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HAL Id: hal-02115966
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Submitted on 30 Apr 2019

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A Kriging-based elliptic extended anisotropic model for the turbulent boundary layer wall pressure spectrum

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The present study addresses the computation of the wall pressure spectrum for a turbulent boundary layer flow without pressure gradient, at high Reynolds numbers, using a new model, the Kriging-based elliptic extended anisotropic model (KEEAM). A space–time solution to the Poisson equation for the wall pressure fluctuations is used. Both the turbulence–turbulence and turbulence–mean shear interactions are taken into account. It involves the mean velocity field and space–time velocity correlations which are modelled using Reynolds stresses and velocity correlation coefficients. We propose a new model, referred to as the extended anisotropic model, to evaluate the latter in all regions of the boundary layer. This model is an extension of the simplified anisotropic model of Gavin (PhD thesis, 2002, The Pennsylvania State University, University Park, PA) which was developed for the outer part of the boundary layer. It relies on a new expression for the spatial velocity correlation function and new parameters calibrated using the direct numerical simulation results of Sillero \textit{et al.} (\textit{Phys. Fluids}, vol. 26, 2014, 105109). Spatial correlation coefficients are related to space–time coefficients with the elliptic model of He & Zhang (\textit{Phys. Rev. E}, vol. 73, 2006, 055303). The turbulent quantities necessary for the pressure computation are obtained by Reynolds-averaged Navier–Stokes solutions with a Reynolds stress turbulence model. Then, the pressure correlations are evaluated with a self-adaptive sampling strategy based on Kriging in order to reduce the computation time. The frequency and wavenumber–frequency wall pressure spectra obtained with the KEEAM agree well with empirical models developed for turbulent boundary layer flows without pressure gradient.

\textbf{Key words:} turbulence modelling, turbulent boundary layers, turbulent flows

1. Introduction

Wall pressure fluctuations induced by turbulent boundary layer (TBL) flows raise several engineering issues such as trailing edge noise, vibrations of structures which in turn generate acoustic radiations and disturbances for towed underwater sonar and

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sonar domes. In the case of military naval architecture, the prediction and reduction of noise generated by elongated vibrating structures is crucial. In order to analyse their stochastic response to wall pressure excitations, the space–time correlation function or the wavenumber–frequency spectrum is required.

Different regions can be defined for the pressure wavenumber–frequency spectrum: the supersonic region, $k < k_a$, the subconvective region, $k_a < k < k_c$, the convective region (or convective peak), $k \approx k_c$, and the viscous region, $k \gg k_c$ (where $k_a = \omega/c$ is the acoustic wavenumber, $k_c = \omega/U_c$ is the convective wavenumber, $\omega$ is the radian frequency, $U_c$ is the convection velocity and $c$ is the sound velocity). The most energetic one is the convective region with pressure levels approximately 40 dB above the levels in the subconvective range (for an attached TBL with zero pressure gradient) (Bonness, Capone & Hambric 2010). In the case of aircraft applications with high Mach number flows, significant vibrations are caused by pressure-field components with wavenumbers of the convective range which match the wavenumbers of the structure bending waves (Graham 1997). However, for underwater applications, the convective peak may not be the only important part of the pressure spectrum. For low Mach number flows over elongated structures, strong structural responses result from excitation by pressure-field components with low wavenumbers which match those of the subconvective region (Bull 1996; Bonness et al. 2010). Moreover, as the sound velocity is higher in water and the vehicle speeds are lower, the acoustic and convective wavenumbers are shifted downward and upward respectively. Thus, the subconvective range is larger for underwater applications.

Accurate predictions of the pressure spectrum may be obtained using direct numerical simulation (DNS) or large-eddy simulation (LES) (Choi & Moin 1990; Chang, Piomelli & Blake 1999; Hu, Morfey & Sandham 2006; Wang et al. 2009). However, the computation time for these simulations is still too large for complex industrial problems where numerous simulations have to be done. Thus, modelling techniques still have to be developed and used to predict the pressure loading.

Numerous semi-empirical models for the pressure spectrum can be found in the literature. They were proposed during the last decades for the frequency spectrum $\Phi_{pp}(\omega)$ and the wavenumber–frequency spectrum $\Phi(k_1, k_3, \omega) = \Phi(k, \omega)$ (where $k_1$ and $k_3$ are the streamwise and spanwise wavenumbers respectively, $k = (k_1, 0, k_3)$ is the wave vector). Detailed reviews of frequently used models for the pressure spectrum can be found in the papers of Graham (1997) and Hwang, Bonness & Hambric (2009). A few of them are presented in table 1. These models are based on data fitting and a certain degree of analytical and theoretical modelling. They only need few boundary layer characteristics as input so they are well adapted for design phases. However, they were mostly developed for zero pressure gradient turbulent boundary layers with ideal conditions, which may be far from the conditions of interest. Rozenberg, Robert & Moreau (2012) showed that empirical models for the frequency spectrum underestimate the spectral levels, particularly at low frequencies, in case of adverse pressure gradient (APG). Thus, they proposed a frequency spectrum model based on Goody (2004) which takes into account pressure gradient effects but it is restricted to APG. Moreover, for the wavenumber–frequency spectrum, the available semi-empirical models agree around the convective peak but show large discrepancies in the low wavenumber range which is also an area of interest.

Another method is to express a solution to the Poisson equation for the pressure which relates wall pressure fluctuations to velocity sources in the turbulent flow. These source terms consist of the turbulence–mean shear (TMS) terms and the turbulence–turbulence (TT) terms, which are also referred to as rapid and slow terms
Table 1. Wall pressure spectra semi-empirical models. Flow: ZPG-TBL, zero pressure gradient turbulent boundary layer; C, channel; A, airfoil; APG, adverse pressure gradient. $Re_\theta = U_c \theta / \nu$ with: $U_c$, free-stream velocity; $\nu$, kinematic viscosity; $\theta$, momentum thickness. Input: $U_c$, convection velocity; $u_\tau$, friction velocity; $\delta$, boundary layer thickness; $\delta^*$, displacement thickness; $\tau_{\text{max}}$, maximum shearing stress along the normal; $dP/dx_1$, mean pressure gradient. Output: $\Phi_{pp}(\omega)$, wall pressure frequency spectrum; $\Phi(\mathbf{k}, \omega)$, wall pressure wavenumber–frequency spectrum.

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respectively. Two kinds of approaches can be considered: a spectral solution which gives an expression for the wavenumber–frequency pressure spectrum; a space–time solution which gives an expression for the pressure fluctuation correlations. A few examples of both strategies are presented in table 2. The former approach was first applied by Kraichnan (1956) who expressed the wavenumber–frequency spectrum for a TBL flow over a flat plate as a double integral over the wall-normal coordinate. His expression was obtained by taking into account only the linear TMS term and supposing the homogeneity of the TBL in planes parallel to the wall. This form involves the gradient of the mean streamwise velocity in the wall-normal direction and the Fourier transform of the vertical velocity correlations. The latter can be modelled using the Reynolds stress $\overline{u'v'}$ and the vertical velocity correlation coefficient $C_{22}$ or a spectral correlation function $\Phi_{22}$. Based on the work of Kraichnan (1956), several authors have proposed similar relationships for the pressure spectrum (see table 2). Panton & Linebarger (1974) used empirical models for the mean velocity and the velocity correlations as input for their expression. Blake (1986) used a derived expression for the frequency spectrum and deduced some characteristics for the spectrum due to the TMS term. Lysak (2005) applied the method to turbulent pipe flows. In more recent works, Reynolds-averaged Navier–Stokes (RANS) solutions are used as input data for relationships similar to Kraichnan (1956) which are applied to TBL flows without a pressure gradient (e.g. Aupoix 2015) or to more complex flows such as flow over a backward-facing step (Lee, Blake & Farabee 2005) or over an airfoil (Remmler et al. 2010; Bertagnolio, Fischer & Jun Zhu 2014).

