the frequencies 0.08 and 0.2.

is constant for all snapshots; see also Ref. [26] for more details), is much more energetic than the other modes compared with 2D analysis (Fig. 3) since the entire domain contains a large steady contribution of the kinetic energy outside the wake. Due to the extent of the spatial domain, the convergence of the fluctuations is much slower than in the 2D wake measurements. The second



FIG. 9. 3D POD eigenvalue spectrum.



FIG. 10. Streamwise velocity contours on body midheight plane z = -0.5; top: projection of an instantaneous field \underline{u} onto the first ten modes $u_{\text{recons}} = \sum_{n=1}^{10} a_n(t) \underline{\phi}(\underline{x})$ where $a_n = \int \underline{u}(\underline{x}, t) \cdot \underline{\phi}(\underline{x}) d\underline{x}$; bottom: difference field $|u - u_{\text{recons}}|$.

mode represents only 8% of the total fluctuating kinetic energy $\sum_{p \ge 2} \lambda_p$, which is equivalent to the combined energy of the next eight most energetic modes. We note that the modes 3–6 have nearly similar eigenvalues, which are about 10% of mode 2. This is in good agreement with the 3D POD experimental results of Ref. [16].

Generally speaking, the physical structures are three-component modes extending over the full domain, linking the upstream flow, the four boundary layers along the body, the near and the far wake. We reconstructed an instantaneous velocity field from its projection on the first ten POD modes, which capture about 20% of the total fluctuating energy. The reconstruction is shown in Fig. 10, along with the difference between the full instantaneous field and the reconstructed field. It is clear that only the large scales are captured by the projection. One can see that most of the unresolved modes (which represent the major portion of the total kinetic energy) are located in the boundary layers, the shear layers, and the far wake.

We now investigate the properties of the POD spatial modes. Spatial characteristics of the first two 3D POD modes identified by the streamwise velocity contour are represented in Fig. 11. Each figure displays the streamwise velocity of the 3D POD mode in both a horizontal cross section at midheight of the body (z = -0.5), and in a vertical plane located in an off-center spanwise location. The location of the vertical plane was chosen at y = -0.4 since asymmetric motions will cancel on the symmetry plane. Although the representation of the first two POD modes in 2D and 3D is different (Figs. 4 and 11), it can be seen that the restriction of the first two 3D POD modes to the cross section of the wake is similar to the 2D POD results, which is not entirely surprising since they are expected to represent the symmetric and antisymmetric part of the time-averaged velocity field. This can also be seen from the similarity between the temporal amplitudes in two dimensions and in three dimensions.

One can see in Fig. 12 that the temporal evolution of the first two 3D POD modes is slightly different from that of the 2D POD coefficients. Unlike its 2D counterpart (Fig. 5, right), the amplitude of the first 3D POD mode is essentially constant, which shows the statistical convergence of the database. Moreover, the correlation between the amplitudes of the 2D and 3D modes a_2^{2d} and a_2^{3d} is 0.6, which indicates that the 2D measurements are able to describe to some extent the evolution of the 3D deviation mode. However, although the amplitude of 2D POD mode 2



FIG. 11. Streamwise velocity contours of 3D POD modes 1 (top row) and 2 (bottom row); (left) horizontal section on body mid-height plane z = -0.5; (right) vertical section at y = -0.4.

changes sign, the corresponding 3D POD amplitude a_2^{3d} always remains positive. This means that there are no full switches in the total wake deviation although a local planar measure may indicate otherwise. In both cases, their corresponding spectrum indicates fluctuations at low frequencies thus confirming their very long-time dynamics evolution. In addition to its permanent asymmetry, the POD mode 2 corresponds to the very low-frequency deviation mode as reported in Refs. [8,19] for an axisymmetric body and which was shown by [1,4] to be responsible for the bistable dynamics. To get a better insight into the action of the deviation mode, Fig. 13 compares streamlines in the recirculation zone for mode 1, which corresponds to the symmetric part of the time-averaged field, and the sum of modes 1 and 2, which corresponds to the time-averaged field. We can see that the effect of the deviation mode is to gather the streamlines around one of the base diagonals.

