A GLOBAL ANALYSIS OF THE ATMOSPHERIC POLLUTANT MODELING

UMBERTO RIZZA¹, JONAS C. CARVALHO², DAVIDSON M. MOREIRA³, MARCELO R. MORAES⁴ & ANTÔNIO G. GOULART⁵

¹ CNR-ISAC Institute of Atmospheric Sciences and Climate, Itália, E-mail: u.rizza@isac.cnr.it.
², ³ Universidade Luterana do Brasil, Eng. Ambiental, PPGEAM, Canoas-RS, Brasil, E-mail: jonascc@yahoo.com.br; davidson@mecanica.ufrgs.br.
⁴ EPAGRI, Centro de Informações de Recursos Ambientais e de Hidrometeorologia, Florianópolis-SC, Brasil.
⁵ Universidade Federal do Pampa, Alegrete-RS, Brasil, E-mail: agoulart@mail.ufsm.br.

Recebido Maio 2004 - Aceito Junho 2005

ABSTRACT
In this article is carried out a comparison between lagrangian and eulerian modelling of the turbulent transport of pollutants within the Planetary Boundary layer (PBL). The Lagrangian model is based on a three-dimensional form of the langevin equation for the random velocity. The Eulerian analytical model is based on a discretization of the PBL in N sub-layers; in each of the sub-layers the advection-diffusion equation is solved by the Laplace transform technique. In the Eulerian numerical model the advective terms are solved using the cubic spline method while a Crank-Nicholson scheme is used for the diffusive terms. The models use a turbulence parameterization that considers a spectrum model, which is given by a linear superposition of the buoyancy and mechanical effects. Observed ground-level concentrations measured in a dispersion field experiment are used to evaluate the simulations.

Keywords: atmospheric pollutant modelling, turbulence parameterization, model evaluation.

RESUMO: UMA ANÁLISE GLOBAL DA MODELAGEM DA POLUIÇÃO ATMOSFÉRICA.

Palavras-chave: modelagem da poluição atmosférica, parametrização da turbulência, avaliação de modelos.

1. INTRODUCTION

The pollutant dispersion is usually investigated by two main approaches: Eulerian and Lagrangian. While the Eulerian reference system is fixed in relation to earth the Lagrangian system follows the atmospheric movement. Each one of these models presents advantages and disadvantages, which have been intensively investigated along last twenty years by the scientific community.
Laplace transform technique, considering an average value for eddy diffusivity and wind speed. In the Eulerian numerical model the time dependent Eulerian equation is splitted in a set of one-dimensional time-dependent equations, then the advective terms are solved using the cubic spline method and a Crank-Nicholson implicit scheme is used for the diffusive terms.

A fundamental issue in all kind of dispersion modelling is the parameterizations of PBL turbulence. This is very important as they define the dispersion properties, that is how pollutant is transported inside the PBL. The main question is thus to relate the pollutant spreading with the spectral characteristic of the PBL turbulence when it is generated by the two forcing mechanisms: buoyant and mechanical. In this work we utilize a formulation for turbulence spectra functions, which take into account these concepts.

In this study, we perform a comparison between models first, then a comparison with measured crosswind-integrated concentrations obtained from the well-known Copenhagen field experiment. This is used to compare observed and predicted concentrations. The results are evaluated through a statistical analysis suggested by Hanna (1989).

2. DESCRIPTION OF THE MODELS

2.1 Eulerian Numerical Model

A typical problem in air pollution studies is to seek the solution for the cross-wind (y direction) integrated concentration of pollutant emitted by a continuous source (being lateral concentration distribution usually assumed Gaussian), that is:

$$\frac{\partial \overline{C}}{\partial t} + U \frac{\partial \overline{C}}{\partial x} + W \frac{\partial \overline{C}}{\partial z} = \frac{\partial}{\partial z} \left( K_x \frac{\partial \overline{C}}{\partial z} \right) + \frac{\partial}{\partial z} \left( K_z \frac{\partial \overline{C}}{\partial z} \right) + S_q$$

(1)

where \(S_q = Q \delta(x) \delta(z - H_s)\) is the source term, \(\delta\) is the Dirac delta function, \(Q\) is the emission rate, \(H_s\) is source height, \(U, V, W\) are mean wind velocity components and \(K_x, K_y, K_z\) are the eddy diffusivities.

$$\overline{C} = \int_{-\infty}^{+\infty} C(x, y, z, t) dy = \overline{C}(x, z, t)$$

(2)

is the cross-wind integrated concentration.

