Air Pollution Steady-State Advection-Diffusion Equation: The General Three-Dimensional Solution

Daniela Buske1, Marco Túllio Vilhena2, Tiziano Tirabassi3, Bardo Bodmann2

1Department of Mathematics and Statistics (IFM/DME), Federal University of Pelotas (UFPel), Pelotas, Brazil; 2Graduate Program in Mechanical Engineering, Federal University of Rio Grande do Sul (UFRGS), Porto Alegre, Brazil; 3Institute of Atmospheric Sciences and Climate (ISAC), National Research Council (CNR), Bologna, Italy.

Email: t.tirabassi@isac.cnr.it

Received June 15th, 2012; revised July 14th, 2012; accepted August 17th, 2012

ABSTRACT

Atmospheric air pollution turbulent fluxes can be assumed to be proportional to the mean concentration gradient. This assumption, along with the equation of continuity, leads to the advection-diffusion equation. Many models simulating air pollution dispersion are based upon the solution (numerical or analytical) of the advection-diffusion equation assuming turbulence parameterization for realistic physical scenarios. We present the general steady three-dimensional solution of the advection-diffusion equation considering a vertically inhomogeneous atmospheric boundary layer for arbitrary vertical profiles of wind and eddy-diffusion coefficients. Numerical results and comparison with experimental data are shown.

Keywords: Advection-Diffusion Equation; Analytical Solution; Laplace Transform; Air Pollution Modeling; Atmospheric Boundary Layer; Eulerian Models

1. Introduction

The processes governing the transport and diffusion of pollutants are numerous, and of such complexity that it would be impossible to describe them without the use of mathematical models. Such models therefore constitute an indispensable technical instrument of air quality management.

The theoretical approach to the problem essentially assumes two basic forms. In the Eulerian approach, diffusion is considered, at a fixed point in space, proportional to the local gradient of the concentration of the diffused material and is based on the resolution, on a fixed spatial-temporal grid, of the equation of the mass conservation of the pollutant chemical species. Lagrangian models are the second approach and they differ from Eulerian ones in adopting a system of reference that follows atmospheric motions. Initially, the term Lagrangian was used only to refer to moving box models that followed the mean wind trajectory [1]. Currently, this class includes all models that decompose the pollutant cloud into discrete “elements”, such as segments, puffs or computer virtual particles [1]. In Lagrangian particle models pollutant dispersion is simulated through the motion of computer virtual particles whose trajectories allow the calculation of the concentration field of the emitted substance. The underlying hypothesis is that the combination of the trajectories of such particles simulates the paths of the air particles situated, at the initial moment, in the same position. The motion of the particles can be reproduced both in a deterministic way and in a stochastic way.

In this paper we limited ourselves to the Eulerian approach, and in particular to the K model, where the flow of a given field is assumed to be proportional to the gradient of an appropriate mean variable [2]. K-theory has its own limits, but its simplicity has led to a widespread use as the mathematical basis for simulating pollution dispersion. Most of Eulerian models are based on the numerical solution of the equation of mass conservation of the pollutant chemical species. Such models are most suitable to confronting complex problems, for example, the dispersion of pollutants over complex terrain or the diffusion of non-inert pollutants.

However, a progressive and continuous effort to obtain analytical solutions of the advection-diffusion equation (ADE) has been made in the last years. In fact analytical solutions of equations are of fundamental importance in understanding and describing physical phenomena. Analytical solutions explicitly take into account all the parameters of a problem, so that their influence can be reliably investigated and it is easy to obtain the asymptotic behavior of the solution, which is usually very much
more tedious to generate through numerical calculations. Moreover, like the Gaussian solution, that was the first solution of ADE with the wind and the eddy diffusivity coefficients supposed constant in space, they may suggest the construction of operative analytic model. Gaussian models, so named because they are based on the Gaussian solution, are forced to represent real situations by means of empirical parameters, referred to as “sigs”. Gaussian models are fast, simple, do not require complex meteorological input, and describe the diffusive transport in an Eulerian framework, making easy use of the Eulerian nature of measurements. For these reasons they are widely employed for regulatory applications by environmental agencies all over the world although their well known intrinsic limits.

A significant number of works regarding ADE analytical solution (mostly two-dimensional solutions) is available in the literature. References [3-16] are considered relevant by the authors. However, the above solutions are valid for very specialized situations: only ground level sources or infinite height of the Atmospheric Boundary Layer (ABL) or specific wind and eddy diffusivities vertical profiles. Vilhena et al. [17] presented an analytical solution, called ADMM (Advection Diffusion Multilayer Method) method, for a limited ABL and general wind and eddy diffusivities vertical profiles, but expressed by a stepwise function (see also [18-20]). Many of the above solutions were utilized in operative air pollution models [21].