Figure 14 shows that modes 3 and 4 are antisymmetric modes, which display a periodicity in the streamwise direction. The time evolution of these modes is quasiperiodic as shown by the time series and the spectra in Fig. 15. The characteristics frequencies lie in the range 0.19–0.24,



FIG. 12. (Left) amplitudes of the first two POD modes; (right) power spectral density of the quasisteady deviation mode $|\hat{a}_2|^2$; the red and black lines, respectively, correspond to the two frequencies 0.08 and 0.2.



FIG. 13. Streamlines of the field viewed from downstream (left) mode $1 \sqrt{\lambda_1} \underline{\phi}_1$; (right) combination of mode 1 and mode $2 \sqrt{\lambda_1} \underline{\phi}_1 + \sqrt{\lambda_2} \underline{\phi}_2$.

with a maximum around 0.19. This frequency matches satisfactorily the vortex shedding mode observed for oscillations in the spanwise direction in Grandemange *et al.* [1], Volpe *et al.* [29]. Similar observations are made for the symmetric modes 5 and 6 in Fig. 14 and the corresponding temporal properties in Fig. 15. Their spectra display a maximum around 0.23, that is ascribed to the vortex shedding mode with oscillations in the vertical direction. The existence of two frequencies, a higher one corresponding to the vertical (smaller) and a lower one corresponding to the spanwise (larger) dimension of the body is in agreement with Grandemange *et al.* [1], Volpe *et al.* [29]. This is a general result for three-dimensional geometries (Kiya and Abe [30]). It is possible to measure the magnitude of vortex shedding in, respectively, the horizontal (r_{Kh}) and the vertical (r_{Kv}) direction by adding the energies of the corresponding modes so that $r_{Kh}^2 = a_3^2 + a_4^2$ and $r_{Kv}^2 = a_5^2 + a_6^2$.

The modes 7 and 8 in Fig. 14 are symmetric and clearly do not display streamwise periodicity. The action of mode 7 in the horizontal mid-plane is to modulate the bubble zone which is either inflated or shrunk, depending on the sign of a_7 (Fig. 15, left). When $a_7 > 0$, Fig. 14 shows that strong negative fluctuations are present in the bubble, which delays reattachment, and vice versa. The spectra shown in Fig. 15 (right) show that both modes 7 and 8 are characterized by a strong energetic content at a frequency of 0.08, which suggests that these modes contribute significantly to the wake pumping, as characterized by Rigas *et al.* [2], Volpe *et al.* [29], and Pavia *et al.* [14,31].

Modes 9 and 10, which are antisymmetric (Fig. 14), are more difficult to interpret. As noted earlier, there is no guarantee that an individual POD mode corresponds to a well-defined physical mechanism. The spectra in Fig. 15 (right) show that modes 9 and 10 are characterized by a mixture of frequencies in an intermediate range 0.08–0.2, with a peak for both modes around 0.15. It is therefore likely that these modes correspond to a superposition of different physical processes.

As a summary, we find some correspondence between the POD modes and the main global modes that contribute to the wake dynamics reported in the literature: POD mode 2 is related to the very low-frequency deviation global mode, modes 3 and 4 to vortex shedding with oscillation in the horizontal direction, modes 5 and 6 to vortex shedding with oscillations in the vertical direction, and modes 7 and 8 to (symmetric) wake pumping.

The POD decomposition allows to investigate quantitatively the correlations between the temporal amplitudes of the modes associated with the global wake dynamics. We first compare the intensity of the deviation mode given by (a_2^2) with the magnitudes of the shedding mode in the horizontal (spanwise) and the vertical direction. We can see in Fig. 16 that minima of a_2^2 are associated with high energy in horizontal vortex shedding. The correlation coefficient between a_2^2 and r_{Kv}^2 is strongly negative (-0.6). Due to the three-dimensional nature of vortex shedding, horizontal and vertical shedding motions are correlated, so that a negative correlation is also obtained between a_2^2 and r_{Kh}^2 (-0.33). The negative correlation is consistent with the idea that a



FIG. 14. Streamwise velocity contours of 3D POD spatial modes 3–10 (from top to bottom); (left) horizontal section on body midheight plane y = -0.5; (right) vertical section at z = -0.4.