As the Poisson equation for the wall pressure fluctuations with the nonlinear TT source terms cannot be solved directly in the spectral domain, these terms are disregarded in the Kraichnan (1956) relation and similar works. Neglecting the TT source terms is often considered as a satisfactory assumption, particularly at mid and high frequencies (Kraichnan 1956; Panton & Linebarger 1974; Remmler et al. 2010). However, when the area of interest is located at low frequencies and wavenumbers, the TT source terms should not be neglected. Besides, Chang et al. (1999) showed,
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<td>$Re_\tau = 1.6$, $3$, $4$, $5 \times 10^6$</td>
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<td>$C_{ij}$, $R_{pp}(r_s, \Delta t)$, $\Phi(k, \omega)$</td>
<td></td>
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<tr>
<td>Present paper</td>
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<td>$Re_\theta = [4.8; 21] \times 10^3$</td>
<td>Space–time</td>
<td>TMS, TT</td>
<td>$U_1$, $K$, $\overline{u_2u_2}$, $\Phi_{22}$</td>
<td>$R_{pp}(r_s, \Delta t)$, $\Phi(k, \omega)$</td>
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</table>

**Table 2.** Wall pressure spectra models solving the Poisson equation for the pressure fluctuations. Flow: ZPG, zero pressure gradient; APG, favourable pressure gradient; TBL, turbulent boundary layer; C, channel; P, pipe; BFS, backward-facing step; FP, over a flat plate; Cy., over a cylinder; A, over an airfoil. Reynolds number: $Re_\tau = u_\tau \delta / \nu$; $Re_\theta = U_\infty \delta / \nu$; $Re_\varepsilon = U_\infty c / \nu$, $Re_D = u_D D / \nu$ and $Re_\ell = U_\infty a / \nu$ with: $u_\tau$, friction velocity; $U_\infty$, free-stream velocity; $\delta$, boundary layer thickness; $\theta$, momentum thickness; $c$, chord length; $D$, pipe diameter; $a$, cylinder radius; $\nu$, kinematic viscosity. Input: input data from RANS solutions, $U_1$, mean streamwise velocity; $\overline{u_2u_2}$, Reynolds stresses; $K$, kinetic energy; $\epsilon$, turbulent dissipation rate; $dP/dx_1$, mean pressure gradient. Modelled terms: $C_{ij}$, space–time velocity correlation coefficients; $\Phi_{22}$, spectral correlation function associated with the vertical velocity fluctuations; $L_2$, vertical correlation length. Output: $\Phi_{pp}(\omega)$, wall pressure frequency spectrum; $\Phi(k, \omega)$, wall pressure wavenumber–frequency spectrum; $R_{pp}$, wall pressure correlations; $r_s$, space separation vector; $\Delta t$, time separation.
using DNS data for a fully developed turbulent channel flow at low Reynolds number, that the TMS and TT terms are of the same order throughout the wavenumber range. In order to take into account the TT terms, Lee et al. (2009) tried to include heuristically a nonlinear source term, based on the DNS results of Kim (1989), in a spectral relation. They obtained satisfactory results for a flow over a backward-facing step. As an alternative, a more accurate and general way to estimate the TT terms contribution may be to solve the Poisson equation in the space–time domain. This approach is performed by Peltier & Hambric (2007) who proposed a stochastic model to estimate wall pressure spectra. A space–time solution to the Poisson equation for wall pressure correlations is used and the pressure spectrum is obtained by their Fourier transform. The TMS and TT source terms are expressed as a double integral over the space above the wall. The relation of Peltier & Hambric (2007) involves the gradient of the mean streamwise velocity in the wall-normal direction and four space–time velocity correlation functions. These latter are modelled as a product of velocity correlation coefficients (denoted by $C_{ij}$), which are obtained using a semi-empirical model developed by Gavin (2002), and Reynolds stresses which are obtained by RANS solutions. Their model was able to correctly predict the effects of pressure gradients on wall pressure frequency spectra for a channel flow. Their results show that the TT contribution is smaller than the TMS one but not negligible. Monté (2013) applied a similar method to estimate the wavenumber–frequency spectrum for TBL flows over towed underwater antennas.

In the present work, we propose a new model to compute the pressure correlations, referred to as the Kriging-based elliptic extended anisotropic model (KEEAM), which relies on an expression similar to Peltier & Hambric (2007) for the pressure correlations and a new closure model for the velocity correlations based on the DNS results of Sillero, Jiménez & Moser (2014). RANS data, obtained with Reynolds stress models, are also used as an input. The evaluation of wall pressure correlations is detailed in § 2 and the velocity correlation model is described in § 3. The present work focuses on the validation of the method and the velocity correlation model for a TBL flow over a flat plate. The results are presented in § 4 where they are discussed. Contrary to Peltier & Hambric (2007), both frequency and wavenumber–frequency spectra are computed. The influence of both the closure model and the Reynolds number on the pressure spectrum are investigated. The results are also compared to other approaches previously described.

### 2. Wall pressure correlations and spectrum

The instantaneous velocity components in the streamwise ($x_1$), wall-normal ($x_2$) and spanwise ($x_3$) directions are denoted by $\tilde{u}_1$, $\tilde{u}_2$, $\tilde{u}_3$, respectively. The instantaneous pressure is indicated by $\tilde{p}$. Points on the wall are denoted by $x_s$ and $y_s$. Points $x$ and $y$ are located in the volume above the wall (see figure 1). Times are denoted by $t$ and $\tau$. An instantaneous variable $\tilde{a}$ is decomposed into mean part and fluctuation as: $\tilde{a} = A + a$ (Reynolds decomposition). Mean field quantities are represented in upper case and fluctuations in lower case.

#### 2.1. Equation for the wall pressure correlation

##### 2.1.1. General formulation

The Poisson equation for the instantaneous pressure is obtained by taking the divergence of the incompressible Navier–Stokes equations. By integrating the resulting equation using an appropriate Green function $G$ and by neglecting the normal
derivative of the pressure at the wall, a solution for the wall pressure can be written (Kraichnan 1956; Chassaing 2000):

\[
\frac{1}{\rho} \tilde{p}(x_s, t) = - \int_{\Omega_s} \frac{\partial}{\partial x_i} \left( \tilde{u}_i(x, t) \frac{\partial \tilde{u}_i(x, t)}{\partial x_j} \right) G(x, x_s) \, dx,
\]

(2.1)

where \( \rho \) is the fluid density. This equation shows that the surface pressure is determined by contributions from all parts of the velocity field over the wall.

The expression of the wall pressure fluctuations is obtained by applying Reynolds decomposition to separate the mean and fluctuating wall pressures and then subtracting the mean pressure:

\[
\frac{1}{\rho} p(x_s, t) = \int_{\Omega_s} \left[ 2 \left( \frac{\partial U_i(x)}{\partial x_j} u_j(x, t) \right) \frac{\partial G(x, x_s)}{\partial x_i} \right. \\
- \left. (u_i(x, t)u_j(x, t) - \overline{u_iu_j(x)}) \frac{\partial^2 G(x, x_s)}{\partial x_i \partial x_j} \right] dx.
\]

(2.2)

In (2.2), the spatial derivative were transferred to the Green function via applications of integration by parts and the divergence theorem (see Peltier & Hambric 2007). The overbar indicates an ensemble average.