A further simplification is obtained by considering longitudinal diffusion much smaller than longitudinal advection that is: \((\partial/\partial x)K_x \left( \partial \overline{C}/\partial x \right) \ll U \left( \partial \overline{C}/\partial z \right)\) and \(W \left( \partial \overline{C}/\partial z \right) \equiv 0\) so the Eq. (1) can be written as:

$$\frac{\partial \overline{C}}{\partial t} + U \frac{\partial \overline{C}}{\partial x} = \frac{\partial}{\partial z} \left( K_x \frac{\partial \overline{C}}{\partial z} \right) + S_q$$

(3)

The coefficients \(U\) and \(K_x\) of Eq. (3), are functions of the different parameters characterising the turbulent regimes of the PBL.

We will derive the full 2-D numerical scheme for the eq.1 and then the equivalent scheme for eq.3. This is not trivial because of numerical noises generates non-physical results. The difficulties stem from the radically different character of the advection and the turbulent diffusion operators. Even though Eq. (1) is formally parabolic in most PBL flows, transport is dominated by advection, leading to hyperbolic like characteristic.

The 2-D numerical algorithm consists in splitting Eq. (1) into a set of Locally One-Dimensional (LOD) time dependent equations (Yanenko, 1971; Rizza et al., 2003):

$$\frac{\partial \overline{C}}{\partial t} = \Lambda_x \overline{C} + \Lambda_z \overline{C}$$

(4)

where

$$\Lambda_x = A_x + D_x = -U \frac{\partial}{\partial x} + \frac{\partial}{\partial z} \left( K_x \frac{\partial}{\partial z} \right)$$

(5a)

$$\Lambda_z = A_z + D_z = -W \frac{\partial}{\partial z} + \frac{\partial}{\partial z} \left( K_z \frac{\partial}{\partial z} \right)$$

(5b)

This allows to reduce the magnitude of computational task and to transform the multidimensional problem into a sequence of one-dimensional equations. In this context, the numerical code is easier to develop and each single operator may be or not switched off. The matrices arising from the one-dimensional spatial discretization are usually tridiagonal, so the cost of using stable implicit procedures is small.

Using Crank-Nicholson time integration we have

$$\overline{C}^{n+1} = \left[ I - \frac{\Delta t}{2} \Lambda_x \right]^{-1} \left[ I + \frac{\Delta t}{2} \Lambda_x \right] \overline{C}^n$$

(6)

or equivalently

$$\overline{C}^{n+1} = T_x^* T_z^* \overline{C}^n$$

(7)

where \(I\) is the unity matrix. To obtain second order accuracy, it is necessary to reverse the order of the operators at each alternate step to cancel the two non-commuting terms. Therefore, it is possible to replace the scheme (7) with the following double-sequence equations

$$\overline{C}^n = T_x^* T_z^* \overline{C}^{n-1}$$

(8a)

$$\overline{C}^{n+1} = T_x^* T_z^* \overline{C}^n$$

(8b)

In order to develop a scheme that preserves peaks, retains positive quantities, and does not severely diffuse sharp gradients, a filtering procedure must be applied after each advective step. This is necessary in order to damping out the small-scale perturbations before they completely corrupt the basic solution. The general 2-D numerical scheme may be written as following:
\[ \bar{C}^0 = (A, fD_x) (A, fD_z) \bar{C}^{0-1} \]  
(9a)

\[ \bar{C}^{n+1} = (A, fD_x) (A, fD_z) \bar{C}^n \]  
(9b)

where the operator \( f \) represents the filter operation described by Forester et al. (1979). In our case, as a consequence of hypothesis leading to Eq. (4), the effective 2-D scheme is:

\[ \bar{C}^0 = (A, fD_x) (A, fD_z) \bar{C}^{0-1} \]  
(10a)

\[ \bar{C}^{n+1} = (D_x) (A, f) \bar{C}^n \]  
(10b)

The advective term (operator \( A_x \)), which is usually plagued by numerical noises is here solved using a method based on cubic spline interpolations, while a Crank-Nicholson implicit scheme is used for the diffusive term \( D_x \) (Rizza et al., 2003).

