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OFFLINE APPROACH FOR HIGHER ORDER CONCENTRATION MOMENTS

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Talk outline

- Motivations
- The basic idea of a fluctuating plume model
- Simulating plume centre of mass trajectory
- Parameterization of the relative dispersion
- Case studies:
 - i)* validation of the model in homogeneous isotropic turbulence
 - ii)* comparison with the water tank experiments by Willis and Deardorff (1978) and Hibberd (2000) in the convective boundary layer
 - iii)* comparison with the water tunnel measurements and Gaussian model by Huq and Franzese (2013) in a canopy layer

Concentration fluctuations

- The concentration fluctuations play an essential role in a great number of environmental issues, such as:
 - Prediction of air pollution especially in short scales
 - Determinations of chemical reaction rates of pollutants (e.g. No_x and O_3)
 - Estimation of odour threshold
 - Analysis of turbulent combustion
- Only a few models are available to calculate at least the second moment
- The available models are subject to limitations such as:
 - Applicability only in strongly idealised conditions
 - Very elaborate numerical implementation and expensive computation
 - Reduction of reliability for small-scale turbulence.

Fluctuating plume model

Plume meandering

Internal mixing

Numerically simulated in a fixed coordinate system relative to the source

$$p_c(c; x, z) = \int_0^H p_{cr}(c; x, z, z_m) p_{zm}(x, z_m) dz_m$$

Parameterized on a local reference frame around the centre of mass

Gifford's [1959] assumption: the absolute dispersion of a passive tracer can be divided into two independent contributions: **the meandering part** and the **relative-diffusion part**.

Fluctuating plume model FPM

$$p_c(c; x, z) = \int_0^H p_{cr}(c; x, z, z_m) p_{zm}(x, z_m) dz_m$$

Absolute concentration moments

$$\langle c^n(x, z) \rangle = \int_0^\infty c^n p_c(c; x, z) dc$$

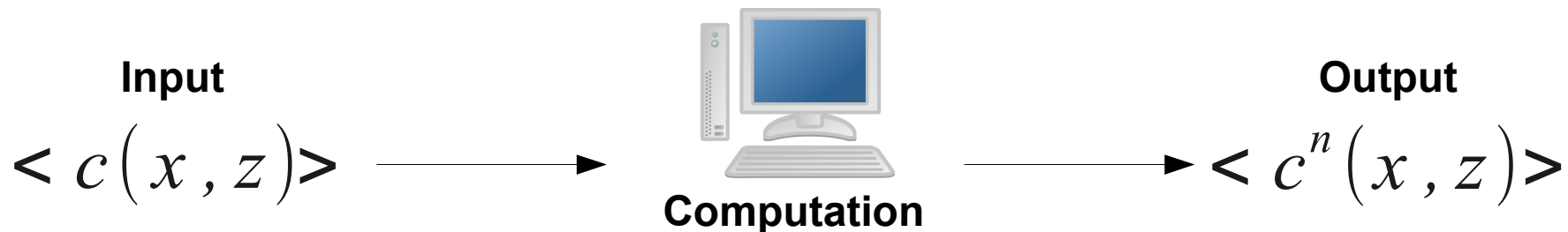
$$\langle c^n(x, z) \rangle = \int_0^H \left[\int_0^\infty c^n p_{cr}(c; x, z, z_m) dc \right] p_{zm}(x, z_m) dz_m$$

Relative concentration moments

$$\langle c_r^n(x, z) \rangle = \int_0^\infty c^n p_{cr}(c; x, z) dc$$

$$\langle c^n(x, z) \rangle = \int_0^H \langle c_r^n(x, z) \rangle p_{zm}(x, z_m) dz_m$$

Proposed FPM version



Evaluation of centre of mass vertical position PDF: Cassiani and Giostra's [2002] approach

Input: mean concentration field

$$\langle c(x, y, z) \rangle$$

Output: centre of mass position PDF

$$p_{zm} = \begin{cases} p_z \frac{\Delta z}{\Delta z_m} & \text{in the compressed axis} \\ 0 & \text{out of the compressed axis} \end{cases}$$