The space–time correlations of the wall pressure fluctuations are obtained from the ensemble average of the product of (2.2) evaluated at two different points and two different times \((x_s, t)\) and \((y_s, \tau)\):

\[
\frac{1}{\rho^2} p(x_s, t)p(y_s, \tau) = \int_{\Omega_s} \int_{\Omega_s} \left[ \frac{4}{u_jv_j} \frac{\partial G(x, x_s)}{\partial x_i} \frac{\partial G(y, y_s)}{\partial y_k} \right. \\
- 2 \frac{\partial U_i}{\partial x_j} \frac{\partial G(x, x_s)}{\partial x_i} \frac{\partial^2 G(y, y_s)}{\partial y_k \partial y_l} \\
- 2 \frac{\partial V_k}{\partial y_l} \frac{\partial G(y, y_s)}{\partial y_k} \frac{\partial^2 G(x, x_s)}{\partial x_i \partial x_j} \\
+ \left. \left( u_iu_jv_kv_l - \overline{u_iu_j} \overline{v_kv_l} \right) \frac{\partial^2 G(x, x_s)}{\partial x_i \partial x_j} \frac{\partial^2 G(y, y_s)}{\partial y_k \partial y_l} \right] \, dy \, dx,
\]

(2.3)

with \( U_i(x) \) the mean velocity at point \( x \); \( V_k(y) \) the mean velocity at point \( y \); \( u_i(x, t) \) the fluctuating velocity at point \( x \) and time \( t \); \( v_j(y, \tau) \) the fluctuating velocity at point \( y \) and time \( \tau \).
Assuming that the joint distribution of the fluctuating velocities is normal, equation (2.3) can be simplified. Indeed, the third-order correlation tensor is zero and the fourth-order correlation tensor can be written as a function of the second-order correlation tensor:

\[
\begin{align*}
\overline{u_i u_j v_k v_l} &= \overline{u_i u_j} \overline{v_k v_l} + \overline{u_i v_k} \overline{v_j v_l} + \overline{u_i v_l} \overline{v_j v_k}.
\end{align*}
\] (2.4)

This hypothesis is not accurate for the velocity but it allowed Peltier & Hambric (2007) to obtain satisfactory results for wall pressure correlations.

As a remark, in the case of isotropic turbulence, only the fourth-order correlation tensor appears in the pressure correlation equation. This tensor is expressed as a function of the second-order correlation tensor as in the case of a normal distribution but nothing is assumed for the third-order tensor. This is referred to as the quasi-normality hypothesis (Millionschikov 1941), which is known to yield accurate results for the pressure evaluation (Batchelor 1951; Lesieur, Ossia & Métais 1999; Meldi & Sagaut 2013).

Finally, with the normality hypothesis, the space–time correlations of the wall pressure fluctuations take the form:

\[
\begin{align*}
\frac{1}{\rho^2} p(x_s, t)p(y_s, \tau) &= \int_{\Omega_1} \int_{\Omega_2} \left[ 4 \frac{\partial U_i}{\partial x_j} \frac{\partial V_k}{\partial y_l} \frac{\partial G(x, x_s)}{\partial x_i} \frac{\partial G(y, y_s)}{\partial y_k} \right. \\
&\quad + 2 \overline{u_i v_k} \frac{\partial^2 G(x, x_s)}{\partial x_i \partial x_j} \frac{\partial^2 G(y, y_s)}{\partial y_k \partial y_l} \left. \right] dy dx.
\end{align*}
\] (2.5)

Two types of terms can be identified in (2.5): the first group on the right-hand side denotes turbulence–mean shear (TMS) interactions source terms and the second group on the right-hand side represents turbulence–turbulence (TT) interactions source terms.

In order to estimate turbulent pressure correlations from (2.5), several quantities are required: a Green function, the mean velocity field at two different points and two-point and two-time velocity correlations \( u_i(x, t)v_k(y, \tau) \). The velocity field can be obtained by RANS solutions but the two-point and two-time velocity correlations (or space–time velocity correlations) have to be modelled.

The two-point two-time velocity correlations, \( R_{ik}(x, y, t, \tau) = \overline{u_i(x, t)v_k(y, \tau)} \), can be written as a function of Reynolds stresses (i.e. one-point velocity correlations) and a space–time correlation coefficient:

\[
R_{ik}(x, y, t, \tau) = \sqrt{\overline{u_i u_i}}(x) \sqrt{\overline{v_k v_k}}(y) \ C_{ik}(x, y, t, \tau).
\] (2.6)

Reynolds stresses can be obtained from RANS solutions and a closure model is needed for the space–time correlation coefficients. This model is detailed in § 3.

2.1.2. Flow over a flat plate

We consider the case of a fully developed turbulent boundary layer over a flat rigid plate in a flow with zero pressure gradient at low Mach number. In this flow, the boundary layer thickness increases slowly in the streamwise direction and the velocity field can be considered homogeneous in the \((x_1, x_3)\) plane (parallel to the wall) with a good approximation. Moreover, we assume that the flow is statistically two-dimensional so Reynolds stresses are reduced to four terms (Pope 2000): \( \overline{u_i u_i}, \)
\( \rho_{\text{pp}}(x_0, t)p(x_1, \tau) = \int_{R_1} \int_{R_2} \left[ 4 \frac{\partial U_1}{\partial x_2} \frac{\partial V_1}{\partial y_2} \sqrt{u_{2u_2}} \sqrt{u_{2v_2}} C_{22} \frac{\partial G(x, x_s)}{\partial x_1} \frac{\partial G(y, y_s)}{\partial y_1} \right] \, dy \, dx. \)  \hfill (2.7)

Finally, the Green function for a flat plate is:

\[ G(x, x_s) = -\frac{1}{2\pi} \frac{1}{||x - x_s||}. \quad \hfill (2.8) \]

The relation for pressure correlations (2.7) is slightly different from the one of Peltier & Hambric (2007) because their definition of the velocity correlation coefficients differs from (2.6). There is also a difference for the velocity gradients: Peltier & Hambric (2007) assume that \( \partial U_1/\partial x_2 \approx \partial V_1/\partial y_2 \) so only \( \partial U_1/\partial x_2 \) appears in their equation.

### 2.2. Wall pressure spectrum

We assume that wall pressure correlations can be written as a function of the spatial separation \( r_s = y_s - x_s \) and the time separation \( \Delta t = \tau - t \):

\[ p(x_0, t)p(y_0, \tau) = R_{\text{pp}}(r_s, \Delta t). \quad \hfill (2.9) \]

The wall pressure wavenumber–frequency spectrum is commonly obtained by the Fourier transform of the pressure fluctuation correlations:

\[ \Phi(k_1, k_3, \omega) = \frac{1}{(2\pi)^3} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} R_{\text{pp}}(r_{s1}, r_{s3}, \Delta t)e^{-i(k_1 r_{s1} + k_3 r_{s3} + \omega \Delta t)} \, dr_{s1} \, dr_{s3} \, d\Delta t. \quad \hfill (2.10) \]

The frequency spectrum is related to the pressure correlations and the wavenumber–frequency spectrum by:

\[ \Phi_{\text{pp}}(\omega) = \frac{1}{2\pi} \int_{-\infty}^{\infty} R_{\text{pp}}(0, 0, \Delta t)e^{-i\omega \Delta t} \, d\Delta t = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \Phi(k_1, k_3, \omega) \, dk_1 \, dk_3. \quad \hfill (2.11) \]

### 2.3. Numerical method

To evaluate (2.5) or (2.7), a double volume integral must be estimated numerically. Moreover, in the case of a flow over a flat plate, the Green function and its derivatives are singular for \( x = x_s \) and \( y = y_s \). Since the velocity and the Reynolds stresses are zero on the wall, numerical problems occur only when \( x \) and \( y \) are close to \( x_s \) and \( y_s \) but above the wall. In this case, the integrand varies strongly. To take these variations into account, we use an adaptive quad-tree grid. This grid generation is based on an error estimator of the integration of the Green function derivatives. Integrations on each subspace of the grid are realized with a seventh-order discrete Gauss–Legendre method. The size of the integration domain is taken sufficiently large to guarantee the convergence of the results.
Besides, to obtain the pressure spectrum for broad frequency and wavenumber ranges, a considerable number of pressure correlation values are required. In order to reduce the computation time, we evaluate the pressure correlations only at a few points and create a surrogate model using Kriging (Krige 1951). A self-adaptive sampling strategy is used to increase the accuracy of the surrogate model (see Braconnier et al. 2011; Margheri & Sagaut 2016). Figure 2 shows the steps of the method. Starting from an initial sample with a uniform distribution, the pressure correlations are calculated. Then, using the leave-one-out cross-validation process, a local error is estimated for each point. Precisely, the sample is split into a training set (used to build the surrogate model with Kriging) and a test set (with only one point). The local error is evaluated between the pressure correlation value at the test point, \( R_{pp}(X_{test}) \), and the value obtained with the surrogate model at the test point, \( R'_{pp}(X_{test}) \):

\[
LE = \frac{|R_{pp}(X_{test}) - R'_{pp}(X_{test})|}{R_{pp}(0, 0, 0)}.
\]  (2.12)