### 2.2. Eulerian Analytical Model

The mathematical description of the dispersion problem represented by the Eq. (3) is well defined when it is provided by initial and boundary conditions. It is indeed assumed that at the beginning of the contaminant release the dispersion region is not polluted, this means:

\[ \bar{C}(x, z, 0) = 0 \quad \text{at} \quad t = 0 \]  
(11)

The boundary conditions are zero flux at ground and PBL top:

\[ K_s \frac{\partial \bar{C}}{\partial z} = 0 \quad \text{at} \quad z = 0, h \]  
(12)

where \( h \) is the PBL height.

Bearing in mind the dependence of the \( K_s \) coefficient and wind speed profile \( U \) on variable \( z \), the height \( z \) of a PBL is discretized in \( N \) sub-intervals in such a manner that inside each interval \( K_s \) and \( U \) assume an average value. Therefore the solution of Eq. (3) is reduced to the solution of \( N \) problems of the type:

\[ \frac{\partial \bar{C}_m}{\partial t} + U_m \frac{\partial \bar{C}_m}{\partial x} = K_m \frac{\partial^2 \bar{C}_m}{\partial z^2} \]  
\[ z_m \leq z \leq z_{m+1} \]  
(13)

for \( m = 1:N \).

The analytical solution is obtained by using the Laplace Transform method (Vilhena et al., 1998; Moreira et al., 2004). Indeed the solution can be readily written as:

\[ \bar{C}_m(s, z, p) = A_u e^{x U_s} + B_u e^{s U_s} + \frac{Q}{2R_m K_m} \left( e^{x U_s (z + H)} - e^{x U_s z} \right) \]  
(14)

where

\[ A_u = L \{ \bar{C}(x, z, t); x \rightarrow s, \quad t \rightarrow s \} \]  
and

\[ R_m = \sqrt{\left( \frac{p}{K_m} \right) + \left( s U_m / K_m \right)} \].

Now, given a closer look to the solution in equation (14), we promptly realize that \( 2N \) integration constants are present. Therefore, to determine these integration constants, we impose (2N-2) interface conditions, namely the continuity of concentration and flux concentration at the interface. These constants are expressed as:

\[ \bar{C}_m = \bar{C}_{m+1} \]  
\( m = 1, 2, \ldots (N-1) \)  
(15a)

\[ K_m \frac{\partial \bar{C}_m}{\partial z} = K_{m+1} \frac{\partial \bar{C}_{m+1}}{\partial z} \]  
\( m = 1, 2, \ldots (N-1) \)  
(15b)

Finally, applying the interface and boundary conditions we obtain a linear system for the integration constants. Henceforth the concentration is obtained inverting numerically the transformed concentration \( \bar{C} \) by Gaussian quadrature scheme:

\[ \bar{C}_m(x, z, t) = \sum_{i=1}^k a_i \left( \frac{p_{i, t}}{t} \right) \sum_{j=1}^v a_j \left( \frac{p_{j, x}}{x} \right) A_m e^{- \left( \frac{p_{i, t}}{t} + \frac{p_{j, x}}{x} \right) K_m} \]  
(16)

where \( H \) \((z - H)\) is the Heaviside function. The solution is only valid for \( x > 0 \) and \( t > 0 \), as the quadrature scheme of Laplace inversion does not work for \( x = 0 \) and \( t = 0 \). The values of \( a_i, a_j \) (weights) and \( p_{i, t}, p_{j, x} \) (roots) of the Gaussian quadrature scheme are tabulated in the book by Stroud and Secrest (1966) and \( k \) and \( v \) are the quadrature points. In this work was utilised \( k, v = 8 \), because these values provides the required accuracy with small computational effort (Moreira et al., 2004).

### 2.3. Lagrangian Particle Model

Lagrangian stochastic particle model are based on a three-dimensional form of the Langevin equation for the random velocity (Thomson, 1987). The velocity and the displacement of each particle are given by the following equations (Rodene, 1996):

\[ d \mathbf{u}_i = a_i (\mathbf{x}, \mathbf{u}, t) dt + b_i (\mathbf{x}, \mathbf{u}, t) dW_i (t) \]  
(17)

and

\[ d \mathbf{x} = (U + \mathbf{u}) dt \]  
(18)

where \( i, j = 1, 2, 3 \) is the displacement vector in the directions \((x, y, z)\), \( U \) is the mean wind velocity vector \((U, V, W)\) in each direction, \( \mathbf{u} \) is the Lagrangian velocity vector in each direction \((u, v, w)\), \( a_i (\mathbf{x}, \mathbf{u}, t) dt \) is a deterministic term and \( b_i (\mathbf{x}, \mathbf{u}, t) dW_i (t) \) is a stochastic term and the quantity \( dW_i (t) \) is the incremental Wiener process. The Wiener process is a continuous but not
differentiable time integral of the “white noise”, $\xi(t)$. $\xi(t)$ is a hypothetical stationary, Gaussian, stochastic process with constant spectral density on the real frequency axis.

