

# GENERATION OF TURBULENT INFLOW CONDITIONS FOR LARGE EDDY SIMULATION FROM STEREOSCOPIC PIV MEASUREMENTS

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## ABSTRACT

A novel method to generate inflow data for unsteady numerical simulations is proposed. This method consists in coupling an experimental database obtained by stereoscopic PIV measurements which are under-resolved in time to a numerical code. Based on the Proper Orthogonal Decomposition of this database, the proposed approach enables the adaptation of the experimental mesh to the grid of the simulation and the modelling of the temporal dynamics of the flow field in the inlet section. To test this method, a LES of a turbulent plane mixing layer configuration is performed.

## INTRODUCTION

The prescription of well-suited boundary conditions is a crucial issue when performing unsteady numerical computations of turbulent flows. In the case of the simulation of a spatially developing flow, the specification of the flow field at the inlet of the computational domain conditions the downstream development of the flow. The simplest and historical method to address this problem is to impose well chosen infinitesimal perturbations, where intrinsic most amplified modes are present, on a laminar mean flow profile at the inflow section and to let the flow evolve to a fully turbulent state. The main drawback of this approach is the need of a large computational domain required for the transition to the turbulent state to occur. Hence, the computational cost of the simulation can be dramatically increased. Consequently, several studies (see Keating *et al.* (2004) for a detailed review) have been devoted to design more efficient methods to prescribe realistic turbulent inflow conditions which present correct mean values, proper one-point and two-point correlations and given spectrum as well as correct phase information. This can be achieved by running a separate precursor calculation (Li *et al.*, 2000) from which a velocity field in a plane normal to the streamwise direction is extracted. This sequence of planes is used afterward as inlet conditions for the main computation. To avoid the additional cost of the precursor simulation, Lund *et al.* (1998) proposed a recycling method which consists in extracting velocity data from a downstream section, rescaling the velocity profile to reintroduce them at the inlet section of the domain. Nevertheless, this method involves the existence of a region where similarity laws enable the velocity profiles to be rescaled. Moreover, the periodicity effect of this recycling procedure can lead to the excitation of some particular modes. An alternative to these methods is to use synthetic turbulence generated by using random numbers satisfying constraints e.g. first and second order moments and spectra (Lee *et al.*, 1992). However, simulations using these approaches re-

quire an adaptation zone downstream of the inlet section, due to the *unphysical* random character of the synthesized velocity fields. Moreover, correct target statistics, especially two-point information, can be difficultly obtained. Recently, Druault *et al.* (2004) developed a method which enables the unsteady coupling between an experimental database obtained through hot-wire measurements and a spatially-developing Large Eddy Simulation (LES). This technique was based on the joint use of the Proper Orthogonal Decomposition (POD) and the Linear Stochastic Estimation (LSE), permitting the reconstruction of the velocity field on the full mesh of the inlet section from only a few measurement locations. Although this method takes advantage of the good temporal resolution of hot-wire measurements, it suffers from a low spatial resolution due to the limited number of probes that can be implemented simultaneously.

The goal of the proposed method, based on the use of the POD, is to couple a database obtained by Stereoscopic Particle Image Velocimetry (SPIV) to a computational code. This measurement technique provides a spatial resolution consistent with that of the numerical simulation. To compensate the low temporal resolution offered by the SPIV measurements, the temporal behaviour of the flow in the inlet section is modelled by synthetic random time series.

In the present paper, this method is used to perform a LES of a turbulent plane mixing layer, the inlet section of the computational domain being located at the beginning of the self-similarity region.

## PRINCIPLES

The proposed inlet condition generation method aims at interfacing experimental data which are under-resolved in time to an unsteady computational code. Once the database is acquired, the method consists of two steps. Firstly, the experimental mesh must be adapted to the numerical grid, which can imply spatial interpolation and extrapolation in the trasversal and spanwise directions, and the generated velocity fields must be consistent with the other boundary conditions retained for the simulation (namely periodic condition in the spanwise direction, free-slip conditions at the horizontal boundaries). Secondly, the temporal dynamics of the flow in the inlet section must be correctly reproduced. The POD, as described later, is used to uncouple these two steps by decomposing the velocity field into a set of spatial and temporal modes (respectively  $\phi_i^n(\mathbf{x})$  and  $a^n(t)$ ):

$$u_i(\mathbf{x}, t) = \sum_{n=1}^{N_s} a^{(n)}(t) \phi_i^{(n)}(\mathbf{x}) \quad (1)$$