\( R_{pp}(0, 0, 0) \) is the maximal value of the pressure correlations and is computed in the initial sample. This error is estimated for each element of the sample and points are added with quad-tree sampling where the error is maximal. The pressure correlations are evaluated for the new points and the previous steps are repeated until a satisfactory error is reached. A minimal value is imposed for the distance between points in order to avoid accumulations near discontinuities or high gradients. Finally, the surrogate model is used to interpolate the pressure correlations on a fine grid.

For two-dimensional results with a local error lower than 5 %, approximately 500 values of pressure correlations are required to compute a streamwise wavenumber–frequency spectrum on a 2048 × 2048 grid. An example of the pressure frequency spectrum obtained with and without Kriging is presented in § 4.
3. Two-point two-time velocity correlation model

3.1. Velocity correlation characteristics

Space and time dependencies are here systematically specified so the fluctuating velocity at point $y$ and time $\tau$ is also denoted $u_i$ without any confusion. Since the turbulent boundary layer is considered as stationary in time, the space–time velocity correlations can be written:

$$R_{ij}(x, y, t, \tau) = R_{ij}(x, r, \Delta t) = u_i(x, t)u_j(x + r, t + \Delta t), \quad (3.1)$$

with $r = y - x$ the spatial separation vector and $\Delta t = \tau - t$ the time separation.

Some conditions can be written for the form of the velocity correlations, which in turn, have an impact on the form of the correlation coefficients $C_{ij}$. It is assumed that the motion at one point of the fluid and one time is statistically independent of the motion at a point sufficiently distant in space or time. Thus, $R_{ij}$ should become negligible for large values of $|r|$ or $1t$. $R_{ij}$ also vanishes when a point is located on the wall (i.e. when $x_2 = 0$ or $r_2 = -x_2$). The former requirement applies directly to $C_{ij}$ and the last one is automatically satisfied using (2.6).

The form of the velocity correlations should also satisfy the continuity condition (Rotta 1962):

$$\frac{\partial}{\partial r_j}R_{ij}(x, r, \Delta t) = 0. \quad (3.2)$$

Since the instantaneous flux across any closed surface is zero and $R_{ij}$ tends to zero at large $|r|$, the integration of (3.2) with respect to all values of $r_2$ and $r_3$ for a constant value of $r_1$ leads to (Rotta 1962; Townsend 1980):

$$\int_{-\infty}^{\infty} \int_{-x_2}^{\infty} R_{i1}(x, r, \Delta t) \, dr_2 \, dr_3 = 0, \quad i = 1, 2, 3. \quad (3.3)$$

Equation (3.3) implies that $R_{i1}$ has positive and negative values within the plane $(r_2, r_3)$. Similar equations can be written for the other velocity correlations by integrating over planes $(r_1, r_3)$ and $(r_1, r_2)$. The condition (3.3) does not apply exactly to $C_{ij}$ even if the coexistence of negative and positive regions can be observed (Sillero et al. 2014).

Finally, $R_{ij}$ should tend to one-point correlations for $r = 0$ and $\Delta t = 0$:

$$R_{ij}(x, 0, 0) = \bar{u}_i\bar{u}_j(x), \quad (3.4)$$

which implies for the correlation coefficient $C_{ij}$ that:

$$C_{ii}(x, 0, 0) = 1 \quad (3.5)$$

$$C_{ij}(x, 0, 0) = \frac{\bar{u}_i\bar{u}_j(x)}{\sqrt{\bar{u}_i\bar{u}_j(x)\bar{u}_j\bar{u}_j(x)}}, \quad i \neq j. \quad (3.6)$$

For the sake of simplicity, we introduce $\tilde{C}_{i2}$ such as:

$$\tilde{C}_{12}(x, r, \Delta t) = \frac{C_{12}(x, r, \Delta t)}{C_{12}(x, 0, 0)}, \quad (3.7)$$

which satisfies $\tilde{C}_{12}(x, 0, 0) = 1$.

The previous requirements help to define a form for the velocity correlation coefficients. To model these space–time coefficients we proceed in two steps:

(i) modelling the spatial evolution of the correlation coefficients (§ 3.2);

(ii) relating spatial correlations to space–time correlations (§ 3.3).
To model the spatial evolution of the velocity correlation coefficients we proposed a semi-analytical model (for $C_{ii}$ and $\tilde{C}_{12}$) using the DNS data of Sillero et al. (2014) as reference. The spatial characteristics of these data are detailed in § 3.2.1. The main characteristics of the isotropic model and the simplified anisotropic model (SAM) are given in §§ 3.2.2 and 3.2.3 respectively. Both approaches are used as a basis for the new $C_{ij}$ form, referred to as extended anisotropic model (EAM), proposed in this work and presented in § 3.2.4. Lastly, these three models are compared in § 3.2.5.

3.2.1. Spatial characteristics from reference data

Sillero et al. (2014) give detailed results of two-point statistics, particularly velocity correlation coefficients, for the buffer, logarithmic and outer regions of turbulent boundary layers. Data are available for three Reynolds number $Re_\theta \approx 4000, 4860$ and 6500 (where $Re_\theta = U_\infty \theta / \nu$, $U_\infty$ is the free-stream velocity, $\nu$ is the kinematic viscosity and $\theta$ is the momentum thickness). These data can be found at http://torroja.dmt.upm.es/turndata/blayers/high_re/correlations.

An example of sections of the four correlation coefficients in the plane $(r_1, r_2)$, obtained by post-processing the Sillero et al. (2014) data, are presented in figure 3. The variables are scaled by the boundary layer thickness $\delta$. Correlation coefficients are
more or less elongated structures tilted with respect to the wall. Sillero et al. (2014) observed that their size in the three directions as well as their inclination angle depend on the variable being considered and the wall distance \( x_2 \). The streamwise velocity correlation coefficient \( \mathcal{C}_{11} \) is more elongated than the others in the streamwise direction (see figure 3). Its maximum length observed for the weakly correlated structures (\( \mathcal{C}_{11} \approx 0.025 \)) is of the order of \( 7\delta \) (see also the experimental results of Tutkun et al. 2009) whereas for \( \mathcal{C}_{22} \) and \( \mathcal{C}_{33} \) the maximum streamwise length is of the order of \( 2\delta - \delta \). On the contrary, these three coefficients have similar spanwise widths of the order of \( \delta \). However, the size of these coefficients in each direction evolve differently with the wall distance \( x_2 \). The value of the inclination angle depends on the correlation level being considered (see figure 3) but Sillero et al. (2014) define a characteristic inclination angle which is the maximum angle observed as a function of the wall distance \( x_2 \). For \( \mathcal{C}_{11} \), the angle is \( \alpha_1 \approx 10^\circ \). For \( \mathcal{C}_{33} \), the inclination angle tends quickly to a steeper one, \( \alpha_3 \approx 32^\circ \). Both angles are principally uniform across the boundary height. \( \mathcal{C}_{22} \) is mainly vertical except in a region close to the wall where \( \alpha_2 \approx \alpha_1 \). Finally, \( \mathcal{C}_{12} \) looks like a shorter version of \( \mathcal{C}_{11} \) (Sillero et al. 2014).