Thomson (1987) considered the Fokker-Planck equation as Eulerian complement of the Langevin equation to obtain the deterministic coefficient $a_i(x, u, t)$:

$$a_i P_e = \frac{\partial}{\partial x_i} \left( \frac{1}{2} b_{ij} b_{ik} P_e \right) + \phi_i(x, u, t) \quad (19a)$$

and

$$\frac{\partial \phi_i}{\partial t_i} = -\frac{\partial P_e}{\partial t} - \frac{\partial}{\partial x_i} \left( u P_e \right) \quad (19a)$$

subject to the condition

$$\phi \to 0 \text{ when } u \to \infty. \quad (20)$$

where $P_e(x, u, t)$ is the non-conditional PDF of the Eulerian velocity fluctuations. While in the two horizontal directions the $P_e$ is considered to be Gaussian, in the vertical direction the PDF is assumed to be non-Gaussian (to deal with non-uniform turbulent conditions and/or convection). For the vertical direction the Gram-Charlier PDF has been chosen. Gram-Charlier PDF is a particular type of expansion that uses orthonormal functions in the form of Hermit polynomials (Kendall and Stuart, 1977; Anfossi et al., 1997; Ferrero and Anfossi, 1998).

Comparing the Lagrangian velocity structure functions obtained from Langevin equation with that determined according to Kolmogorov’s theory of local isotropy in the inertial subrange, Thomson (1987) determined $b_{ij} = b_i \sqrt{C_0 \varepsilon}$, where $C_0$ is a Kolmogorov constant and $\varepsilon$ is the rate of turbulence kinetic energy dissipation. The product $(C_0 \varepsilon)^2$ can also be written as a function of the turbulent velocity variance $\sigma_i^2$; and the Lagrangian decorrelation time scale $\tau_{\ell_i}$ (Hinze, 1975):

$$C_0 \varepsilon = 2 \frac{\sigma_i^2}{\tau_{\ell_i}}. \quad (21)$$

The discretization of the Eqs. (17) and (18) is necessary for their practical application. The present model uses an explicit Euler scheme for velocities and an implicit scheme for displacement (Flash and Wilson, 1995). The concentration field is determined by counting the particles in a cell or imaginary volume in the position $x, y, z$.

3. TURBULENCE PARAMETERIZATION

The Eulerian models [Eqs. (10a and 10b) and (16)] and the Lagrangian model [Eqs. (17 and 18)] depend on turbulent parameters like the eddy diffusivities, Lagrangian decorrelation time scales and wind velocity variances. In this section we present the derivation of these parameters using a model for the turbulence spectra. These are modelled by a superposition of a buoyancy-produced part and a shear-produced part, neglecting the interaction between them through the localness hypothesis (Hinze, 1975, p. 232; Højstrup, 1982; Frisch, 1995, p. 105). The linear superposition of the two mechanisms occurs only when there is statistical independence between their Fourier components; this happens when the energy-containing wavenumber ranges of the two spectra are well separated (Mangia et al., 2000).

On the basis of Taylor’s theory, Batchelor (1949) proposed the following time-dependent relationships for the eddy diffusivities:

$$K_n = \frac{1}{2} \frac{d\sigma_i^2}{dt} = \frac{\sigma_i^2 \beta_i}{2\pi} \int_0^\infty F_i^e(n) \frac{\sin(n \pi \beta_i / \beta)}{n} dn, \quad (22)$$

where $\alpha = x, y, z$, $i = u, v, w$, $n$ is the frequency, $F_i^e(n)$ is the Eulerian spectrum normalized by the Eulerian velocity variance $\sigma_i^2$, $\sin^2(n \pi \beta_i / \beta)$ is a low-pass filter function that accounts for the travel time of the plume and $\beta_i$ is the ratio of the Lagrangian to the Eulerian integral timescales of the turbulence field. According to Wandel and Kofoed-Hansen (1962), $\beta_i$ can be written as

$$\beta_i = \left( \frac{\pi \bar{U}_i}{16 \bar{c_i}^2} \right)^{1/2} \quad (23)$$

where $\bar{U}$ is the mean wind.