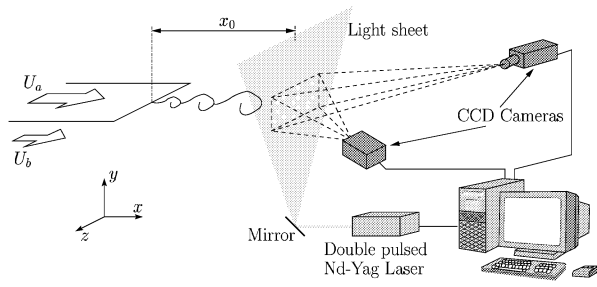


Figure 1: Experimental set up.

Thus, the first step is performed by independently processing each spatial eigenvector  $\phi_i^{(n)}(\mathbf{x})$ . Furthermore, the anisotropy and the inhomogeneity of the velocity field is taken into account implicitly by these spatial modes. The global dynamics of the whole flow field is represented by the temporal projection coefficients  $a^{(n)}(t)$ . In the present case, due to the low sampling frequency of the SPIV system compared to the characteristic frequencies of the flow, these temporal coefficients are synthetically modelled.

## FLOW CONFIGURATION AND EXPERIMENTAL SET UP

An experimental database is obtained from SPIV measurements. A sketch of the experimental set up is presented in Figure 1. A plane subsonic turbulent air/air mixing layer with a high-speed velocity  $U_a$  of 35.2 m/s and a low-speed velocity  $U_b$  equal to 23.8 m/s (velocity ratio  $r = 0.67$ ) is used. The coordinate system is  $x$  in the streamwise direction,  $y$  in the vertical direction and  $z$  in the spanwise direction. The origin is located at the splitter plate trailing edge. The measurements were performed in a section of the flow normal to the streamwise direction, located at  $x_0 = 300$  mm from the trailing edge of the splitter plate, corresponding to the beginning of the self-similarity region of the mean flow quantities. The Reynolds number, based on the mean velocity  $U_m = (U_a + U_b)/2$  and the vorticity thickness in the measurement section  $\delta_{\omega_0} = 18.7$  mm, is  $Re = \delta_{\omega_0} U_m / \nu \simeq 36,000$ . The typical frequency  $f_c$  associated to the large scale structures in the mixing layer, corresponding to a Strouhal number  $St = f_c \delta_{\omega_0} / U_m = 0.3$ , is about 480 Hz.

The SPIV set up comprises two double frame CCD cameras  $1350 \times 1049$  pixels<sup>2</sup> in angular stereoscopic configuration to record images of the flow and a double-pulsed Nd-Yag laser to create the light sheet. The flow was seeded with olive oil tracer particles. Acquisition, synchronisation and SPIV post-processing were performed with the *Davis 6.2 Lavision* software. More details about the flow investigated and the SPIV set up can be found in Perret (2004) and Perret *et al.* (2004). The size of the field of SPIV measurement,  $L_y \times L_z = 4\delta_{\omega_0} \times 6.1\delta_{\omega_0}$ , is chosen large enough to capture the main scales of the flow in the  $y$ - and  $z$ -directions while preserving a good spatial resolution of about  $0.088 \delta_{\omega_0}$ . A set of  $N_s = 2,000$  velocity fields was acquired at a frequency of 1 Hz.