Finally a general two-dimensional solution without any restriction in the spatial function of wind and eddy diffusion coefficients was presented in Wortmann et al. [22] and Moreira et al. [23,24]. The solving methodology was the Generalized Integral Laplace Transform Technique (GILTT) that is an analytical series solution including the solution of an associate Sturm-Liouville problem, expansion of the pollutant concentration in a series in terms of the attained eigenfunction, replacement of this expansion in the ADE and, finally, taking moments. This procedure leads to a set of differential ordinary equations that is solved analytically by Laplace transform technique. A complete review of the GILTT method is given in [25].

In this paper we extend these last results and we present a three-dimensional solution for a limited height ABL and without any restriction in the spatial function of wind and eddy diffusion coefficients. Cauchy-Kowalewski theorem [26] guarantees the existence and uniqueness of an analytical solution of the ADE, so we are presenting the general solution of three-dimensional ADE.

2. The Three-Dimensional Solution

The ADE for air pollution in the atmosphere is essentially a statement of conservation of the suspended material and in Cartesian geometry it can be written as [27]:

$$\begin{align*}
\nabla \overline{c} + \nabla \cdot \overline{D} + \nabla \cdot \rho = 0,
\end{align*}$$

Here $\overline{c}$ denotes the mean concentration of a passive contaminant (g/m³), $\nabla, \nabla$, and $\nabla$ are the Cartesian components of the mean wind (m/s) in the directions $x$ (0 < $x$ < $L_x$), $y$ (0 < $y$ < $L_y$) and $z$ (0 < $z$ < $H$) and $K_x, K_y, K_z$ are the eddy diffusivities (m²/s). Equation (1) is subjected to the following boundary and source conditions:

$$\begin{align*}
K \nabla \overline{c} |_{(0,0,0)} = 0,
\overline{w} (0, y, z) = Q \delta (y) \delta (z - H_s),
\end{align*}$$

where $K = \text{diag} (K_x, K_y, K_z)$ is the diagonal diffusion matrix, $Q$ is the emission rate (g/s), $h$ the height of the ABL (m), $H_s$ the height of the source (m), $L_x$ and $L_y$ are the limits in the $x$ and $y$-axis and far away from the source (m) and $\delta$ represents the Dirac delta function. The source position is at $x = 0, y = y_0$ and $z = H_s$.

In order to solve problem (1), taking advantage of the well-known solution of the two-dimensional problem with advection in the $x$-direction by the GILTT method [25], we initially apply the integral transform technique in the $y$ variable. To this end, we expand the pollutant concentration as:

$$\begin{align*}
\overline{c} (x, y, z) = \sum_{n=0}^{n} \overline{c}_m (x, z) Y_n (y),
\end{align*}$$

where $Y_n (y)$ are a set of orthogonal eigenfunctions, given by $Y_n (y) = \cos (\lambda_n y)$, and $\lambda_n = mn / L_y$ ($m = 0, 1, 2, \cdots$) are respectively the set of eigenvalues.

To determine the unknown coefficient $\overline{c}_m (x, z)$ we manipulate Equation (1) applying the chain rule for the diffusion terms. After substitute Equation (4) in the resulting equation and taking moments, meaning applying the operator $\int_0^{L_y} Y_n (y) dy$, we obtain the result:

$$\begin{align*}
\sum_{n=0}^{n} \left[ \frac{\partial}{\partial x} \left( K_x \overline{c}_m (x, z) \right) + \frac{\partial}{\partial z} \left( K_z \overline{c}_m (x, z) \right) + \rho \overline{c}_m (x, z) \right] Y_n (y) + \overline{w} (x, z) \int_0^{L_y} Y_n (y) Y_n (y) dy + \lambda_n^2 \overline{c}_m (x, z) \int_0^{L_y} Y_n (y) Y_n (y) dy + \int_0^{L_y} K_y Y_n (y) Y_n (y) dy,
\end{align*}$$

Air Pollution Steady-State Advection-Diffusion Equation: The General Three-Dimensional Solution 1125