For large diffusion times, the low-pass filter function in Eq. (22) selects the characteristic frequency ($n \to 0$) describing the energy-containing eddies. In this case, $F_i^e(n) = F_i^e(0)$ so that the eddy diffusivity becomes independent of the travel time from the source and can be expressed as a function of the local properties of turbulence:

$$K_n = \frac{\sigma_i^2 \beta_i F_i^e(0)}{4}. \quad (24)$$

From the Eq. (24) we can also determine the Lagrangian decorrelation time scale, given by:

$$T_{\ell_i} = \frac{K_n}{\sigma_i^2} = \frac{\beta_i F_i^e(0)}{4}. \quad (25)$$

Assuming the hypothesis of linear superposition of the buoyancy and mechanical process, we can model the dimensional Eulerian spectra as:

$$S_i^b(n) = S_i^b(n) + S_i^b(n), \quad (26)$$

where the subscripts $b$ and $s$ indicate buoyancy and shear production terms, respectively.

The adimensional Eulerian spectra is obtained normalising the dimensional Eulerian spectra with its total variance ($\sigma_i^2 = \sigma_i^2 + \sigma_i^2$):

$$F_i^e(n) = \frac{S_i^b(n) + S_i^b(n)}{\sigma_i^2} = \frac{S_i^b(n) + S_i^b(n)}{\sigma_i^2 + \sigma_i^2}. \quad (27)$$
It is important to point out that when dealing with both components (buoyancy and shear) sometimes it is easier to normalise each spectral component with own variance, as suggested by Degrazia et al. (2000) (Eq. 12, page 3577), but this is not possible as in this case it would be 
\[
\int_0^\infty [(S_b^E/\sigma_b^2) + (S_m^E/\sigma_m^2)]dn = 2.
\]

According to Olesen et al. (1984), \( S_b^E(n) \) can be given by 
\[
\frac{nS_b^E(n)}{w_*^2} = 0.98 \cdot c_4 \cdot f_n^{2/3} \cdot \left( \frac{z}{h} \right)^{2/3} \cdot \left( 1 + 1.5 \cdot \frac{f}{f_n^{1/3}} \right)^{5/3},
\]
where \( w_* \) is the convective velocity scale, \( \Psi_{eb} = \epsilon_b h \) is the nondimensional molecular dissipation rate function associated to buoyancy production, \( \epsilon_b \) is the buoyant rate of TKE dissipation given by Højstrup (1982), \( h \) is the convective PBL height, \( f = nz/U \) is the reduced frequency, \( f_n = z/(\lambda_m h) \) is the reduced frequency of the convective spectral peak, \( \lambda_m \) is the peak wavelength of the turbulent velocity spectra obtained according to Kaimal et al. (1976) and \( c_1 = \alpha_i \alpha_b (2\pi\kappa)^{-2/3} \) with \( \alpha_i = 0.5 \pm 0.05 \) and \( \alpha_b = 1.4/3, 4/3 \) for \( u \), \( v \) and \( w \) components, respectively.

Following Degrazia and Moraes (1982), \( S_m^E(n) \) can be written as 
\[
\frac{nS_m^E(n)}{u_*^2} = 1.5 \cdot c_1 \cdot f_n^{2/3} \cdot \left( \frac{z}{h} \right)^{2/3} \cdot \left( 1 + 1.5 \cdot \frac{f}{f_n^{1/3}} \right)^{5/3},
\]
where \( u_* \) is the local friction velocity, \( \phi_{ms} = (\epsilon_b/kz)/u_*^2 \) is the dissipation rate function associated to mechanical production, \( \epsilon_m \) is the mechanical rate of TKE dissipation given by Højstrup (1982), \( \kappa \) is the Von Karman constant and \( f_n = z/(\lambda_m h) \) is the reduced frequency of the neutral spectral peak obtained according to Olesen et al. (1984). The values of \( f_n \), \( \lambda_m \), \( \epsilon_b \) and \( \epsilon_m \) are found in the appendix.