FIG. 15. (Left) amplitudes of the modes a_n , n = 3, 5, 7, 9; (right) power spectral density of the 3D POD mode amplitudes in the simulation $|\hat{a}_n|^2$; the red and black lines, respectively, correspond to the two frequencies 0.08 and 0.2.

reduced deviation or asymmetry is accompanied by an increase in vortex shedding, which was observed in control experiments of Refs. [9,18], as well as in Refs. [14,16].

The next step is to compare the evolution of the POD amplitudes with the base suction coefficient $C_B = -C_{pb}$ where the pressure coefficient C_{pb} corresponds to the integral of the pressure over the base of the body, which is shown in Fig. 17 (left). Figure 17 (right) shows that the base suction coefficient is positively correlated with the amplitude of the wake deviation a_2^{3d} with a positive delay of $\Delta t U/H \sim 2.5$ and with a maximum correlation coefficient of 0.55.

The time delay appears to have some significance as the correlation without time delay drops to 0.34. This suggests that the variations of the drag follow those of the mean deviation amplitude. This is in agreement with the idea that an increase of the wake symmetry is followed by a pressure change at the body base reducing the drag. The correlation between wake asymmetry and drag has already been reported in Refs. [3,14], but the present study derives the cause and effect relationship with a time delay of the order of the convective time. Figure 17 also shows that



FIG. 16. (Left) energy of POD modes corresponding to quasisteady deviation a_2^2 , spanwise $(r_{Kh}^2 = a_3^2 + a_4^2)$ and vertical $(r_{Kv}^2 = a_5^2 + a_6^2)$ vortex shedding intensities; vertical lines correspond to minima of a_2^2 ; (right) correlation coefficient between vortex shedding energy and quasisteady deviation energy. The largest value of the correlation coefficient is -0.6 at zero time delay and is related with spanwise vortex shedding.



FIG. 17. (Left) base suction coefficient $C_B = -C_p$ obtained by integrating the pressure over the rear of the body; (right) correlation between the base suction coefficient and the POD amplitudes corresponding to the steady deviation (a_2 , top), wake pumping mode (a_7 , middle), and vortex shedding modes ($r_{Kh,v}^2$, bottom). The time delays corresponding to vertical lines in the figure correspond to a value of four time units for a_2 and -1.5 for a_7 .

a correlation coefficient of -0.55 with a negative delay of $\Delta t U/H \sim -1.5$ is observed between the pressure and the amplitude of a_7 . Unlike the previous case, the correlation remains about the same without no time delay (-0.51), so it is not possible to assign a physical relevance to the small time delay observed. As seen above, a_7 is associated with wake pumping: from Fig. 14 one can see in the near wake that when $a_7 > 0$ there are more negative fluctuations within the zone, so that the size of the recirculation actually increases. This is consistent with the observation that the base suction coefficient decreases as the recirculation length increases [32]. In contrast, the correlation of the drag with the intensity of vortex shedding appears to be weaker (and negative).

The observations made above suggest the following picture: the flow is characterized by a quasisteady wake deviation, vortex shedding modes, and low-frequency wake-pumping motion. The drag coefficient depends on the global (symmetric) size of the bubble, which is associated with wake pumping, as a longer bubble corresponds to a lower drag. It also depends on the magnitude of the deviation mode: a lower drag corresponds to a decrease of asymmetry, and to an increase in the intensity of vortex shedding.