## METHOD FOR EXPERIMENT/COMPUTATION INTERFACING

### Proper Orthogonal Decomposition

In this section, the Proper Orthogonal Decomposition is briefly presented. More details can be found in Berkooz *et al.* (1993) and Sirovich (1987). Lumley (1967) first proposed the POD technique to identify the coherent structures in turbulent flows. It consists in extracting from the flow the structure  $\phi(\mathbf{X})$  with the largest mean-square projection onto the velocity field  $\mathbf{u}(\mathbf{X}, t)$ . This maximization problem leads to the solving of the integral problem of eigenvalues:

$$\int_{\mathcal{D}} R_{ij}(\mathbf{X}, \mathbf{X}') \phi_j^{(n)}(\mathbf{X}') d\mathbf{X}' = \lambda^{(n)} \phi_i^{(n)}(\mathbf{X}) \quad (2)$$

where  $\lambda^{(n)}$  corresponds to the  $n$ -th eigenvalue and represents the amount of energy contained in the mode  $\phi^{(n)}(\mathbf{X})$ .  $R_{ij}$  is the two-point space correlation tensor over the domain  $\mathcal{D}$ :

$$R_{ij}(\mathbf{X}, \mathbf{X}') = \langle u_i(\mathbf{X}, t) u_j(\mathbf{X}', t) \rangle \quad (3)$$

where  $\langle \cdot \rangle$  is the ensemble average operator. The fluctuating field can be projected onto the POD basis composed of the eigenfunctions  $\phi_i^{(n)}(\mathbf{X})$ :

$$u_i(\mathbf{X}, t) = \sum_{n=1}^{\infty} a^{(n)}(t) \phi_i^{(n)}(\mathbf{X}) \quad (4)$$

The projection coefficients are computed as:

$$a^{(n)}(t) = \int_{\mathcal{D}} u_i(\mathbf{X}, t) \phi_i^{(n)}(\mathbf{X}) d\mathbf{X} \quad (5)$$

and are uncorrelated in time:

$$\langle a^{(n)}(t) a^{(m)}(t) \rangle = \lambda^{(n)} \delta_m^n \quad (6)$$

In the present work, we used the version proposed by Sirovich (1987), called the Snapshot POD, which is more adapted to data which are well-resolved in space with a limited number of time samples. Moreover, the spatial domain considered here is the  $y - z$  plane. Consequently, the spatial eigenvectors are function of two variables:  $\phi_i^{(n)}(y, z)$ ,  $i = 1, 2, 3$ .

### Horizontal extrapolation

In the homogeneous spanwise direction, a periodic condition is retained in the numerical simulation. Hence, inlet conditions must satisfy this constraint at each time step. Periodicity has then to be imposed to experimental data that are *a priori* non-periodic. Two steps are involved: 1) determination of the size of the domain on which the periodic condition must be imposed, 2) extrapolation and periodization of the data on this domain:

- 1) POD eigenvectors are known to degenerate into harmonic modes when the POD is performed in a homogeneous direction (Berkooz *et al.*, 1993). As proposed by Coiffet *et al.* (2002), the POD can then be used in the spanwise direction to detect a global wave length in the data. The advantage of this approach is that the wave length of the POD eigenvectors is not constrained by the size of the domain on which the POD is applied, as a Fourier transform would be. Thus, at each vertical location  $y$ , a one-dimensional POD is performed in the direction  $z$  in order to detect a common wave length that is not directly related to the size of the measurement domain.

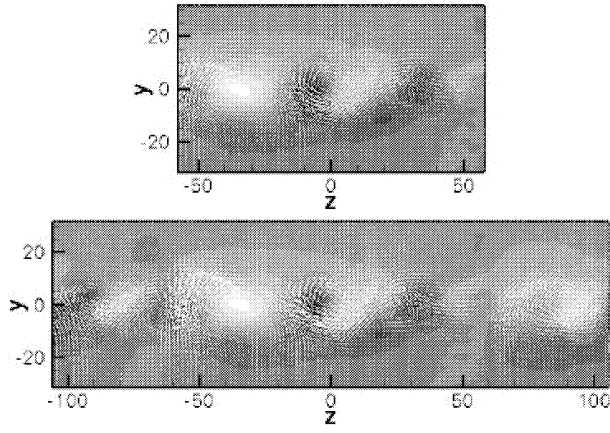


Figure 2: Example of a Snapshot POD eigenvector  $\phi_i^{(n)}(y, z)$  (top) and its spanwise extrapolation (bottom) ( $n = 8$ ).