Copyright © 2012 SciRes.
Defining the integrals appearing in the above equation like:

\[
\begin{align*}
\int_0^{L_y} Y_m(y) Y_n(y) \, dy &= \alpha_{mn}, \\
\int_0^{L_y} K^*_{mn} Y_n(y) \, dy &= \gamma_{nm}, \\
\int_0^{L_x} K_{mn} Y_n(y) \, dy &= \eta_{nm}
\end{align*}
\]

the Equation (5) is rewritten as:

\[
\sum_{m=0}^{\infty} \left[ \frac{\partial}{\partial x} \left( K_{mn} \frac{\partial \tilde{\varphi}_m(x,z)}{\partial x} \right) + \frac{\partial}{\partial z} \left( K_{mn} \frac{\partial \tilde{\varphi}_m(x,z)}{\partial z} \right) + \frac{\partial}{\partial x} \left( \lambda_{mn} \tilde{\varphi}_m(x,z) \right) = 0 \right] + \frac{\partial}{\partial x} \left( \lambda_{mn} \tilde{\varphi}_m(x,z) \right) = \alpha_{mn} - \tilde{\varphi}_{mn} \tilde{\varphi}_m(x,z)
\]

(6)

Without losing generality, we specialize the application for a pollutant dispersion problem in ABL, assuming that the speeds \( \bar{v} \) and \( \bar{w} \) takes the null value. We neglect the diffusion component \( K_r \) because we assume that the advection is dominant in the x-direction, i.e.,

\[
\bar{u} = \frac{\partial \bar{\varphi}}{\partial x} \Rightarrow \frac{\partial}{\partial x} \left( K_{mn} \frac{\partial \tilde{\varphi}_m(x,z)}{\partial x} \right)
\]

Further we also consider that \( K_r \) has only dependence on the z-direction. After these assumptions, Equation (6) is rewritten in matrix fashion as a set of \( M + 1 \) two-dimensional diffusion equations:

\[
\frac{\partial \tilde{\varphi}_m(x,z)}{\partial x} = \frac{\partial}{\partial x} \left( K_{mn} \frac{\partial \tilde{\varphi}_m(x,z)}{\partial x} \right) - \lambda_{mn} K_r \tilde{\varphi}_m(x,z)
\]

(6)

The problem (6) is then solved by the GILTT method. Following the work of Moreira et al. [25] and taking advantage of the well known solution for the stationary problem with advection in the x direction, we pose the solution of problem (6) in the form:

\[
\tilde{\varphi}_m(x,z) = \sum_{i=0}^{L_x} \tilde{\varphi}_{m,i}(x) \zeta_i(z)
\]

(7)

where \( \zeta_i(z) \) are a set of orthogonal eigenfunctions, given by \( \zeta_i(z) = \cos \left( \gamma_i z \right) \), and \( \gamma_i = \ln h / l \) (\( l = 0, 1, 2, \ldots \)) are the set of eigenvalues.

Replacing Equation (7) in Equation (6) and taking moments, we get the first order matrix differential equation:

\[
\frac{d}{dx} \left( P_m(x) + F P_m(x) \right) = 0
\]

(8)

for \( m = 0:M \), where \( P_m(x) \) is the column vector whose components are \( \tilde{\varphi}_{m,i} \) for \( i = 0:L \). The matrix \( F \) is defined as \( F = B_1^{-1} B_1 \). The entries of matrices \( B_1 \) and \( B_2 \) are, respectively, given by:

\[
\begin{align*}
(b_{11})_{ij} &= -\int_0^L u \zeta_i(z) \zeta_j(z) \, dz \\
(b_{12})_{ij} &= \int_0^L K^*_{ij} \zeta_i(z) \zeta_j(z) \, dz - \lambda_j \int_0^L K_{ij} \zeta_j(z) \zeta_j(z) \, dz \\
(b_{21})_{ij} &= \int_0^L K^*_{ij} \zeta_i(z) \zeta_j(z) \, dz \\
(b_{22})_{ij} &= \lambda_j \int_0^L K_{ij} \zeta_j(z) \zeta_j(z) \, dz
\end{align*}
\]

Similar procedure leads to the boundary condition of problem (9):

\[
P_m(0) = \tilde{\varphi}_{m,i}(0) = Q \zeta_j \left( H_i \right) Y(y_0) A^{-1},
\]

(9)

where \( A^{-1} \) is the inverse of matrix \( A \) having the entry \( a_{ij} = \int_0^L u \zeta_i(z) \zeta_j(z) \, dz \).