VI. LOW-DIMENSIONAL MODEL

We now examine whether it is possible to model the behavior of the largest scales using a PODbased model, even if the scales considered represent only a fraction of the total fluctuating kinetic energy. Following the approach described in Refs. [27,33], we build a low-dimensional model to reproduce the dynamics. We use a Galerkin approach to project the Navier-Stokes equations onto the basis of spatial modes for a selected truncation, and obtain a set of ordinary differential equations

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n	1	2	3	4	5	6	7	8	9	10
$\overline{L_i}$		-0.05	-0.05	-0.05	-0.05	-0.05	-0.05	-0.05	-0.05	-0.05
$\langle a_n^2 \rangle_{M_0}$	1	1	0.93	0.93	0.86	0.86	1.45	1.51	1.05	1.05
$\langle a_n^2 \rangle_M$	1	1.05	0.29	0.29	0.18	0.18	1.01	1.03	2.05	2.33

TABLE II. Model linear coefficients and predicted energy.

for the normalized amplitudes $a_n(t)$. Using tensor notation, the model is of the form

$$\dot{a}_n = L_{nm}a_m + Q_{nmp}a_ma_p + T_n,\tag{6}$$

where

(i) the linear terms contain the viscous dissipation

$$L_{nm} = \int v \Delta \underline{\phi}_{m} \underline{\phi}_{n} d\underline{x}$$
⁽⁷⁾

and are indicated in Table II. For the ten-mode truncation they form a diagonal matrix $L \sim -0.05I$. The diagonal form of the matrix was also observed in channel flow [33].

(ii) T_n is a closure term representing the contribution of the unresolved stresses (associated with the modes excluded from the truncation) to the evolution of the amplitude a_n ;

(iii) the quadratic terms Q_{nmp} are obtained by projection of the nonlinear term, which yields contributions of the form

$$Q_{mpn}^{T} = \frac{\sqrt{\lambda_{m}\lambda_{p}}}{\lambda_{n}} \int (\underline{\phi}_{p} \cdot \nabla \underline{\phi}_{m})) \cdot \underline{\phi}_{n} d\underline{x}.$$
(8)

In order to obtain a symmetric expression in a_m and a_p , we therefore define the coefficient Q_{mpn} as follows:

$$Q_{nmp} = \frac{\sqrt{\lambda_m \lambda_p}}{\lambda_n} \frac{1}{2} (2 - \delta_{mp}) \int (\underline{\phi}_p \cdot \nabla \underline{\phi}_m) + \underline{\phi}_m \cdot \nabla \underline{\phi}_p) \cdot \underline{\phi}_n d\underline{x}.$$
(9)

 Q_{mpn} is equal to Q_{mpn} when m = p and $\frac{1}{2}(Q_{mpn} + Q_{pmn})$ otherwise.

For the evolution equation of the amplitude a_n , $1 < n \le 10$, shown in Eq. (6), the interaction coefficient of a_n with the mean mode, Q_{n1n} , is essentially independent of n and its value is about 0.2. A physical interpretation of this is that each of the modes interacts directly and equally with the mean mode, in particular its strong shear layers.

Generally speaking, the magnitudes of the quadratic coefficients Q_{nmp} provide insight into the interactions between the different modes. Table III contains the values of the quadratic coefficients

TABLE III. Quadratic interaction coefficients with the main mode Q_{n1m} . Only coefficients larger than 0.1 are indicated; only coefficients larger than 0.4 (indicated in bold) are included in the model.

Q_{n1m}	m = 3	m = 4	m = 5	m = 6	m = 7	m = 8	m = 9	m = 10
n = 3	0.21	0.91				-0.22		0.37
n = 4	-0.94	0.23				0.0	0.29	0.17
n = 5			0.22	1.09	0.	0.34		
n = 6			-1.22	0.20	-0.12	-0.10		
n = 7				0.17	0.16	-0.42		
n = 8			0.55		0.4	0.17		
n = 9	0.32		0.59				0.22	-0.86
n = 10	-0.88	-0.18					0.98	0.24

 Q_{n1m} for $3 \le n, m \le 10$. The values of largest magnitude, which were kept for the model are indicated in bold.