- 2) Once the spanwise extent of the periodic domain is known, each two-dimensional Snapshot POD eigenvector  $\phi_i^{(n)}(y, z)$  is individually extrapolated and periodized. In that manner, the final reconstructed velocity field  $u_i(\mathbf{x}, t) = \sum_{n=1}^{N_s} a^{(n)}(t) \phi_i^{(n)}(\mathbf{x})$  satisfies the periodicity condition. To perform this extrapolation, the technique of selective deconvolution, proposed by Franke (1987) to extrapolate signals, the spectrum of which shows dominant spectral frequencies, is retained. This algorithm is iterative and is based on direct and inverse fast Fourier transforms (FFT). The final signal is periodic by construction on the final domain. Moreover, this approach enables to remove the windowing effect from the signal caused by the limited extent of the measurement domain.

This extrapolation procedure is applied to the velocity fields obtained from the SPIV measurements. The first step, which consists in detecting a wave length to periodize the domain, leads to a domain which spanwise extent is 1.83 time larger than the original measurement window.

Each Snapshot POD eigenvector  $\phi_i^{(n)}(y, z)$  is then extrapolated in the  $z$ -direction via a bidimensional approach of the algorithm of Franke (1987) using bidimensional FFT.

As can be seen in Figure 2, the extrapolation procedure does not introduce any spatial discontinuity in the eigenvectors and correctly spreads the motives corresponding to the turbulent structures extracted by the POD. Moreover, it can be noticed that the extrapolated eigenvectors are periodic in the final domain.

### Vertical extrapolation

Conditions chosen on the upper and lower horizontal boundaries of the computational domain are free-slip conditions. Moreover, to prevent the flow from being confined in this direction, an extension of the inlet domain in the vertical direction is required. Hence, the experimental velocity fields must also be extrapolated in this direction. A one-dimensional POD approach is used to extrapolate each Snapshot POD eigenvector  $\phi_i^{(n)}(y, z)$  in the  $y$ -direction. Here, each spanwise location  $z$  is processed independently by the following steps:

- 1) compute the two-point correlation tensor  $R(y, y') = \langle u_i(y)u_i(y') \rangle$  from the original velocity fields where  $\langle \cdot \rangle$

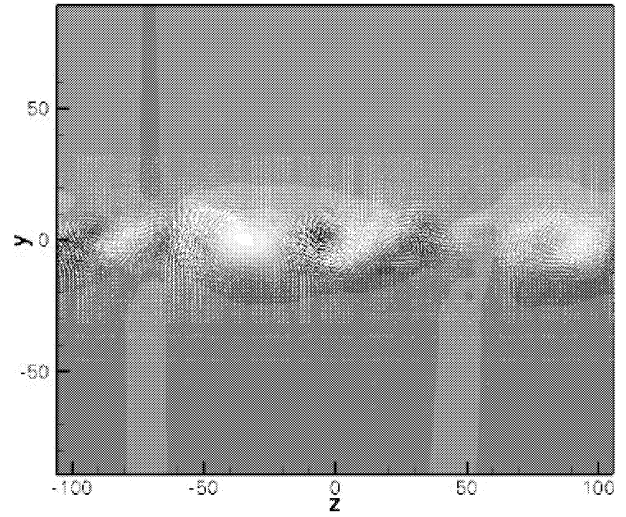


Figure 3: Vertical and spanwise extrapolation of the Snapshot POD eigenvector  $\tilde{\phi}_i^{(n)}(y, z)$  presented on Figure 2 ( $n = 8$ ).

is the average operator in both the temporal and spanwise directions.

- 2) perform a POD decomposition of this tensor to obtain a POD basis of  $N_y$  one-dimensional eigenvectors  $\psi_i^{(m)}(y)$ ,  $N_y$  being the number of points in the vertical direction. The snapshot POD eigenvectors  $\phi_i^{(n)}(y, z)$  can be projected onto the basis of the vectors  $\psi_i^{(m)}(y)$ :