Problem (9) is solved applying Laplace transform and diagonalization. Firstly, transforming \( x \) in \( s \) and \( P \) em \( \bar{P} \) the equation becomes

\[
s\bar{P}(s) - P(0) + F \bar{P}(s) = 0,
\]

(10)

where the overbar represents the transformed potential.

The matrix \( F \) is decomposed in eigenvectors and eigenvalues as \( F = X D X^{-1} \) where \( X \) is the matrix of the eigenvectors and \( D \) is the diagonal matrix of the eigenvalues of \( F \). Then, the Equation (10) became

\[
\left( sI + X D X^{-1} \right) \bar{P}(s) = P(0),
\]

(11)

where \( I \) is the matrix identity. After algebraic manipulation we get

\[
\bar{P}(s) = X \left( sI + D \right)^{-1} \xi,
\]

(12)

where \( \xi \left( \xi = X^{-1} P(0) \right) \) is found from the equation \( X \xi = P(0) \), and their values are calculated by LU factorization. The elements of the matrix \( (sI + D) \) have the form \( \{s + d_i\} \), where \( d_i \) are the eigenvalues of the matrix \( F \) and the elements of \( (sI + D)^{-1} \) are \( 1/(s + d_i) \) whose inverse Laplace transform is \( e^{-d_i s} \). Let be \( G(x) \) the diagonal matrix whose elements are \( e^{-d_i s} \) the final solution is then given by

\[
P(x) = X G(x) \xi.
\]

(13)

Then, using Equation (7),

\[
\tilde{\varphi}_m(x,z) = \sum_{i=0}^{L_x} \tilde{\varphi}_{m,i}(x) \zeta_i(z)
\]

(7)

we obtain the solution of the 2D problem, where \( \xi(z) = \cos \left( \gamma_i z \right) \) and \( \tilde{\varphi}_{m,i}(x) \) comes from the solution of the transformed problem given by Equation (8). Once \( \tilde{\varphi}_m(x,z) \) is known we are in a position to write the final three-dimensional solution of problem (1) which is given by Equation (4),

\[
\tilde{\varphi}(x,y,z) = \sum_{m=0}^{\infty} \tilde{\varphi}_m(x,y) Y_m(y),
\]

(14)

where \( Y_m(y) = \cos \left( \lambda_m y \right) \). This solution is named as 3D-GILTT (three-dimensional GILTT solution).

Since this problem is a special case of the Cauchy-Kowalewsky theorem, existence and uniqueness is guaranteed. Convergence of the solution may be shown by a genuine mathematical convergence criterion. Note that the only numerical error comes from truncation, which is determined from the Sturm-Liouville problem. Recalling that the structure of the contaminant is essentially determined by advection and diffusion, present in form of a velocity vector \( U = (\bar{u}, \bar{v}, \bar{w}) \) and a diagonal diffusion
matrix $K = \text{diag}(K_x, K_y, K_z)$, which define a smoothness length scale by the maximum norm of the expression $\sigma = \|K\|/\|L\|$. Thus one may conclude that with decreasing length (either $m$ or $n$) and an increasing integer number) variations in the solution become spurious.

Upon interpreting $\sigma^{-1}$ as a sampling density, one may now employ the Cardinal Theorem of Interpolation Theory [28] in order to find the truncation

$$n = \text{int} \left[ m \int \pi_{\tau z;\gamma} / 2 \pi \sigma + 1 / 2 \right]$$

in Equation (4) that leaves the analytical solution almost exact, i.e. introduces only functions that vary significantly in length scales beyond the mentioned limit. Thus, the Cardinal Theorem of Interpolation theory may be cast in form of a convergence criterion as follows.

The square integrable function $\chi = \int \pi \, dx \, d\eta \in L^2$

($\eta = y$ or $z$) with spectrum $\{\lambda_i\}$ which is bounded by $m \sigma$ has an exact solution for a finite expansion. This statement expresses the Cardinal Theorem of Interpolation Theory for our problem (see Bodmann et al. [29]).

Since the cut-off, i.e. the aforementioned supreme defines some sort of a sampling density, its introduction is an approximation and is related to convergence of the approach and Parseval’s theorem may be used to estimate the error. In order to keep the solution error within a specified order of magnitude, the expansion in the domain of interest has to contain $n + 1$ terms. For the bounded spectrum and according to the theorem the solution is then exact. In our approximation, if $m$ is properly chosen such that the cut-off part of the spectrum is negligible, then the found solution is almost exact.