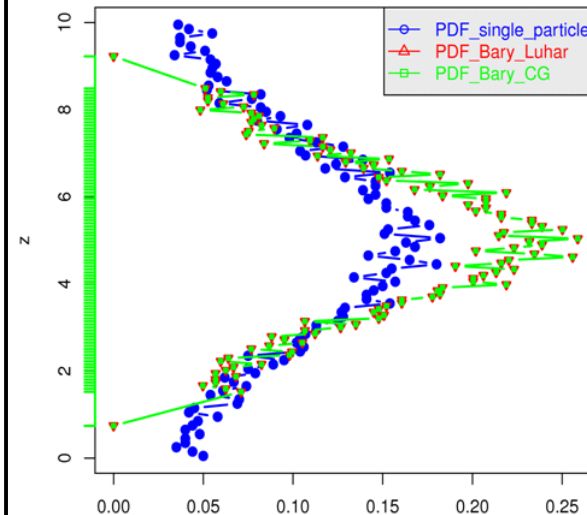
- Computation of $p_z(t, z)$, $\langle z \rangle$, σ_z^2 normalising the mean concentration

- Parameterisation of the relative spread:

$$\sigma_{zr}^2 = \frac{g \epsilon (t + t_s)^3}{[1 + \alpha (g \epsilon t^3)^{2/3}]^{3/2}}$$

Franzese [2003], Mortarini et al. [2009]

Centre of mass grid compression



$$p_{zm} > p_z$$

$$\Delta z_m < \Delta z$$

$$\frac{\sigma_z^2(t) - \sigma_{zr}^2(t)}{\sigma_z^2(t)} < 1$$

$$p_z \Delta z = p_{zm} \Delta z_m$$

Hypothesis: Luhar et al. [2000] **linear transformation** applied on the **calculation grid spacing**

$$z_{mi} = \frac{\sigma_z^2(t) - \sigma_{zr}^2(t)}{\sigma_z^2(t)} (z_i - \langle z(t) \rangle) + \langle z(t) \rangle$$

$$\Delta z_m = \frac{\sigma_z^2(t) - \sigma_{zr}^2(t)}{\sigma_z^2(t)} \Delta z$$

Parameterisation of the relative concentration PDF

$$p_{cr} = \frac{\lambda^\lambda}{\langle c_r \rangle \Gamma(\lambda)} \left(\frac{c}{\langle c_r \rangle} \right)^{\lambda-1} \exp\left(\frac{-\lambda c}{\langle c_r \rangle} \right)$$

Luhar et al. [2000]
Gailis et al. [2007]

$$\langle c_r^n \rangle = \int_0^\infty c^n p_{cr} dc$$

$$\langle c_r^n \rangle = \frac{\Gamma(\lambda+n)}{\lambda^n \Gamma(\lambda)} \langle c_r \rangle^n$$

$$\lambda^{-1} = i_{cr}^2 = \left(1 + i_{cr0}^2 \right) \left(\frac{\langle c_r \rangle}{\langle c_{r0}^n \rangle} \right)^\xi$$

Ferrero et al. [2012] Gailis et al. [2007]