$$\phi_i^{(n)}(y, z) = \sum_{m=1}^{N_y} c^{(m,n)}(z) \psi_i^{(m)}(y) \quad (7)$$

- 3) As pointed out by Druault and Delville (2000), POD eigenvectors  $\psi_i^{(m)}(y)$  obtained in the direction of inhomogeneity  $y$  in free turbulent shear flows must decrease to zero as the energy tends to 0 when  $y$  tends to  $\pm\infty$ . Moreover, in the case of the mixing layer, the  $\psi_i^{(m)}(y)$  decrease exponentially. Hence, the basis  $\{\psi_i^{(m)}(y)\}$  can be extrapolated in the  $y$ -direction on the final domain into a new set of modes  $\{\tilde{\psi}_i^{(m)}(y)\}$
- 4) Extrapolated bidimensional Snapshot POD eigenvectors  $\tilde{\phi}_i^{(n)}(y, z)$  can then be reconstructed over the larger domain using the extrapolated basis  $\{\tilde{\psi}_i^{(m)}(y)\}$ :

$$\tilde{\phi}_i^{(n)}(y, z) = \sum_{m=1}^{N_y} c^{(m,n)}(z) \tilde{\psi}_i^{(m)}(y) \quad (8)$$

Example of an extrapolated Snapshot POD mode in the vertical direction is shown in Figure 3. Eventually, the eigenvectors  $\tilde{\phi}_i^{(n)}(y, z)$  are used to reconstruct the velocity field in the final domain:  $\tilde{u}_i(y, z, t) = \sum_{n=1}^{N_s} a^{(n)}(t) \tilde{\phi}_i^{(n)}(y, z)$ .

### Temporal modelling

The temporal behaviour of the POD modes is modelled through the projection coefficients  $a^{(n)}(t)$ . Given the fact that no spectral information is available about these coefficients except their total energy content (e.g. the eigenvalues  $\lambda^{(n)}$ ), the

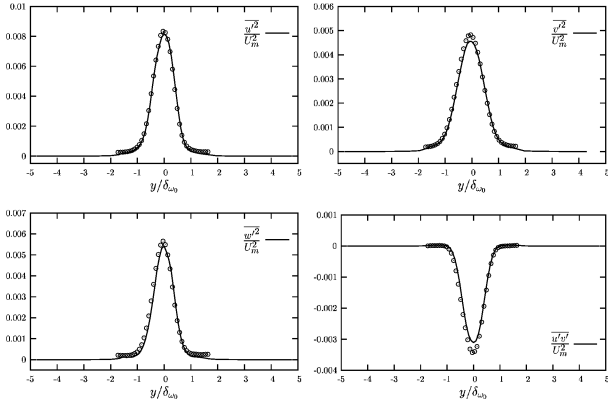


Figure 4: Reynolds stress profiles  $\overline{u_i' u_j'} / U_m^2$  of the synthesized (—) and experimental (o) velocity fields.

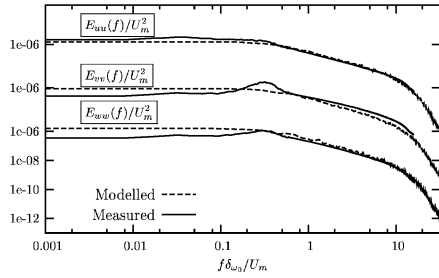


Figure 5: Spectra of the generated velocity fields obtained at  $y/\delta_{\omega_0} = 0.5$ . —, hot-wire measurements; - - -, modelled data.

$a^{(n)}(t)$  are modelled by time series of random numbers. Since these coefficients are mutually uncorrelated in time (Equation 6), they can be generated independently.  $N_s$  gaussian time series  $g^{(n)}(t)$  of random numbers of zero mean and rms equals to 1 are first generated. Then, to enforce a realistic spectral repartition of energy, a spectral transfer function, deduced from a modified Von Karman spectrum  $E_{Pao}(f)$  proposed by Pao (Hinze, 1975), is applied:

$$\hat{a}^{(n)}(f) = \hat{g}^{(n)}(f) \left( \frac{E_{Pao}(f)}{E_{gg}(f)} \right)^{1/2} \quad (9)$$

where  $\hat{g}^{(n)}(f)$  is the Fourier transform of  $g^{(n)}(t)$  and  $E_{gg}(f)$  the spectrum of  $g^{(n)}(t)$ . Characteristic frequencies delimiting the inertial range of the velocity spectra are evaluated from prior hot-wire measurements (Perret, 2004). Random time series are finally normalized to ensure that  $\langle a^{(n)} a^{(n)} \rangle = \lambda^{(n)}$ . Hence, the velocity field which results from the linear combination of the different POD modes presents the correct one- and two-point spatial statistics with a spectral repartition of the energy given by the shape of the chosen theoretical spectrum.