3. Model Application to Different Meteorological Scenarios and against Experimental Data

In order to illustrate the behavior of the discussed solution we report a simulation of contaminant dispersion in the ABL for different scenarios. Moreover, we evaluate the performance against experimental ground-level concentration. To do this we first have to introduce a boundary layer parameterization.

3.1. Atmospheric Boundary Layer Parameterization

Define abbreviations and acronyms the first time they are used in the text, even after they have been defined in the abstract. In air pollution diffusion modeling the choice of a turbulence parameterization represents a fundamental aspect for the contaminants dispersion modeling. From a physical point of view a turbulence parameterization is an approximation to nature in the sense that we are putting in mathematical models an approximated relation that in principle can be used as a surrogate for the natural true unknown term [30]. The reliability of each model strongly depends on the way as turbulent parameters are calculated and related to the current understanding of the ABL [31].

The literature reports many, greatly varied formulae, for the calculation of the vertical turbulent diffusion coefficient [2]. As an example of application of our new solution we tested the following vertical and lateral diffusion parameterization suggested by Degrazia et al. [32], derived from Batchelor [33], for convective conditions:

$$K_z = \frac{0.22 \left( \frac{z}{h} \right)^{1/3} \left( 1 - \frac{z}{h} \right)^{1/3} \left[ 1 - e^{-\frac{4z}{L}} - 0.0003 e^{\frac{4z}{L}} \right]}{w_h}$$

$$K_y = \frac{\sqrt{\pi} \sigma_v}{16 (f_m) \psi_e}$$

with $\sigma_v^2 = 0.98 c_v \left( \psi_e \right)^{2/3} \left( \frac{z}{h} \right)^{2/3} \frac{w^2}{f_m} \psi_e = 4.16 \frac{z}{h}$

$$f_m = 0.16 \quad \text{and} \quad \psi_e^{1/3} = \left[ 1 - \frac{z}{h} \left( \frac{z}{L} \right)^{-2/3} + 0.75 \right]^{1/2}.$$ 

More, $k$ is the von Karman constant ($k = 0.4$), $w_h$ is the convective velocity scale, $L$ is the Monin-Obukhov length, $\sigma_v$ is the Eulerian standard deviation of the longitudinal turbulent velocity, $q_e$ is the stability function, $\psi_e$ is the non-dimensional molecular dissipation rate function, $f_m$ is the peak wavelength of the turbulent velocity spectra and $c_v = a_v \alpha_e (2 \pi \kappa)^{-2/3}$ with $a_v = 0.5 \pm 0.05$ and $\alpha_e = 4/3$ [34,35].

Following Pleim and Chang [36] during convective conditions at $h/L \leq 10$ the following relation is used:

$$K_z = k \psi_e (1 - z/h)^2$$

$$K_y = k \psi_e (1 - z/h)^2 \psi_e$$

where $\psi_e = 1 + 5 (z/L)$ in stable conditions and $\psi_e = 1$ in neutral conditions.

Degrazia et al. [37] proposed for the stable boundary layer an algebraic formulation for the eddy diffusivities in the $y$-direction. It takes the form:

$$\frac{K_y}{w_h} = \frac{2 \sqrt{\pi} 0.64 a^2 (1 - z/h)^{6/5} (z/h)^2}{2 \sqrt{\pi} 0.64 (z/h) + 16 a (f_m) (1 - z/h)^{6/5} X'}$$

$$\times \left[ 2 \sqrt{\pi} 0.64 a^2 (z/h) + 8 a (f_m) (1 - z/h)^{6/5} X' \right]^{1/2}$$

where $f_m = \left( f_m \right)_{\infty} (1 + 3.7(z/L))$ is the frequency of the spectral peak, $f_m = 0.33$ is the frequency of the spectral peak in the neutral stratification [38], $\Lambda = L (1 - z/h)^{(1.5a_v - a)}$ ($\alpha_v = 1.5$; $\alpha_v = 1$ [39]) is the
local Monin-Obukhov length, \( a_v = (2.7c_v)^{1/2}/(f)x \), where \( c_v = 0.4 \), \( u_* \) is the friction velocity and \( X^* = xu_*/f \) represents the non-dimensional distance.

The wind speed profile can be described by a power law expressed as follows [40]:

\[
\frac{\bar{u}_v}{\bar{u}_1} = \left( \frac{z}{z_1} \right)^n
\]

(20)

where \( \bar{u}_v \) and \( \bar{u}_1 \) are the mean wind speeds horizontal to heights \( z \) and \( z_1 \) and \( n \) is an exponent that is related to the intensity of turbulence [41].