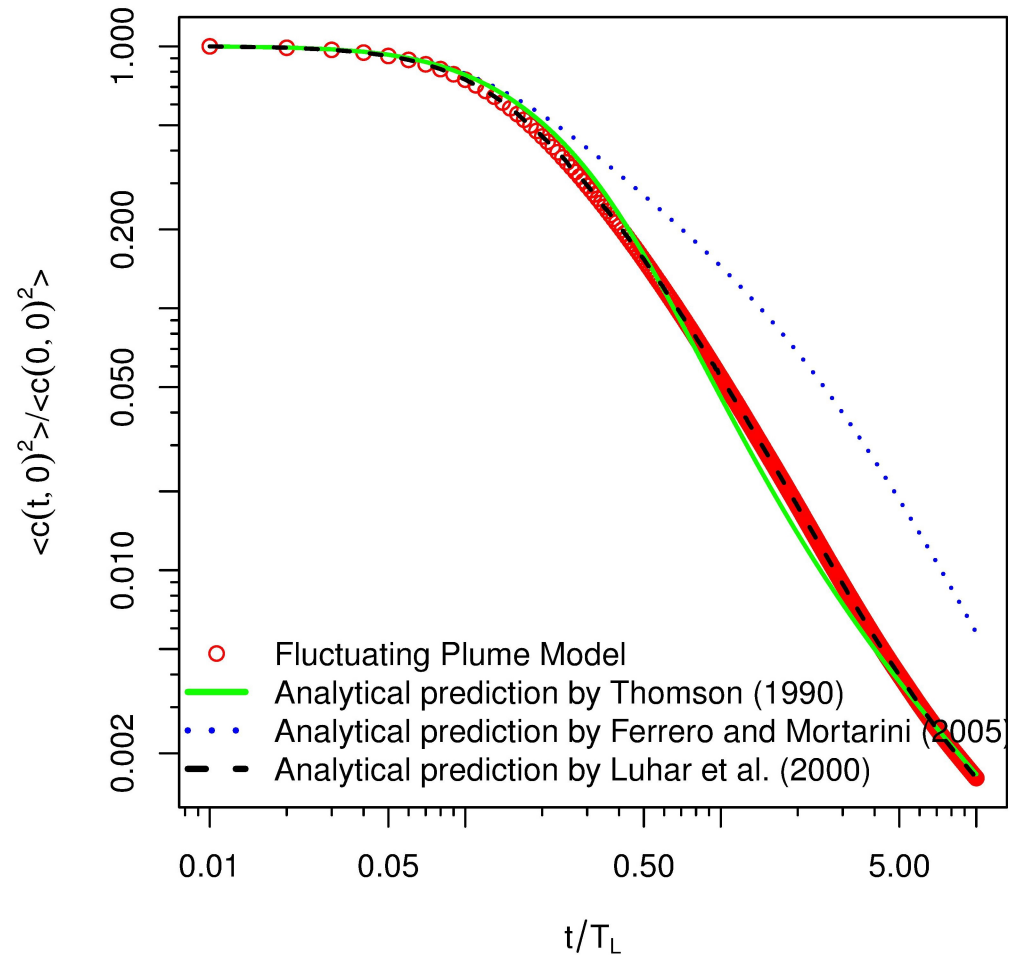
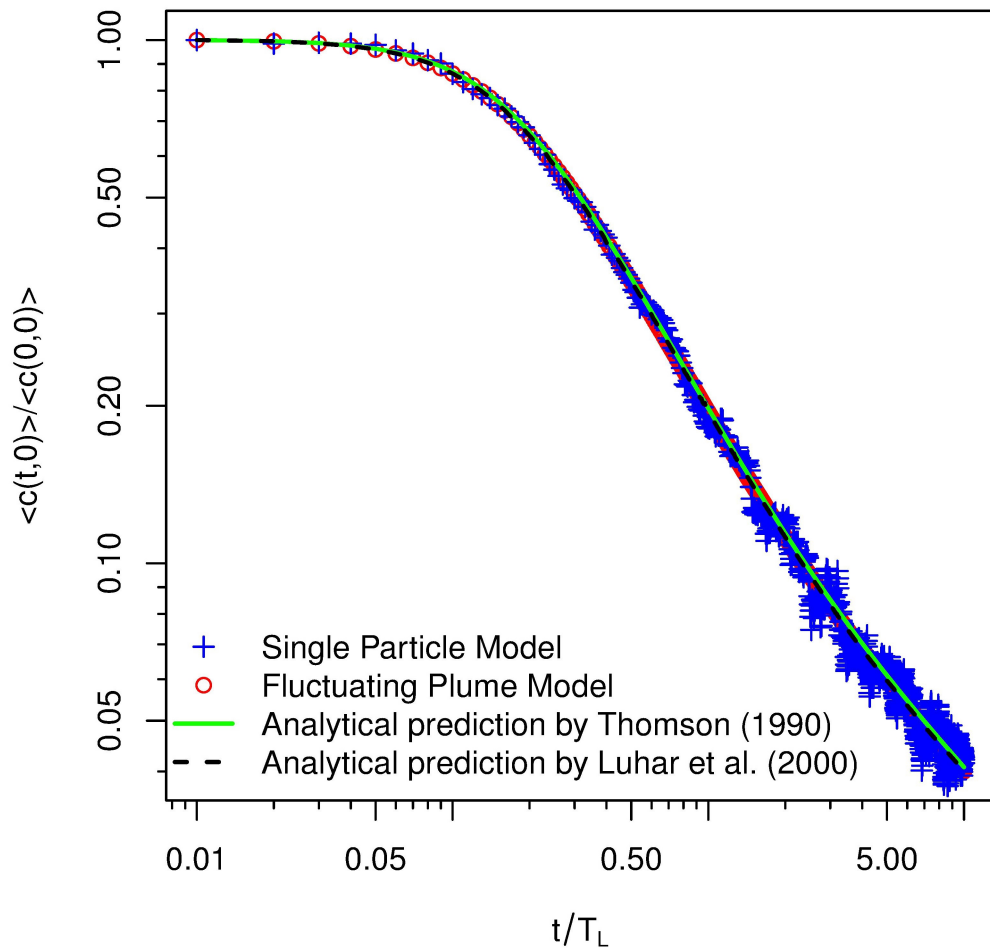
$$\langle c_r \rangle = \frac{Q}{U} p_{zr}$$

p_{zr} Gaussian or skewed bi-Gaussian

Franzese [2003] Dosio and De Arellano [2006]