## GENERATED INFLOW CONDITIONS

A database of synthetic velocity fields has been generated with a temporal sampling frequency corresponding to the time step used in the LES computation. One- and two-point statistics of these velocity fields are presented in this section.

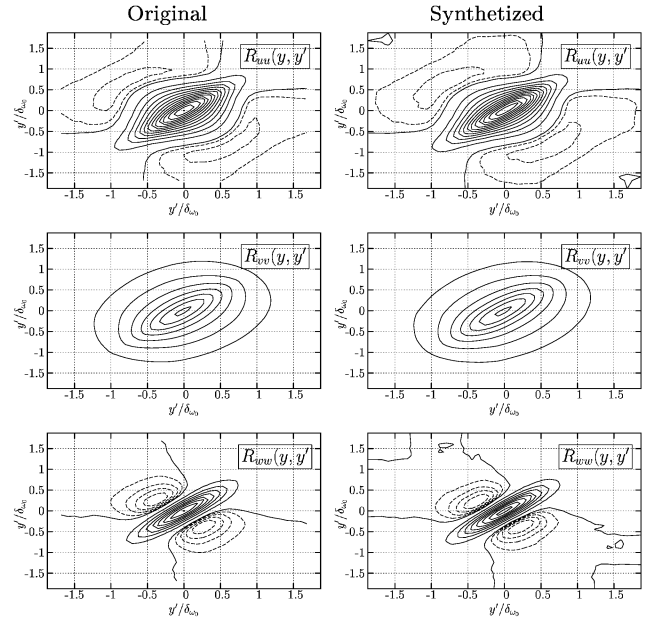


Figure 6: Two-point correlation  $R_{ii}(y, y')$  of the original velocity fields (left) and synthetic velocity fields (right). —: positive increment between contours of +0.5, - - -: negative increment between contours of -0.1.

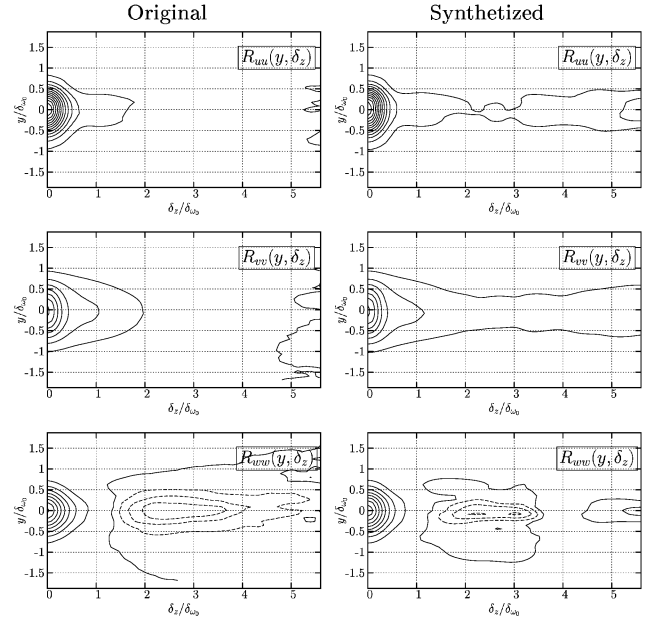


Figure 7: Two-point correlation  $R_{ii}(y, \delta_z)$  of the original velocity fields (left) and synthetic velocity fields (right). —: positive increment between contours of +0.5, - - -: negative increment between contours of -0.1.

## One-point statistics

Vertical distribution of the energy of each velocity component, integrated over  $z$ , is presented in Figure 4. When compared to the statistics of the original experimental data, synthesized velocity fields exhibit correct energy levels as well as spatial repartition. In particular, the anisotropy of the flow and the shear stress  $\overline{u'v'}$  are well reproduced.