Thus, in this study we introduce the eddy diffusivities and the wind profile described above in the 3D-GILTT model (Equation (14)) to calculate the ground-level concentration of emissions released from an elevated continuous source point in an unstable/neutral ABL. For the application to different meteorological scenarios the eddy diffusivities (17)-(19) are used, while in the comparisons against experimental data eddy diffusivities (15) and (16) are used.

### 3.2. Application to Different Meteorological Scenarios

Given the complexity of the solution, it is useful to show the behavior of the solution in different scenarios. It is possible to see a graphical representation in Figure 1 of the ground level concentrations predicted by the three dimensional solution for different source heights in convective condition.

Also to show the influence of the atmospheric turbulence we present in Figure 2 the non-dimensional concentration in function of the non-dimensional distance from the source \((H_s = 0.1 \, \text{h})\) for five different meteorological scenarios (exponent of power wind profile \( \alpha \) and inverse of Monin-Obukhov length \((1/L)\) values for different meteorological scenarios are showed in Table 1).

In Figure 3, the influence of the source height is showed. In fact, the non-dimensional vertical concentrations at three distances, and four different source heights is presented in convective condition.

### 3.3. Application against Experimental Data

In order to show the performance of the present solution of the ADE and the performance of the proposed ABL parameterizations we have applied the model using the Copenhagen and Kinkaid experimental datasets. The first experiment is carried out in the northern part of Copenhagen, described by Gryning and Lyck [42]. It consisted of tracer released without buoyancy from a tower at a height of 115 m, and collection of tracer sampling units at the ground-level positions at the maximum of three crosswind arcs. The sampling units were positioned at

![Figure 1. Ground level concentrations predicted by the three dimensional solution for different source heights in convective conditions \((1/L = -0.01 \, \text{m}^{-1})\).](image)

![Figure 2. Non-dimensional concentration \((C^* = cuh^2/Q)\) in function of the non-dimensional distance \((X^* = xu_*/uh)\) from the source \((H_s = 0.1 \, \text{h})\) for 5 different meteorological scenarios.](image)

<table>
<thead>
<tr>
<th>Scenario</th>
<th>(\alpha)</th>
<th>(1/L , (\text{m}^{-1}))</th>
</tr>
</thead>
<tbody>
<tr>
<td>Unstable</td>
<td>0.07</td>
<td>-0.10</td>
</tr>
<tr>
<td>Unstable</td>
<td>0.1</td>
<td>-0.02</td>
</tr>
<tr>
<td>Neutral</td>
<td>0.15</td>
<td>0</td>
</tr>
<tr>
<td>Stable</td>
<td>0.35</td>
<td>0.01</td>
</tr>
<tr>
<td>Stable</td>
<td>0.55</td>
<td>0.03</td>
</tr>
</tbody>
</table>

Table 1. Exponent of power wind profile \((\alpha)\) and inverse of Monin-Obukhov length \((1/L)\) values for different meteorological scenarios.
two to six kilometers from the point of release. The site was mainly residential with a roughness length of 0.6 m.

The Kinkaid experiment was conducted at Illinois, USA, during convective conditions (for \(-h/L > 10\)) and is described in the work of Hanna and Paine [43]. The Kinkaid field campaign concerns an elevated release in a flat farmland with some lakes. During the experiment, SF\(_6\) was released from 187 tall stacks and recorded on a network consisting of roughly 200 samplers positioned in arcs from 0.5 to 50 km downwind of the source. The data set includes the meteorological parameters as friction velocity, Monin-Obukhov length and height of boundary layer. The measured concentration level is frequently irregular with high and low concentrations occurring intermittently along same arc, moreover there are frequent gaps in the monitoring arcs. For the above reasons a variable has been assigned as a quality factor in order to indicate the degree of readability of data [44]. The quality indicator (from 0 to 3) has been assigned. Here, only the data with quality factor 3 were considered.

Figures 4 and 5 show the comparison of 3D-GILLT predicted concentrations against observed data in the Copenhagen and Kinkaid experiments. We can observe that the obtained concentrations reproduce acceptably the observed data.