A. Simplified model

We first consider a simplified version of the model by making the following assumptions:

(i) we assume that the first two modes a_1 and a_2 are constant, since they respectively correspond to the symmetric and antisymmetric part of the time-averaged field;

(ii) we neglect quadratic terms of small magnitude, since the dynamics are expected to be dominated by the largest interaction coefficients;

(iii) we model the energy transfer to the unresolved modes by assuming local equilibrium, which means that their effect is to compensate for the production term, i.e., the interaction with the mean shear (mode 1). This means that

$$T_n = (-L_n - Q_{n1n}a_1)a_n. (10)$$

This leads to the following form for the model, which we will refer to as M_0 :

$$\dot{a}_1 = 0 \tag{11}$$

$$\dot{a}_2 = 0 \tag{12}$$

$$\dot{a}_3 = 0.91a_4$$
 (13)

$$\dot{a}_4 = -0.94a_3 \tag{14}$$

$$\dot{a}_5 = -1.09a_6 + 0.49a_9 \tag{15}$$

$$\dot{a}_6 = 1.22a_5$$
 (16)

$$\dot{a}_7 = -0.42a_8 \tag{17}$$

$$\dot{a}_8 = -0.55a_5 + 0.44a_7 \tag{18}$$

$$\dot{a}_9 = -0.6a_4 - 0.86a_{10} \tag{19}$$

$$\dot{a}_{10} = -0.88a_3 + 0.98a_9. \tag{20}$$

We can see that except for a dependence of the evolution of mode 5 on mode 9, there are only interactions between modes with the same symmetry with respect to the vertical midplane reflection: (3,4,9,10) on the one hand and (5,6,7,8) on the other hand. The interactions between the normalized amplitudes $(a_{2i-1}, a_{2i}), 2 \le i \le 3$ are of the form $\dot{a}_3 = -q_3 a_4, \dot{a}_4 = q_4 a_3, \dot{a}_5 = -q_5 a_6, \dot{a}_6 = q_6 a_5$, with $q_i > 0$. It can be shown that the system can be rewritten as $a_{2i-1} = -\omega_i (\frac{\lambda_{2i}}{\lambda_{2i-1}})^{1/2} a_{2i}, \dot{a}_{2i} = \omega_i (\frac{\lambda_{2i-1}}{\lambda_{2i}})^{1/2} a_{2i-1}$, which correspond to propagative oscillatory solutions for the amplitudes $\sqrt{\lambda_n} a_n$, in agreement with the convective dynamics expected for the corresponding Kármán modes.

The frequencies $|\omega_i| = 2\pi f_i$ identified for the pairs (3,4), (5,6), (7,8), and (9,10) are about 0.92, 1.15, 0.43, 0.9, which correspond to frequencies (or Strouhal numbers) of 0.15, 0.18, 0.07, and 0.14. This is in good agreement with the main frequencies identified in the simulation. We emphasize that this prediction of the relevant time scales is based exclusively on the spatial structure of the modes, which are extracted from a set of samples arbitrarily separated in time. Since the computation is based on the derivatives of the spatial modes, some uncertainty exists in the determination of the time scales.

The model was integrated from a random initial condition, and the amplitudes a_i , $3 \le i \le 10$ are represented in spectral space in Fig. 18. As expected, the frequencies of the model coefficients agree



FIG. 18. Power spectral density of the 3D POD mode amplitudes $3-10 |\hat{a}_n^{3D,M_0}|^2$ in the simplified model; the red and black lines, respectively, correspond to the two frequencies 0.08 and 0.2.

well with the dominant frequencies identified in the previous section for the amplitudes of the modes in the simulation. The amplitudes of the modes are also close to their expected values, as shown in Table II. These results indicate that the main temporal dynamics of the flow can be recovered from the quadratic interactions between spatial POD modes, even if the snapshots are obtained with large separations, which is evidence of the predictive abilities of the POD-based model.

B. Switches

We now examine if and how switches can be reproduced by a more complex version of the model, which will be referred to as M. The model is obtained by relaxing the assumptions of the simplified model, presented in the previous paragraph, as follows.

(i) The second mode is allowed to vary.

(ii) Feedback is provided between the unresolved terms and the modes of the truncation.