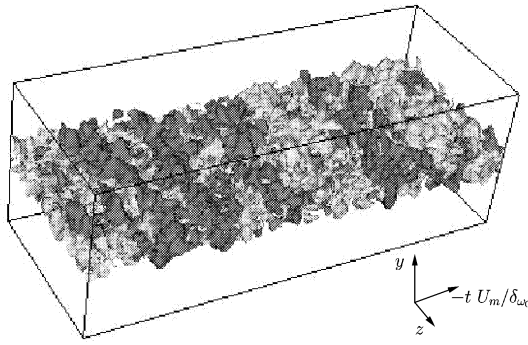


Figure 8: Isosurface of the vertical velocity component,  $v \pm 0.02 U_m$  of the temporal evolution of the synthetic inflow data ( $x = -U_m t$ ).

Nonetheless, given the simple spectral model used for the temporal coefficients  $a^{(n)}(t)$ , typical frequencies of the mixing layer found by hot-wire measurements are not present in the generated data (Figure 5).

### Two-point statistics

Figure 6 shows good agreement between measured and synthesized two-point correlations  $R_{ii}(y, y')$ . It must be stressed that, correlation levels, correlation lengths as well as the presence of regions of negative excursions are correctly reproduced by the proposed modelling approach.

In the spanwise direction (Figure 7), the two-point correlation tensor  $R_{ii}(y, \delta_z)$  is also correctly reproduced. Nevertheless, it appears that the extrapolation procedure affects the long range correlation levels which remain higher than the measured one. These differences can result from the introduction of parasite low wave numbers during the extrapolation-periodisation step.

### Instantaneous flow fields

Temporal evolution of the generated velocity field in the inlet section (Figure 8) exhibits well organized regions both in the vertical and the spanwise directions. The random character of the generated time series leads to a lack of structuration in the temporal direction.

### LES WITH GENERATED INFLOW CONDITIONS

A LES of the mixing layer is performed with *Code\_Saturne* a finite volume solver on unstructured grids developed at EDF, for vectorial and parallel computing (Archambeau et al., 2004). The filtered, incompressible Navier-Stokes equations are solved with the Piomelli and Liu (1995) subgrid scale model. Space discretization is based on a collocation of all the variables at the center of gravity of the cells, using central differencing. Time discretization is based on a second order Crank-Nicolson/Adams-Bashforth method. The velocity/pressure system is solved by a SIMPLEC algorithm, with the Rhie and Chow interpolation. The Poisson equation is solved using a conjugate gradient method, with diagonal preconditioning. Computations are performed using 8 processors of a PC cluster. The inlet of the computational domain corresponds to the measurement section and the above-presented

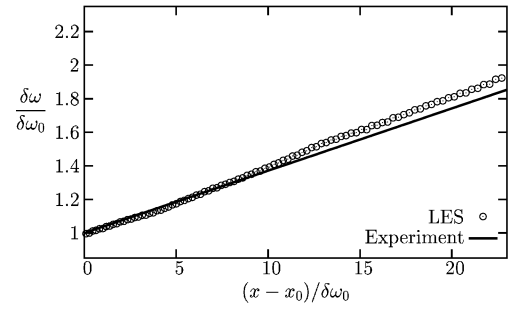


Figure 9: Streamwise evolution of the vorticity thickness  $\delta_\omega$ : —, experimental results;  $\circ$ , LES results using the synthetic inflow conditions.

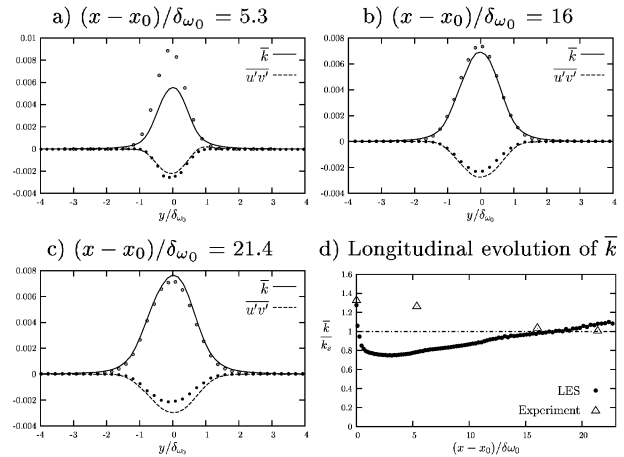


Figure 10: a), b) and c): evolution of  $\bar{k}$  and  $\overline{u'v'}$  at three different downstream locations (lines: LES; symbols: experiment). d): streamwise evolution of  $\bar{k}$  normalized by the value obtained experimentally in the self-similarity region:  $\circ$ , LES results;  $\triangle$ : experiment.