In the further we use standard statistical indices in order to compare the quality of the new approach against other models. While the present approach (3D-GILTT) is based on a genuine three dimensional description an earlier analytical approach called GILTTG uses a Gaussian assumption for the horizontal transverse direction [25]. In the GILTTG the crosswind integrated concentration \(c(x, z, t)\) (i.e., two-dimensional) is obtained analytically using the GILTT method. To calculate the three-dimensional concentration \(c(x, y, z, t)\) lateral diffusion (\(\sigma_y\))...
needs to be included, that is, it is assumed that the plume has a Gaussian concentration distribution in the lateral. So, to calculate the concentration the following expression is assumed:

\[ c(x, y, z, t) = c^*(x, z, t) \frac{e^{-y^2/(2\sigma_y^2)}}{\sqrt{2\pi\sigma_y}} \]

Note that we present the two analytical model approaches, since the earlier one was found to be acceptable in comparison to other approaches found in the literature and both give a solution in closed form. For the Copenhagen data set, our results are also compared with the one obtained with the GIADMT method [18]. Basically the GIADMT method consists on the solution of the GITT (Generalized Integral Transform Technique) transformed problem by the ADMM method. The ADMM approach solves the two-dimensional ADE with variable wind profile and eddy diffusivity coefficient [19]. The main idea here relies on the discretisation of the ABL in a multilayer domain, assuming in each layer that the eddy diffusivity and wind profile take averaged values. The resulting ADE in each layer is then solved by Laplace transform.

Tables 2 and 3 present some performances evaluations of the model results using the statistical evaluation procedure described by Hanna [45] and defined in the following way:

\[ \text{NMSE} = \left( \frac{C_o - C_p}{\sigma_o} \right)^2 / \left( \frac{C_p}{\sigma_p} \right), \]
\[ \text{FA2} = \text{data for which } 0.5 \leq \left( \frac{C_p}{C_o} \right) \leq 2, \]
\[ \text{COR} = \frac{(C_o - C_p)(C_p - C_o)}{0.5(C_o + C_p)}, \]
\[ \text{FB} = \frac{C_o - C_p}{0.5(C_o + C_p)}, \]
\[ \text{FS} = \frac{(\sigma_o - \sigma_p)}{0.5(\sigma_o + \sigma_p)}, \]

where NMSE is the normalized mean square error, COR the correlation coefficient, FA2 is the fraction of data (% normalized to 1), FB the fractional bias, FS the fractional standard deviations. Subscripts \( o \) and \( p \) refer to observed and predicted quantities, respectively, and the overbar indicates an averaged value. The statistical index FB says if the predicted quantities underestimate or overestimate the observed ones. The statistical index NMSE represents the model values dispersion in respect to data dispersion.

Table 2. Statistical comparison between models using the Copenhagen dataset.

<table>
<thead>
<tr>
<th>Model</th>
<th>NMSE</th>
<th>COR</th>
<th>FA2</th>
<th>FB</th>
<th>FS</th>
</tr>
</thead>
<tbody>
<tr>
<td>GILTG</td>
<td>0.33</td>
<td>0.80</td>
<td>0.87</td>
<td>0.28</td>
<td>0.09</td>
</tr>
<tr>
<td>GIADMT</td>
<td>0.15</td>
<td>0.87</td>
<td>0.96</td>
<td>0.01</td>
<td>-0.09</td>
</tr>
<tr>
<td>3D-GILT</td>
<td>0.07</td>
<td>0.93</td>
<td>0.96</td>
<td>0.02</td>
<td>0.03</td>
</tr>
</tbody>
</table>

Table 3. Statistical comparison between models using the Kinkaid dataset.

<table>
<thead>
<tr>
<th>Model</th>
<th>NMSE</th>
<th>COR</th>
<th>FA2</th>
<th>FB</th>
<th>FS</th>
</tr>
</thead>
<tbody>
<tr>
<td>GILTG</td>
<td>0.37</td>
<td>0.68</td>
<td>0.77</td>
<td>0.08</td>
<td>-0.15</td>
</tr>
<tr>
<td>3D-GILT</td>
<td>0.37</td>
<td>0.67</td>
<td>0.71</td>
<td>0.09</td>
<td>-0.09</td>
</tr>
</tbody>
</table>
The best results are expected to have values near to zero for the indices NMSE, FB and FS, and near to 1 in the indices COR and FA2. The statistical indices point out that a reasonable agreement is obtained between experimental data and the 3D-GILTT model for both cases.