Case study (I): Homogeneous Isotropic Turbulence

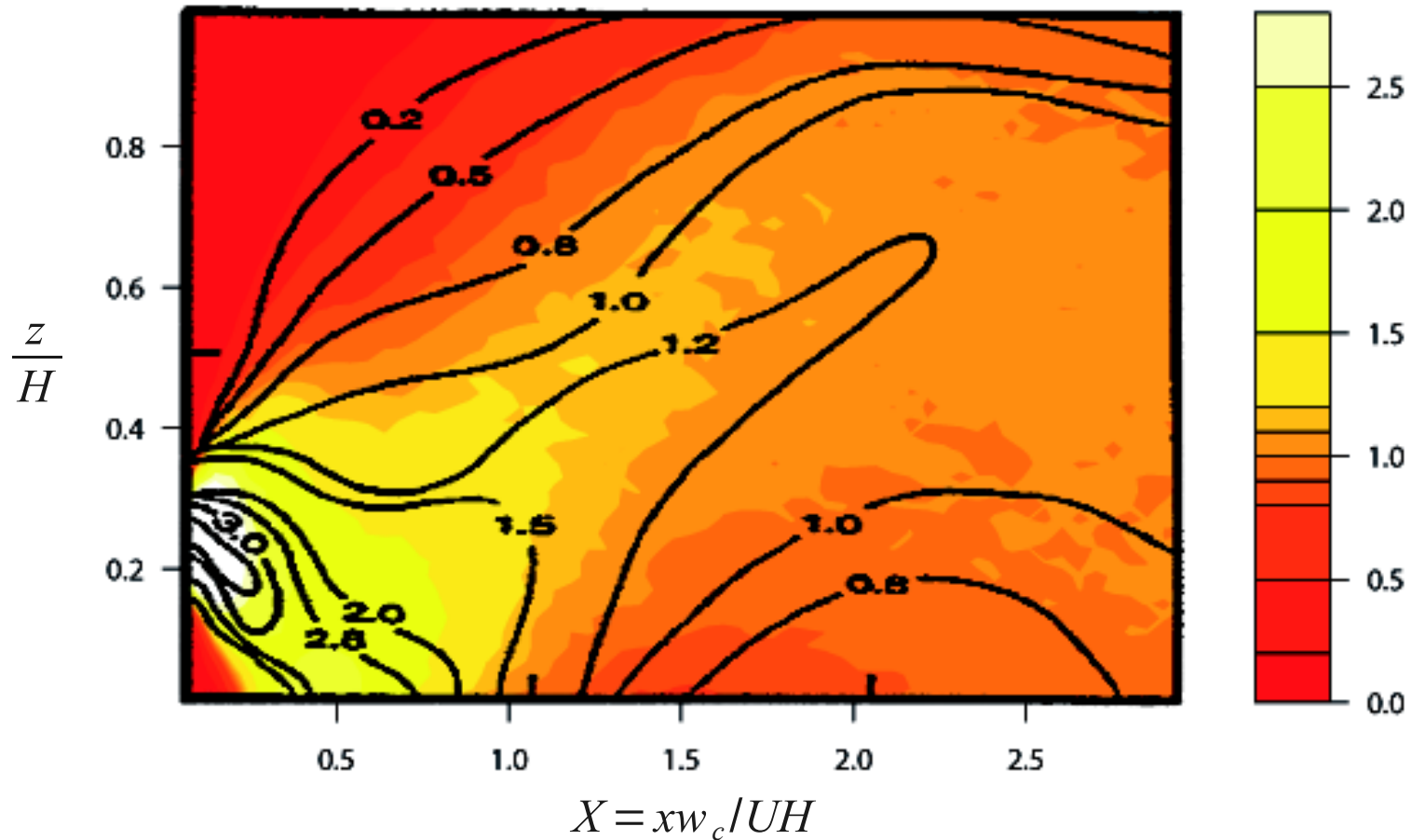
Comparison among the proposed fluctuating plume model and analytical previsions.