3.2.2. Isotropic model
In the case of isotropic turbulence, the velocity correlations can be expressed as (Batchelor 1959; Sagaut & Cambon 2008):

\[
R_{ij}(r) = u_i(x)u_j(x + r) = u^2 \left( f(r) - g(r) \frac{r_i r_j + g(r) \delta_{ij}}{r^2} \right),
\]

(3.8)

where \( u^2 = 2\mathcal{K}/3 \) is a scaling factor function of the kinetic energy \( \mathcal{K} \); \( \delta_{ij} \) is the Kronecker delta; \( f \) and \( g \) are the longitudinal and transverse correlation functions respectively, \( r \) is the norm of the spatial separation vector \( r \).

The correlation function \( g \) is linked to \( f \) via the continuity condition (3.2) which leads to:

\[
g(r) = f(r) + \frac{r}{2} \frac{\partial f(r)}{\partial r}.
\]

(3.9)

These functions have to satisfy \( f(0) = g(0) = 1 \). A common expression for the correlation function \( f \) is (Panton & Linebarger 1974; Peltier & Hambric 2007):

\[
f(r) = \exp \left( -\frac{r}{L} \right),
\]

(3.10)

where \( L \) is a correlation length. Different formulas can be used for \( L \) such as Taylor microscale \( \lambda_f = \sqrt{15\nu u^2/\varepsilon} \) or a characteristic integral scale \( L_u = \mathcal{K}^{3/2}/\varepsilon \) (Sagaut & Cambon 2008) which are both functions of the kinetic energy dissipation rate \( \varepsilon \). With (3.10), the condition that \( R_{ij} \) should vanish for large values of \( |r| \) is satisfied.

From (3.8), the correlation coefficients can be written:

\[
\mathcal{C}_{ii}(r) = \frac{f(r) - g(r)}{r^2} r_i r_i + g(r)
\]

(3.11)

\[
\tilde{\mathcal{C}}_{12}(r) = \frac{f(r) - g(r)}{r^2} r_1 r_2.
\]

(3.12)

The four correlation coefficients obtained with the isotropic model are shown in the plane \((r_1, r_2)\) in figure 4. The DNS results presented in figure 3 are also plotted (grey dotted lines) as a reference. The correlation length used for this figure is proportional to \( L_u \). It shows the necessity to take into account the inclination angles as well as the anisotropy effects. Indeed, only one correlation length for all the \( \mathcal{C}_{ij} \) is not sufficient. Besides, the form of the model for \( \tilde{\mathcal{C}}_{12} \) is clearly not adapted.
3.2.3. Simplified anisotropic model

The simplified anisotropic model, presented in Peltier & Hambric (2007), was developed by Gavin (2002). It is an empirical model based on measurements of turbulent velocity correlations in the outer region of a high Reynolds number boundary layer. In order to take into account the ellipsoidal shape and the inclination of the velocity correlations to the wall, Gavin (2002) defined a rotation angle $\theta$ and stretching coefficients $\gamma_i$ to modify the separation vector $r = y - x$ and then used a form similar to the isotropic model (3.8).

Firstly, a rotation is applied to the separation vector:

$$\xi = \begin{bmatrix} \cos(\theta) & \sin(\theta) & 0 \\ -\sin(\theta) & \cos(\theta) & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} r_1 \\ r_2 \\ r_3 \end{bmatrix}. \quad (3.13)$$

Then, a separation length is defined using the stretching coefficients:

$$r^* = \sqrt{\left(\frac{\xi_1}{\gamma_1}\right)^2 + \left(\frac{\xi_2}{\gamma_2}\right)^2 + \left(\frac{\xi_3}{\gamma_3}\right)^2}. \quad (3.14)$$
Finally, velocity correlation coefficients are written by adapting the model for the isotropic turbulence:

\[ C_{ij}(\xi) = \xi_i \xi_j r^* \left[ f(r^*) - g(r^*) \right] + \delta_{ij} g(r^*), \]  

(3.15)

with \( f(r^*) = \exp(-r^* / L) \). The transverse correlation function \( g \) is obtained using (3.9).

According to Peltier & Hambric (2007), Gavin used \( L = 0.35 \delta \) for the correlation length which is valid only in the outer region of the TBL. Instead, they use \( L(x_2) \equiv 0.54 L_u(x_2) \). The values of the SAM parameters determined by Gavin are presented in table 3. Since no values are given for \( \tilde{C}_{12} \), we use \( \tilde{C}_{12} \approx \sqrt{\tilde{C}_{11} \tilde{C}_{22}} \) as Peltier & Hambric (2007).

The resulting correlation coefficients are plotted in figure 5 and compared with the DNS results of Sillero et al. (2014). The SAM angles appear to be too large, particularly for \( \tilde{C}_{11} \). However, it depends on the correlation level being considered. Except \( \tilde{C}_{22} \), the \( \tilde{C}_{ij} \) coefficients decay too fast which implies that the correlation length is too small for this value of \( x_2 \). Other drawbacks of the SAM are that the lengths of the coefficients in each direction evolve similarly as functions of \( x_2 \) which is not expected (see Sillero et al. 2014). Moreover the definition of \( \tilde{C}_{12} \) makes the existence of negative values impossible which is inconsistent with the continuity condition.

### 3.2.4. Extended anisotropic model

By taking into account the advantages and disadvantages of the previous approaches, we propose a new model, the extended anisotropic model (EAM). Its form is based on the SAM with some improvements. Instead of using the separation length \( r^* \) with the SAM stretching coefficients, we define a scaled separation vector \( \tilde{r} \) with three correlation lengths \( L_i \):

\[ \tilde{r}_i(\xi, x_2) = \frac{\xi_i}{L_i(x_2)} \]  

(3.16)

\[ \tilde{r}(\xi, x_2) = \sqrt{\left( \frac{\xi_1}{L_1(x_2)} \right)^2 + \left( \frac{\xi_2}{L_2(x_2)} \right)^2 + \left( \frac{\xi_3}{L_3(x_2)} \right)^2}. \]  

(3.17)

Then we use an expression similar to (3.11) for the diagonal components of the correlation coefficient tensor, but for \( \tilde{C}_{12} \) we choose an anisotropic form, leading to an additional term in the right-hand side:

\[ C_{ii}(\tilde{r}) = \left( \frac{\tilde{r}_i}{\tilde{r}} \right)^2 \left[ f(\tilde{r}) - g(\tilde{r}) \right] + g(\tilde{r}) \]  

(3.18)

\[ \tilde{C}_{12}(\tilde{r}) = \frac{\tilde{r}_1 \tilde{r}_2}{\tilde{r}^2} \left[ f(\tilde{r}) - g(\tilde{r}) \right] + g(\tilde{r}) \]  

(3.19)

---

<table>
<thead>
<tr>
<th>( C_{ij} )</th>
<th>( \theta )</th>
<th>( \gamma_1 )</th>
<th>( \gamma_2 )</th>
<th>( \gamma_3 )</th>
</tr>
</thead>
<tbody>
<tr>
<td>( C_{11} )</td>
<td>20°</td>
<td>1.000</td>
<td>0.700</td>
<td>0.520</td>
</tr>
<tr>
<td>( C_{22} )</td>
<td>90°</td>
<td>0.500</td>
<td>0.525</td>
<td>0.350</td>
</tr>
<tr>
<td>( C_{33} )</td>
<td>35°</td>
<td>0.800</td>
<td>0.220</td>
<td>0.400</td>
</tr>
</tbody>
</table>

**Table 3.** SAM parameters.
Figure 5. Streamwise \((r_1, r_2)\) sections of the correlation coefficients obtained with the SAM \((L = 0.54L_u)\) for \(r_3 = 0, \Delta t = 0, x_2/\delta = 0.065\) and \(Re_\theta \approx 4860\) (the flow is from left to right). Positive contours (— —) are from 0.1 to 0.9 with increments of 0.1. The grey dotted contours are the DNS results of Sillero et al. (2014) at the same height and Reynolds number.