As shown in Ref. [27], we assume that the rate of energy transferred to the small scales depends on the energy available in the largest scales. If more energy is available in the large scales, then more energy is extracted by the small scales and conversely. This leads to us to introduce a time-varying viscosity term, so that the effect of the unresolved terms is modelled as

$$T_n = A_n a_n + \epsilon_n, \tag{21}$$

where

(i) the time-averaged value $\langle A_n \rangle$ satisfies

$$\langle A_n \rangle = -L_n - Q_{nn1}; \tag{22}$$

(ii) A_n contains a linear part and a quadratic part.

The details of the modelling procedure can be found in Refs. [27,34]. We have

$$T_n = \left(\langle A_n \rangle + \alpha_n \sum_{p \ge 1}^N \lambda_p - \alpha_n \sum_{p \ge 1}^N |a_p|^2 \right) a_n + \epsilon_n,$$
(23)

where

$$\alpha_n = -\frac{\langle A_n \rangle}{2\sum_{p \ge 2}^N \lambda_p}.$$

The value of α_i was evaluated to be around 0.5. Since the model was found to be largely insensitive to the exact values of α_i , in what follows a constant value of $\alpha_i = \alpha = 0.5$ was used for all modes for the sake of simplicity. Examination of (21) and of the POD eigenvalues shows that about 50% of the turbulent viscosity is dependent on the amplitude of mode a_2 . In that sense the structure of the model displays similarities with both Refs. [8] and [28]'s model, which includes a cubic term in a_2 . However, it is derived from entirely different physical arguments.

(iii) ϵ_n represents Gaussian noise representing the high-frequency component due to the presence of unresolved scales. The r.m.s. value σ_n of each noisy perturbation ϵ_n used to integrate the model was determined using $\sigma_n \approx |A_n|$. We used $\sigma_2 = 0.09$ and $\sigma_i = 0.04$ for $3 \le i \le 10$.

The modified model reads:

$$\dot{a}_1 = 0 \tag{24}$$

$$\dot{a}_2 = \left(L_2 - \alpha \sum_{p=1}^{10} a_p^2 \lambda_p\right) a_2 + \epsilon_2 \tag{25}$$

$$\dot{a}_3 = \left(L_3 - \alpha \sum_{2}^{10} a_p^2 \lambda_p\right) a_2 + 0.71 a_4 + \epsilon_3$$
(26)

$$\dot{a}_4 = \left(L_4 - \alpha \sum_{2}^{10} a_p^2 \lambda_p\right) a_2 - 0.74 a_3 + \epsilon_4 \tag{27}$$

$$\dot{a}_{5} = \left(L_{5} - \alpha \sum_{2}^{10} a_{p}^{2} \lambda_{p}\right) a_{2} - 1.09a_{6} + \epsilon_{5}$$
(28)

$$\dot{a}_{6} = \left(L_{6} - \alpha \sum_{2}^{10} a_{p}^{2} \lambda_{p}\right) a_{2} - 1.2a_{5} + \epsilon_{6}$$
⁽²⁹⁾

$$\dot{a}_7 = \left(L_7 - \alpha \sum_{2}^{10} a_p^2 \lambda_p\right) a_2 - 0.42a_8 + \epsilon_7 \tag{30}$$

$$\dot{a}_9 = \left(L_9 - \alpha \sum_{2}^{10} a_p^2 \lambda_p\right) a_2 - 0.6a_4 - 0.86a_{10} + \epsilon_9 \tag{32}$$

$$\dot{a}_{10} = \left(L_{10} - \alpha \sum_{2}^{10} a_p^2 \lambda_p\right) a_2 - 0.88a_3 + 0.98a_9 + \epsilon_{10}.$$
(33)

The quadratic terms in the expression for a_2 are not included in the model as their magnitude was small (we checked that including these terms in the equations did not change the dynamics reported below).



FIG. 19. (Left) POD amplitude a₂; top: experiment (2D); bottom: model M (3D); (right) histogram of a₂.

The effect of the feedback term on the dynamics of the model is that if there is less energy in mode a_2 , the higher-order modes will extract less energy, which will allow mode a_2 to grow. Conversely, if mode a_2 becomes too large, the energy transfer to the higher-order modes will be increased, which will in turn affect the energy of mode a_2 .