By considering Eqs. (24) and (25) and assuming the hypothesis of linear superposition given by Eq. (26), the expressions for the eddy diffusivities and Lagrangian decorrelation time scales for large travel times (diffusion regime) can be obtained as:
\[
K_* = \frac{\beta_s}{4} \left[ S_b^E(0) + S_m^E(0) \right],
\]
and 
\[
T_* = \frac{\beta_s}{4} \left[ \frac{S_b^E(0) + S_m^E(0)}{\sigma_*^2} \right].
\]

Taking the limit of the Eqs. (28) and (29) when \( n \to 0 \), we can obtain the expressions for the dimensional spectra near the origin \( S_b^E(0) \) and \( S_m^E(0) \):
\[
S_b^E(0) = \lim_{n \to 0} nS_b^E(n) = \frac{0.98}{(f_n^{3/2})^3} \left( \frac{z}{h} \right)^{2/3} \cdot \frac{1}{\left( \frac{f_n^{1/3}}{f_n^{2/3}} \right)^{5/3}}.
\]
and 
\[
S_m^E(0) = \lim_{n \to 0} nS_m^E(n) = \frac{1.5}{(f_n^{3/2})^3} \left( \frac{z}{h} \right)^{2/3} \cdot \frac{1}{\left( \frac{f_n^{1/3}}{f_n^{2/3}} \right)^{5/3}}.
\]

By analytically integrating the Eulerian spectra given by Eqs. (28) and (29) over whole frequency domain, we can obtain the buoyant and mechanical wind velocity variance components:
\[
\sigma_b^2 = \int_0^\infty nS_b^E(n)dn = \frac{1.06}{(f_n^{3/2})^3} \cdot \frac{1}{\left( \frac{f_n^{1/3}}{f_n^{2/3}} \right)^{5/3}}.
\]
and 
\[
\sigma_m^2 = \int_0^\infty nS_m^E(n)dn = \frac{2.32}{(f_n^{3/2})^3} \cdot \frac{1}{\left( \frac{f_n^{1/3}}{f_n^{2/3}} \right)^{5/3}}.
\]

From which we get the total variance \( \sigma^2 = \sigma_b^2 + \sigma_m^2 \).

In Figure 1 a-b it is depicted the vertical profiles of eddy diffusivities coefficients and Lagrangian decorrelation time scales. Figure 1a is obtained by inserting Eqs. (32a) and (32b) into Eq. (30). Figure 1b is obtained by using again Eqs. (32a) and (32b) and total variance into Eq. (31). This figure shows a typical PBL profile for both quantities.

4. RESULTS

In this section we report numerical simulations and comparisons with measured data. In a first step a comparison between the models is done considering two source configurations, as it is usually done in air pollution context that is for low and tall sources. As second step we compare the models with Copenhagen data set. We consider this test particularly suited for this validation, since the tracer experiments were performed in Copenhagen area under neutral to convective conditions. The profiles of eddy-diffusivities coefficients and Lagrangian decorrelation time scale were calculated according the Eqs. (30) and (31), respectively. Wind speed profile has been parameterized following the classical logarithmic profile (Berkowicz et al., 1986).

4.1. Comparisons Between the Models

Figure 2 a-b shows the longitudinal profiles of \( C_y \) as predicted by the three models for an elevated (115 m) and low source (30 m), respectively. If we consider the location and the value of maximum \( C_y \) we can see that the three models are
in very good agreement in both cases. These results are very important in sense that the maximum concentration is one of the most significant parameter in air quality assessment.

4.2. Comparisons with Copenhagen Experiment

The performance of the models has been evaluated against experimental ground-level concentration measured in Copenhagen (Gryning and Lyck, 1984) diffusion experiment. Copenhagen experiment was carried out in the northern part of Copenhagen. The pollutant (SF6) was released without buoyancy from a tower at a height of 115 m and collected at the ground-level positions in up to three crosswind arcs of tracer sampling units. The sampling units were positioned 2-6 km from the point of release. The site was mainly residential with a roughness length of 0.6 m. All the available data were used to create the input for the simulations.

The models performance is shown in Table 1 and Figure 3. Table 1 shows the result of the statistical analysis made with observed and predicted values of ground-level cross-wind-integrated concentration ($C_y$). Figure 3 shows the scatter diagram between observed and predicted ground-level cross-wind integrated concentrations. The statistical indices are suggested by Hanna (1989):

![Figure 1](image1.png)  
**Figure 1** - Nondimensional eddy diffusivities (a) and Lagrangian decorrelation time scales (b) according to Eqs. (31) and (32), respectively, for unstable condition.