The longitudinal correlation function is:

\[
    f(\tilde{r}) = \begin{cases} 
    \exp(-\tilde{r}), & \text{for } C_{22}, C_{33} \text{ and } \tilde{C}_{12} \\
    \alpha \exp(-\tilde{r}) + (1 - \alpha) \exp(-\beta \tilde{r}), & \text{for } C_{11},
    \end{cases}
\]

with \(\beta = 10\) and \(\alpha\) is a function of the wall distance \(x_2\). The transverse correlation function \(g\) is obtained using (3.9).

The expression of \(f\) for \(C_{11}\) takes into account the more elongated shape of this coefficient in the streamwise direction. The inclination angle are \(\theta_{11} = 10^\circ\) for \(C_{11}\) and \(\theta_{12} = 115^\circ\) for \(\tilde{C}_{12}\). For the normal and spanwise velocity correlations, inclination angles are linear functions of \(x_2^+ (x_2^+ = x_2 u_\tau / \nu\) with \(u_\tau\) the friction velocity):

\[
    \theta_{22} = \begin{cases} 
    \text{linear evolution from } 5^\circ \text{ to } 90^\circ, & x_2^+ \leq 100 \\
    90^\circ, & x_2^+ > 100
    \end{cases}
\]

(3.21)

\[
    \theta_{33} = \begin{cases} 
    \text{linear evolution from } 5^\circ \text{ to } 30^\circ, & x_2^+ \leq 100 \\
    30^\circ, & x_2^+ > 100.
    \end{cases}
\]

(3.22)
The evolutions of the parameter $\alpha$ and the correlation lengths $L_i$ for each coefficient are presented in figures 6–8. These parameters are calibrated using the DNS data of Sillero et al. (2014) for $Re_\theta \approx 4860$, available at http://hal.dmt.upm.es/raw_database/Boundary_Layer/High-Reynolds/HR2-3D-Correlations. We use an inner and an outer regions, defined as $x^+_2 \leq 100$ and $x^+_2 > 100$, where the parameters are scaled with wall units and outer units respectively. The evolutions of equivalent correlation lengths for the SAM ($L_i \equiv 0.54 \gamma L_n$) are also included in figures 6 and 7 (only for $C_{11}$). The evolutions of the EAM correlation lengths are very different from the SAM ones.

The results obtained with the EAM in the plane ($r_1$, $r_2$) are displayed in figure 9 where they are compared with the Sillero et al. (2014) data. The size of the coefficients and their inclination angles appear to be more adapted, particularly for $C_{11}$. Nevertheless, some discrepancies are still visible for the other coefficients. For instance, these coefficients decay a little too quickly. As the other models, the EAM also fails to accurately take into account the decrease of the coefficient values close to the wall.

Figure 10 shows the evolution of the correlation coefficients, obtained with the EAM, the SAM and DNS data, with respect to $x_2$ for a fixed separation distance. The EAM results are better than the SAM ones for $C_{33}$ and $\tilde{C}_{12}$. For $C_{11}$, both results are quite close for $0.2 < x_2/\delta < 0.6$. For $x_2/\delta < 0.2$ and $x_2/\delta > 0.6$, the coefficient $C_{11}$ estimated with the EAM is closer to DNS data. For the $C_{22}$, the values obtained with the EAM are better for $x_2/\delta > 0.6$.

3.2.5. Model comparison

A comparison of the EAM, SAM and the isotropic model is presented in figure 11. Results are compared with the DNS data of Sillero et al. (2014). As expected, the values obtained with the EAM are closer to the DNS data, particularly for $C_{11}$ in the streamwise direction. The isotropic model and the SAM are not adapted for $C_{11}$ in this direction. For $C_{22}$ and $C_{33}$, all models have similar shapes but the scales are
Figure 7. Correlation lengths for the EAM scaled in outer units: — (black), $C_{11}$; - - - , $C_{22}$; - - - - , $C_{33}$; - - - - - - , $C_{12}$. The grey lines represent equivalent correlation lengths ($L_i \equiv 0.54 \gamma_i L_u$) for the SAM for $C_{11}$.

Figure 8. Parameter $\alpha$ for the EAM scaled in wall units (a) and outer units (b).

different. The results obtained with the isotropic model for $L = \lambda_f$ and the SAM for $L = 0.54L_u$ are too narrow whereas both models give correlations which are too wide for the other correlation lengths ($L = 0.54L_u$ and $L = 0.35\delta$ respectively).

As shown in figures 3 and 11, the shape of the velocity correlation coefficients is complex and the ellipsoidal form of the model derived from the isotropic turbulence theory is not perfectly adapted, even with inclination angles. Some improvements involve more complex functions than the correlation functions $f$ and $g$, in order to take into account the asymmetry of the correlations for separations in the normal direction. However, this is far beyond the scope of the present work. Further comparisons of the $C_{ij}$ models are made in § 4.2 where their influence on the pressure frequency spectrum is discussed.
Figure 9. Streamwise ($r_1$, $r_2$) sections of the correlation coefficients obtained with the EAM for $r_3 = 0$, $\Delta t = 0$, $x_2/\delta = 0.065$ and $Re_\theta \approx 4860$ (the flow is from left to right). Positive contours (——) are from 0.1 to 0.9 with increments of 0.1. The grey dotted contours are the DNS results of Sillero et al. (2014) at the same height and Reynolds number.

3.3. Space–time correlations model

Several theories and models exist for turbulent space–time velocity correlations (see the reviews of Wallace (2014) and He, Jin & Yang (2017)). In this work, we compare two models which relate spatial correlations to space–time correlations.

The first one is Taylor’s frozen turbulence hypothesis (Taylor 1938), stating that turbulent structures are carried past a fixed point by the mean flow at the convection velocity $U_c$ without any major changes. With this hypothesis, space–time velocity correlations for a spatial separation $r_1$ in the streamwise direction and a time separation $\Delta t$ can be written:

$$R_{ij}(x, r_1, \Delta t) = R_{ij}(x, r_1 - U_c \Delta t, 0).$$  

(3.23)

According to He & Zhang (2006), Taylor’s frozen flow hypothesis is only a first approximation which has limitations. Indeed, with this hypothesis, iso-correlation contours of $R_{ij}(x, r_1, \Delta t)$ are straight lines and thus the correlations remain constant even for large values of $r_1$ and $\Delta t$. This result is not consistent because correlation functions decrease with increasing space and time separations. Moreover, several simulations and experiments (He & Zhang 2006; Guo et al. 2012; Wallace 2014; Wang, Guan & Jiang 2014) show that the iso-correlation contours are not straight
FIGURE 10. Evolution of the correlation coefficients with respect to $x_2$ for $r = (0.030\delta, 0, 0.037\delta)$, $\Delta t = 0$ and $Re_\theta \approx 4860$. For each panel: ——, DNS (Sillero et al. 2014); O, EAM; ●, SAM ($L = 0.54L_u$).

lines but elongated ellipses. Nevertheless, Taylor’s hypothesis is often used (Panton & Linebarger 1974; Lee et al. 2005; Bertagnolio et al. 2014) due to its simplicity and ease of implementation.

The elliptic model of He & Zhang (2006) is formulated using a second-order Taylor series expansion of $R_{ij}(x, r_1, \Delta t)$. This model relates space correlation to space–time correlations via two parameters, the convection velocity $U_c$ and a velocity $V$ which characterizes the distortion of small-scale eddies:

$$R_{ij}(x, r_1, \Delta t) = R_{ij}(x, \sqrt{(r_1 - U_c\Delta t)^2 + V^2\Delta t^2}, 0).$$

The iso-correlation contours obtained with this model are elliptic. Contrary to Taylor’s hypothesis, the elliptic model also takes into account the decorrelation induced by the random sweeping of velocity fluctuations and the shearing of mean flows which distort small-scale eddies (He et al. 2017). This model was successfully applied by Zhao & He (2009) and Guo et al. (2012) for turbulent channel flows and by Wang et al. (2014) for a TBL flow.