Figures 19 and 20 show results of the model integration for a noise amplitude of about 0.15. Figure 19 (left) shows the 3D coefficient a_2^{3d} predicted with the model M along with the 2D coefficient a_2^{2d} in the experiment, which appears a relevant comparison since, as shown in the previous section, there is a reasonably good correlation between a_2 in three dimensions and two dimensions (0.6). The model displays time scales of O(1000) between switches, in agreement with experimental observations. Figure 19 (right) shows that the histogram of the amplitude is similar to that observed in the experiment in Fig. 6. This shows that the model is able to reproduce deviations in a way that is consistent with experiments. As Fig. 20 indicates, modes 3 to 10 are characterized by relatively fast oscillating time scales and slower amplitude variations. The frequencies of the amplitudes are shown in Fig. 20 (right) and compare relatively well with those measured in the simulation, given the crudeness of the truncation. Table II (last line) shows that the magnitude of



FIG. 20. (Left) amplitudes of 3D POD amplitudes of modes 3, 5, 7, and 9 in the model; (right) power spectral density of POD amplitudes in the model $|\hat{a}_n^{3d,\mathcal{M}}|^2$; the red and black lines, respectively, correspond to the two frequencies 0.08 and 0.2.

the normalized POD amplitudes is relatively well estimated by the model with values of about 0.2–2 for the last modes of the truncation (we emphasize that the model contains only 20% of the total fluctuating energy). The main dynamics of the largest scales are therefore captured by the model. Again the specificity of the model is that it is able to extract essential dynamical features of a complex flow by combining projection of the Navier-Stokes equations with general, data-independent assumptions.

VII. CONCLUSION

We have applied proper orthogonal decomposition to the 3D numerical simulation of the flow behind an Ahmed body at $Re = 10^4$. Reflection symmetry was applied to the computed data set in order to compensate for the relatively short time of the simulation, which precludes the observation of switches in the wake deviation. Both possible quasisteady states are thus included in in the augmented space of realizations, but it should be noted that the symmetrization procedure cannot provide a description of the switching process itself.

As a consequence of the enforced statistical symmetry, the flow can be decomposed into symmetric and antisymmetric structures. The mean flow consists of a symmetric recirculation bubble and an antisymmetric deviation, the effect of which is to gather flow streamlines around one of the base diagonals. 2D POD analysis was performed in the near wake in the simulation and compared with experimental results obtained for the same geometry. Despite the discrepancy in Reynolds number between the simulation and the experiment, an excellent agreement was observed for both the spatial structure and temporal statistics of the POD modes.

2D results were then confronted with a 3D approach. The energetic importance of the quasisteady wake deviation was established. The evolution of this global 3D deviation mode was relatively well captured by 2D measurements in the near wake. The next most energetic patterns are associated with vortex shedding and wake pumping. Characteristic frequencies were identified for each type of structure. Structures associated with wake pumping were characterized by a low frequency of about 0.08. Both symmetric and antisymmetric structures associated with vortex shedding in, respectively, the vertical and spanwise direction were characterized by dominant frequencies of 0.19 and 0.23 in the far wake. A strong negative correlation mode. In addition, increases in the base drag were found to correspond to an increase of the deviation mode magnitude, along with a shrinkage of the recirculation zone associated with wake pumping.

Finally, POD-based low-dimensional models were derived for the largest scales of the flow. The energy content of the modes was correctly captured by the model. A simplified model was able to single out the main frequencies of the POD amplitudes observed in the simulation from the spatial modes, regardless of the time separation between the snapshots used to compute POD. This predictive ability of the model is remarkable in view of the slow convergence of the decomposition, which reflects the complexity of the flow. A more elaborate version of the model was also considered. The approach is consistent with Rigas *et al.*'s model [8] and in particular the structure of the model would be the same if the POD truncation was limited to two modes. By accounting for the effect of the unresolved modes with a feedback term, the POD-based model was able to reproduce the characteristics of the switches in the wake deviation. The success of the model supports the idea that wake switching is triggered by higher-order modes.

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