In order to validate the two models (GILTTG and 3D-GILTT) we fit the predicted versus observed values by a linear regression (see Figures 6 and 7) for both experiments, where, the closer they intersect to the origin and the closer the slope is to unity the better is the approach. In order to perform a model validation we introduced an index $k = \sqrt{\left( a - 1 \right)^2 + \left( b / \overline{C}_o \right)^2}$ with

$$\overline{C}_o = \frac{1}{n} \sum_{i=1}^{n} C_{oi},$$

which if identical zero indicates a perfect match between the model and the experimental findings. Here $a$ is the slope, $b$ the intersection, $C_{oi}$ of the experimental data and $\overline{C}_o$ its arithmetic mean. Since the experiment is of stochastic character whereas the stochastic properties are hidden in the model parameters, considerable fluctuations are present. Nevertheless, by comparison one observes in Table 4 that the present approach yields the better description of the data for the Copenhagen experiment.

### 4. Conclusions

In the present work we developed a novel analytical description of air pollution dispersion in ABL. We solved the steady-state three-dimensional ADE for general vertical profiles of wind and eddy diffusivity. The closed form solution is obtained using an approach set-up by spectral theory together with an integral transform, which in the present case is the Laplace transform. This solution allows simulating dispersion of pollutant substances in a computationally efficient fashion. The advantage of an analytical procedure over the nowadays usual numerical schemes that take advantage of existing computing power is evident from the fact that once an analytical solution to a mathematical model is found one can claim that the problem has been solved, without the necessity for benchmarking. We derived a closed form solution applicable for numerical simulations in principle to any desired precision, since existence and uniqueness are guaranteed by the Cauchy-Kowalewski theorem. The quality of the solution is controlled by a genuine mathematical convergence criterion. For the Laplace inversion only bi-Lipschitz functions are acceptable, which defines then a unique relation between the original function and its Laplace-transform. This makes the transform procedure manifest exact and the only numerical error comes from truncation in the space of orthogonal functions, which may be estimated by a theorem in close analogy to

<table>
<thead>
<tr>
<th>Model</th>
<th>Regression</th>
<th>$R^2$</th>
<th>$k$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Copenhagen experiment</td>
<td>$\overline{C}_o = 0.77\overline{C}_r + 0.004$</td>
<td>0.8</td>
<td>0.23</td>
</tr>
<tr>
<td>GILTTG</td>
<td>$\overline{C}_o = 0.91\overline{C}_r + 0.34$</td>
<td>0.93</td>
<td>0.1</td>
</tr>
<tr>
<td>3D-GILTT</td>
<td>$\overline{C}_o = 0.93\overline{C}_r + 0.109$</td>
<td>0.67</td>
<td>0.32</td>
</tr>
</tbody>
</table>

Figure 6. Linear regression for the GILTTG (balls) and 3D-GILTT (stars) using the Copenhagen dataset. The bisector was added as an eye guide.

Figure 7. Linear regression for the GILTTG (balls) and 3D-GILTT (stars) using the Kinkaid dataset. The bisector was added as an eye guide.

Table 4. Comparison of the linear regressions of GILTTG and 3D-GILTT using the Copenhagen and Kinkaid datasets.
the Cardinal theorem of interpolation theory together with Parseval’s theorem. Once convergence is under control and has no longer heuristic character, a pathway opened for a genuine model validation, differently to numerical approaches where in general it is not straightforward to disentangle model errors from numerical ones.

The performance of the solution together with a parameterization of the ABL was validated against the data from the Copenhagen and the Kinkaid experiments. By comparison the present approach was found to yield an acceptable solution for the three dimensional ADE and moreover predicted tracer concentrations closer to observed values compared to other approaches from the literature. Although K-closure is known to have its limitations, the comparison of measurements and theoretical predictions showed agreement on a satisfactory level and thus supported the usage of such an approach for micro-scale dispersion phenomena.

We outline that an analytical solution can be useful in evaluating the performances of numerical models that solve numerically the ADE by comparing their results, not only against experimental data but with the analytic solution itself in order to check numerical errors without model uncertainties. Finally, the program of providing analytical solutions for close to realistic physical dispersion problems, leads us to future problems with different closure hypothesis considering full space-time dependence and thus supported the usage of such an approach for micro-scale dispersion phenomena.

We forward to disentangle model errors from numerical approaches where in general it is not straightforward to disentangle model errors from numerical ones.

5. Acknowledgements

The authors thank CNPq (Conselho Nacional de Desenvolvimento Científico e Tecnológico) and FAPERGS (Fundação de Amparo à Pesquisa do Estado do Rio Grande do Sul) for the partial financial support of this work.

REFERENCES


[44] H. R. Olesen, “Datasets and Protocol for Model Valida-