Two different models for the convection velocity are considered. The first one is $U_c = 0.7U_\infty$. Taking a uniform value for the convection velocity is mainly motivated here by its simplicity and ease of implementation. The value $U_c = 0.7U_\infty$ is moreover justified by the DNS results of Choi & Moin (1990), in which it is shown that this value constitutes a reasonable approximation. In more recent efforts, del Álamo &
Jiménez (2009) demonstrate that a uniform convection velocity is not accurate since the convection velocity is a function of the eddy frequency and of their wall distance. Therefore, a second model, taken from Renard & Deck (2015) results for a global correlation-based convection velocity, is evaluated. They obtained a convection velocity which is a function of the wall distance \(x_2\) and is close to the mean velocity except near the wall where it is larger and nearly constant. There are other models available in the literature for \(U_c\), particularly expressions which are functions of the frequency (e.g. Wills 1971; Panton & Linebarger 1974). However their implementation in our space–time correlation model is not straightforward so they are not studied in the present work.

For \(V\), we use the expression proposed by Monté (2013) which defines a velocity \(V\) for each coefficient \(C_{ij}\):

\[
V_{ij}^2(x) = S_{ij}^2(x) \lambda_{ij}^2(x) + \overline{u_i u_j}(x),
\]

(3.25)
where $S_{ij}$ is the mean rate-of-strain tensor and $\lambda_{ij}$ is a Taylor scale defined from Reynolds stresses as: $\lambda_{ij} = \sqrt{15|u_i u_j|\nu/\varepsilon}$.

The influence of the space–time model on the pressure spectrum is discussed in § 4.2.

3.4. Kriging-based elliptic extended anisotropic model

In this work, the EAM and the elliptic model are used to close (2.7). The whole approach to compute the pressure correlations with the numerical method described in § 2.3 is referred to as the Kriging-based elliptic extended anisotropic model (KEEAM). In the following section, the influence of the closure model on the pressure spectrum is discussed. Then the KEEAM is compared to a few models presented in the introduction.

4. Results

4.1. RANS simulation

The RANS data used in this work were generated with the open-source computational fluid dynamics software Code_Saturne developed by EDF R&D (Archambeau, Méchitoua & Sakiz 2004). The elliptic blending Reynolds stress model (EBRSM) (Manceau & Hanjalić 2002) is used for the turbulence closure. With this model, the value of $x_2^+ = u_x x_2/\nu$ for the first cell adjacent to the wall must be lower than 1. Therefore wall functions are not used and the equations are solved in all regions of the boundary layer.

In figure 12, the mean streamwise velocity and the Reynolds stress profiles obtained with the EBRSM are compared with the DNS results of Sillero, Jiménez & Moser (2013) at the same Reynolds number. The agreement of the data are satisfactory although there are slight differences for the streamwise and the spanwise Reynolds stresses.
4.2. Frequency spectrum of the wall pressure fluctuations

4.2.1. Validation of the adaptive sampling method

In order to validate the adaptive sampling strategy, the wall pressure frequency spectrum is computed with and without Kriging. The elliptic model with $U_c = 0.7 U_\infty$ and the EAM are used. The results are presented in figure 13. The spectrum without Kriging is computed with 1024 values of the pressure time correlations whereas only 77 points are necessary to obtain very satisfactory results with Kriging.

4.2.2. Influence of the velocity correlation coefficient model

Wall pressure frequency spectra obtained with the different velocity correlation coefficient models previously described are presented in figure 14. The elliptic model is used to relate spatial correlations to space–time correlations. The Goody (2004) spectrum is also plotted as a reference since it provides satisfactory predictions of the frequency spectrum for a TBL flow over a flat plate (Hwang et al. 2009). Large discrepancies can be observed between the pressure spectra, which shows the strong influence of the correlation coefficients. The choice of the correlation length $L$ has also a great influence for the SAM and the isotropic model. The wall pressure fluctuations estimated with the EAM are satisfactory in the whole frequency range.

The influence of the space–time model (Taylor’s hypothesis or the elliptic model) and the convection velocity model are also studied. The frequency spectra computed with the EAM are shown in figure 15. The pressure spectra obtained with Taylor’s hypothesis and the elliptic model are very close, except at low frequencies and very high frequencies. However, for the wavenumber–frequency spectrum, Taylor’s hypothesis leads to a Dirac delta function so the elliptic model is used in the KEEAM. The convection velocity model has more influence on the form of the pressure spectrum (see figure 16): the pressure levels obtained with $U_c(x_2)$ are larger and the spectrum decay at a lower frequency. The spectrum computed with $U_c = 0.7 U_\infty$ is thus in better agreement with Goody (2004) model, so this expression is retained for the rest of the study.

4.2.3. TMS and TT contributions

The wall pressure spectra with the TMS and TT terms contributions are displayed in figure 17. The magnitude of the TT spectrum is lower than the TMS one, except at
Figure 14. Frequency spectrum of the wall pressure fluctuations at $Re_\theta \approx 4860$. —— black, Goody (2004) model; —— grey, EAM; ––, isotropic (black, $L = \lambda_f$; grey, $L = 0.54L_u$); - - -, SAM (black, $L = 0.35\delta$; grey, $L = 0.54L_u$).

Figure 15. Frequency spectrum of the wall pressure fluctuations at $Re_\theta \approx 4860$ ($U_c = 0.7U_\infty$). ——, elliptic model; - - -, Taylor’s hypothesis.

low frequencies. This is consistent with Kraichnan (1956) and Remmler et al. (2010). However, depending on the precision required, the TT terms should not be neglected at mid frequencies. These results differ from the Chang et al. (1999) results for the streamwise wavenumber pressure spectrum of a fully developed turbulent channel flow at a low Reynolds number. They observed that the TMS and TT terms are of the same order but that the TMS term is greater for the whole streamwise wavenumber range.

4.2.4. Influence of the Reynolds number

The frequency pressure spectra obtained with the KEEAM at two Reynolds numbers are presented in figure 18. They are compared to the Goody (2004) model which is based on experimental pressure spectra measurements on a large range of Reynolds numbers. Since it takes into account the effect of the Reynolds number, it is also used as a reference here. The wall pressure spectra are in agreement with the Goody (2004) model at both Reynolds numbers. It shows that the present model is also able to take accurately into account the influence of the Reynolds number even if the parameters
of the EAM were calibrated with DNS results for a particular Reynolds number. Other predictions of the pressure spectrum were performed with the KEEAM and it gives accurate results (compared to Goody model) for $3 \times 10^3 < Re_\theta < 3 \times 10^4$.

The wall pressure levels predicted by the KEEAM are quite close to the ones obtained with Goody model but there are differences. Indeed, the maximum of the KEEAM spectra are reached at higher frequencies. In the high frequency region, the spectrum evolves as $\omega^{-3}$ whereas Goody model is designed to reproduce a $-5$ slope. According to Blake (1986), the decay rate of the spectrum in this range is due to contributions from the viscous wall region. The EAM could be improved in this area. For $Re_\theta \approx 4860$, the KEEAM does not predict the universal region, characterized by a slope between $-0.7$ and $-1.11$ (Smol’yakov 2000; Goody 2004). However, this region is visible for $Re_\theta \approx 21100$ with a $-0.75$ slope, which is close to the $-0.7$ slope of the Goody spectrum. So the KEEAM reproduced the universal region but it appears at a higher Reynolds number.