![Figure 2](image2.png)  
**Figure 2** - Comparison between longitudinal profiles of ground-level cross-wind integrated concentrations ($C_y$) as predicted by the Eulerian numerical, Eulerian analytical and Lagrangian models for an elevated source (115 m) (a) and low source (30 m) (b).
NMSE = \frac{(C_o - C_p)^2}{\bar{C_o}\bar{C_p}} 
(Normalized Mean Square Error)

FB = \frac{\bar{C_o} - \bar{C_p}}{0.5(\bar{C_o} + \bar{C_p})} 
(Fractional Bias)

FS = 2\left(\frac{\sigma_o - \sigma_p}{\sigma_o + \sigma_p}\right) 
(Fractional Standard Deviation)

RR = \frac{(C_o - C_p)(C_p - C_o)}{\sigma_o \sigma_p} 
(Correlation Coefficient)

FA2 = 0.5 \leq \frac{C_o}{C_p} \leq 2 
(Factor 2)

where C is the analyzed quantity (concentration) and the subscripts “o” and “p” represent the observed and the predicted values, respectively. The overbars in the statistical indices indicate averages. The statistical index FB indicates if the predicted quantity underestimates or overestimates the observed one. The statistical index NMSE represents the quadratic error of the predicted quantity in relation to the observed one. The statistical index FS indicates the measure of the comparison between predicted and observed plume spreading. The statistical index FA2 provides the fraction of data for which 0.5 \leq C_o/C_p \leq 2. As nearest zero are the NMSE, FB and FS and as nearest one are the RR and FA2, better are the results.

Analysing the statistical indices in Tables 1 it is possible to notice that the model simulates quite well the observed concentrations, with NMSE, FB and FS values relatively near to zero and RR and FA2 very close to 1. Fractional bias (FB in Table 1) shows under-prediction for Lagrangian model and over-prediction for Eulerian models. This also confirmed by visual inspection of Figure 3. For the other statistical indices, there are not considerable differences between the results. All the values for the indices are within ranges that are characteristics of those found for other state-of-the-art models applied to other field datasets, thus showing that the models and the turbulence parameterizations are quite effective.

5. CONCLUSIONS

We utilized both Eulerian and Lagrangian approaches to air pollution problems to make a comparison between these different techniques. Furthermore the Eulerian conservation equation for a passive contaminant has been solved both numerically than analytically. A fundamental recipe in all kind of modelling is the parameterization of turbulent quantities describing the dispersion properties of PBL. We proposed a new parameterization for eddy diffusivity and Lagrangian decorrelation time scales that properly model both generation mechanisms of PBL turbulence. A sensitivity analysis has been conducted between the three models first and then with experimental data. Such comparison show an excellent agreement between the three models showing that they can be incorporate in a global modelling system for air-quality estimates. This work is just preliminary. A more detailed comparison will be made using more concentration datasets in different PBL stability conditions.

6. APPENDIX

Physical Constraints

The convective velocity scale \( w_c \) is given by:

\[ w_c = u_c \left( \frac{h}{\kappa L} \right)^{1/3}. \]

The values of the reduced frequency of the neutral spectral peak are (Table 1 of Olesen, 1984):

\( f_m^h = 0.045 \quad f_m^v = 0.16 \quad f_m^w = 0.35. \)

The spectral peaks depending on height and stability are determined according Caughey and Palmer (1979):
\((\lambda_m)_u = (\lambda_m)_v = 1.5h\)

and

\[
\left( \frac{z}{(0.55 - 0.3Lz/L)} \right), \quad 0 \leq z \leq L
\]

\[
(\lambda_m)_w = 5.9z, \quad |L| \leq z \leq 0.1h
\]

\[
1.8h \left[ 1 \exp(-4z/h) \right] 0.0003\exp(-8z/h), \quad 0.1h \leq z \leq h
\]

The buoyancy-shear ensemble average rates of dissipation of TKE are from Højstrup (1982):

\[
\varepsilon_b = (0.75)^{92} \frac{w^3}{h}, \quad \varepsilon_s = \frac{u^3}{kz} \left( 1 - \frac{z}{h} \right)^3.
\]

7. ACKNOWLEDGEMENTS

The authors acknowledge the financial support provided by CNPq (Conselho Nacional de Desenvolvimento Científico e Tecnológico) and FAPERGS (Fundação de Amparo à Pesquisa do Estado do Rio Grande do Sul).

8. REFERENCES