generated data are used as inlet conditions. The Reynolds number in the inlet section of the computation corresponds to the experimental one and is 36,000. The spatial extent of the domain is  $L_x \times L_y \times L_z = 30 \delta_{\omega_0} \times 10 \delta_{\omega_0} \times 11 \delta_{\omega_0}$ , meshed non regularly with  $N_x \times N_y \times N_z = 112 \times 53 \times 65$  points. The streamwise extent of the computational domain is chosen to enable the simulation of a significative evolution of the flow, corresponding to the doubling of the shear layer thickness ( $(x - x_0) / \delta_{\omega_0} \leq 23$ ). The last part of the domain is used to avoid some possible influence of the outflow condition.

Concerning the downstream evolution of the simulated mean flow, it is seen in Figure 9 that the growth rate is correctly reproduced.

Study of the spatial repartition of the kinetic turbulent energy  $\bar{k} = \frac{1}{2}(\overline{u'^2} + \overline{v'^2} + \overline{w'^2})$  and the shear stress  $\overline{u'v'}$  across the shear layer (Figure 10) reveals good agreement with the experimental data, except in the region just downstream the inlet section where a decrease of  $\bar{k}$  can be seen. It traduces the fact that the code must regenerate the correct phase information between modes to produce realistic turbulent structures.

Results obtained with the present method are compared (Figure 10 d)) to those obtained via hot-wire measurements

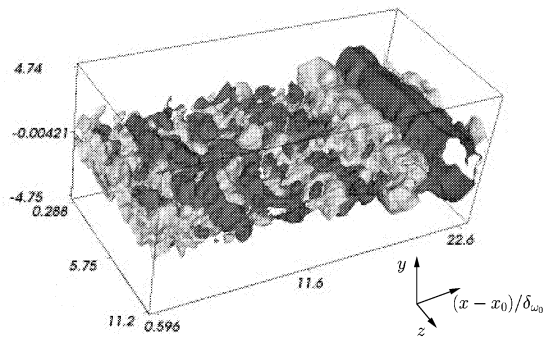


Figure 11: Isosurface of the vertical velocity component,  $v \pm 0.02 U_m$  of the LES results.

at different downstream locations (Perret, 2004). It should be pointed out here that the inlet section is located in a region where the turbulent quantities are out of similarity regime, contrary to the mean flow. Hence, due to the presence of a region of adaptation just downstream the inlet section, the code fails to correctly reproduce the slow decay of energy to the self-similarity level. However, the energy level obtained in the self-similarity region is reached approximately at the same spatial location both in the experiment and the computation.

Instantaneous simulated flow field (Figure 11) shows the rapid regeneration of realistic coherent structures with the downstream location.

## CONCLUSIONS

An original approach to generate synthetic inflow conditions for numerical simulations has been presented and its viability has been demonstrated by performing a LES of a fully turbulent plane mixing layer. The proposed method is designed to interface stereoscopic PIV measurements, which are under-resolved in time, to a computational code via the use of the POD. This technique enables to perform separately the spatial adaptation of the experimental data to the numerical mesh and the temporal modelling of the dynamics of the flow in the inlet section.

The proposed method turns out to be able to generate synthetic data which present correct one- and two-point spatial statistics in the whole section as well as realistic turbulent spectra. Given the random character of the temporal series generated, LES using these data demonstrates the good performance of the proposed approach. Indeed, a turbulent mixing layer can be simulated by starting from a section where the flow is fully developed and highly turbulent.

Future work will concern the improvement of the temporal modelling of each POD modes to be able to take into account characteristic frequencies of the flow under interest. Strategies based on low-order dynamical system will be developed to model the dynamics of the first POD modes representing the large scale organization of the flow in the measurement section.

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