4.2.5. Comparison with a spectral formulation of the wall pressure

Besides the Goody (2004) model, the wall pressure spectrum computed is also compared to the model of Bertagnolio et al. (2014) which uses a spectral solution.
FIGURE 18. Frequency spectra of the wall pressure fluctuations computed at two different Reynolds numbers. Black lines are Goody (2004) model and grey lines are the KEEAM. ——, \( \text{Re}_\theta \approx 4860; - - - , \text{Re}_\theta \approx 21100. \)

To the Poisson equation (see table 2). Using Taylor’s frozen flow hypothesis, the frequency pressure spectrum is given by:

\[
\Phi_{pp}(\omega) = \int_{-\infty}^{\infty} \int_{0}^{\infty} \frac{4\rho^2}{U_c} L_2(x_2)\overline{u_2u_2} \left( \frac{\partial U_1}{\partial x_2} \right)^2 k_x^2 k_\|^2 \Phi_{22}(k_x, k_\|) e^{-2k_x x_2} \, dx_2 \, dk_\|, \tag{4.1}
\]

with:

\[
\Phi_{22}(k_1, k_3) = \frac{4}{9\pi} \frac{\beta_1\beta_3}{k_x^2 k_\|^2} \left[ \frac{\beta_1 k_1}{k_x} + \frac{\beta_3 k_3}{k_\|} + \left( \frac{\beta_1 k_1}{k_x} \right)^2 + \left( \frac{\beta_3 k_3}{k_\|} \right)^2 \right]^{1/3} \tag{4.2}
\]

\[
k = \sqrt{k_x^2 + k_\|^2} \tag{4.3}
\]

\[
L_2 = 0.747 \Lambda \tag{4.4}
\]

\[
\Lambda = 0.519 \frac{k_x^{3/2}}{\varepsilon} \tag{4.5}
\]

\[
k_x = \frac{1}{2\Lambda} \tag{4.6}
\]

\[
\beta_i = \frac{\overline{u_i u_i}}{\overline{u_1 u_1}} \tag{4.7}
\]

Since the Bertagnolio et al. (2014) expressions for \( \beta_1 \) and \( \beta_3 \) are not adapted for TBL flows without pressure gradient, we use the ones proposed by Stalnov, Paruchuri & Joseph (2015) instead. We also use their relation for \( k_x \), which is expected to be more accurate. The mean velocity field and the Reynolds stresses are obtained from RANS solutions. A comparison of the frequency pressure spectra is presented in figure 19. Discrepancies can be observed between the Goody (2004) and Bertagnolio et al. (2014) models. The differences at low frequencies are expected since only the TMS term is taken into account in the latter model. Large discrepancies can also be observed between the TMS spectrum computed with the KEEAM and the Bertagnolio et al. (2014) model.
4.3. Wavenumber–frequency spectrum of the wall pressure fluctuations

In the present work, the pressure correlations $R_{pp}(r_{s1}, 0, \Delta t)$ are computed as well as the streamwise wavenumber–frequency spectrum which is given by:

$$
\Phi(k_1, \omega) = \left(\frac{1}{2\pi}\right)^2 \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} R_{pp}(r_{s1}, 0, \Delta t) e^{-i(k_1 r_{s1} + \omega \Delta t)} dr_{s1} \, d\Delta t
$$

$$
= \int_{-\infty}^{\infty} \Phi(k_1, k_3, \omega) \, dk_3. \quad (4.8)
$$

The results, presented in figures 20 and 21, are compared to the empirical models of Corcos (1964) and Chase (1987) (integrated over $k_3$) at two frequencies. The KEEAM is able to predict a convective peak whose amplitude is in agreement with both empirical models. It also predicts a subconvective plateau that is close to the one obtained with the Chase (1987) model. As mentioned before, large discrepancies exist between the empirical models in the subconvective range and there is no consensus on their validity. Nevertheless, the Corcos (1964) model is known to overestimate the pressure spectrum at low wavenumbers (Bonness et al. 2010).

Figures 22 and 23 show the corresponding TMS and TT contributions for the spectrum obtained with the KEEAM. Around the convective peak, the TT terms are lower than the TMS one at both frequencies. On the contrary, in the viscous region the TT terms dominate. In the subconvective range, the TMS and TT contributions are of the same order for $\omega \delta / U_\infty \approx 5$ whereas the TT terms dominate for $\omega \delta / U_\infty \approx 15$. In any case, it shows that the TT contributions should not be neglected at low wavenumbers.

5. Conclusion

In this work, the wall pressure spectrum for a turbulent boundary layer flow without a pressure gradient is computed using a new model, the Kriging-based elliptic extended anisotropic model. The pressure spectrum is the Fourier transform of the space–time pressure fluctuation correlations, which are obtained by solving a Poisson equation. The solution takes into account both the turbulence–mean shear (TMS) and the turbulence–turbulence (TT) interactions, contrary to spectral wall pressure models.
Figure 20. Wavenumber–frequency spectra of the wall pressure fluctuations at $Re_\theta \approx 4860$ and $\omega \delta / U_\infty \approx 5$. ——, KEEAM; - - -, Chase (1987); - ·-, Corcos (1964).

Figure 21. Wavenumber–frequency spectra of the wall pressure fluctuations at $Re_\theta \approx 4860$ and $\omega \delta / U_\infty \approx 15$. ——, KEEAM; - - -, Chase (1987); - ·-, Corcos (1964).

Figure 22. Wavenumber–frequency spectra of the wall pressure fluctuations computed with the KEEAM at $Re_\theta \approx 4860$ and $\omega \delta / U_\infty \approx 5$. ——, total; grey - - -, TT; black - - - , TMS.

Based on Kraichnan (1956). The mean velocity field and the space–time velocity correlations are required to evaluate the wall pressure. As the velocity correlations
cannot be obtained directly from RANS solutions, they are modelled using Reynolds stresses and space–time velocity correlation coefficients. We propose a new model for the spatial correlation coefficients, referred to as the extended anisotropic model, which takes into account the evolution of the coefficients shape and size with the wall distance. It requires a few parameters which are calibrated using the DNS results of Sillero et al. (2014). Spatial correlation coefficients are related to space–time coefficients with the elliptic model of He & Zhang (2006). Then, the pressure correlations are evaluated with a Kriging self-adaptive sampling strategy in order to reduce the computation time.

The EAM is an extension of the simplified anisotropic model of Gavin (2002) that relies on a new expression for the velocity two-point correlation function and new parameters which evolve with the wall distance. Contrary to the SAM, the EAM is adapted for both inner and outer regions of the boundary layer. The EAM is compared to the isotropic model and the SAM. Correlation coefficients obtained with the EAM are closer to the DNS ones, even though some discrepancies are visible particularly for the weak-level correlations. Nevertheless, the frequency and wavenumber–frequency wall pressure spectra computed with the KEEAM are satisfactory. The wall pressure model is able to take into account the effect of the Reynolds number even at higher Reynolds numbers than the one of the DNS data used to calibrate the EAM. The frequency spectrum is in good agreement with Goody (2004) model and the convective peak of the wavenumber–frequency spectrum is close to empirical models such as that of Chase (1987). The KEEAM is also able to predict the subconvective plateau. Besides, the TT and TMS contributions to the wall pressure spectrum are detailed. The TT terms are lower than the TMS one except at low frequencies, low wavenumbers and in the viscous regions where they are greater.

The interest of the model presented in this work, compared to empirical models, is that it can be applied to more complex flows since RANS solutions are used as input data. Local information of the flow is then taken into account in the computation of the wall pressure spectrum. The influence of adverse pressure gradients on the wavenumber–frequency spectrum of the wall pressure for an attached turbulent boundary layer flow and the ability of the KEEAM to predict it are the subject of an ongoing study.

Figure 23. Wavenumber–frequency spectra of the wall pressure fluctuations at computed with the KEEAM $Re_\theta \approx 4860$ and $\omega \delta / U_\infty \approx 15$. ——, total; grey ---, TT; black ----, TMS.
Acknowledgements

The authors would like to thank J. A. Sillero for providing the DNS results for the velocity correlation coefficients and L. Margheri for fruitful discussions about Kriging methods.

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