Case study (II): Convective Boundary Layer

Non-dimensional crosswind-integrated mean concentration field predicted by Franzese (1999) single particle model

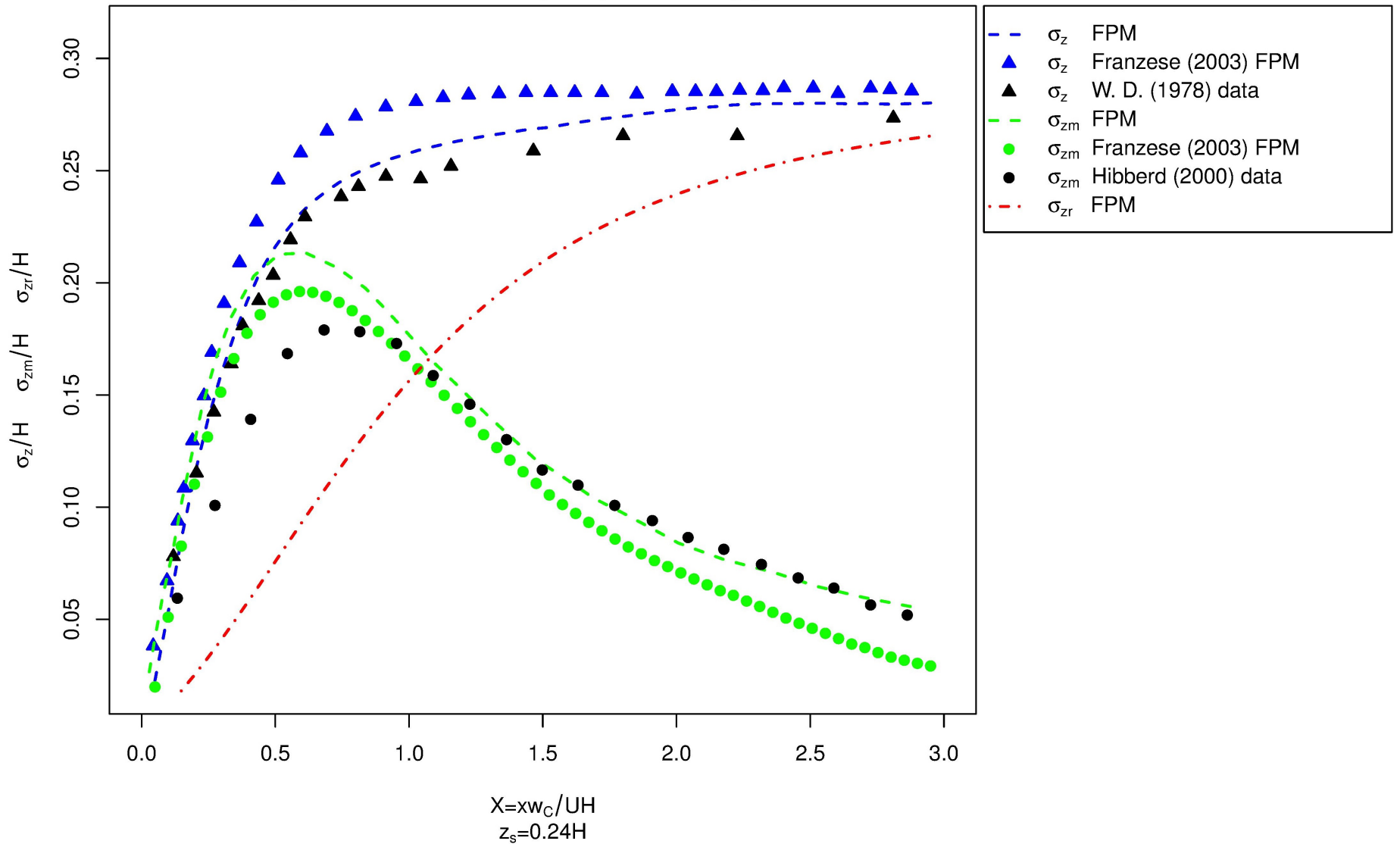
$$z_s = 0.24H$$



Filled surfaces for the fluctuating plume model presented
Black contours for Willis and Deardoff (1978) water tank data.

Case study (II): Convective Boundary Layer

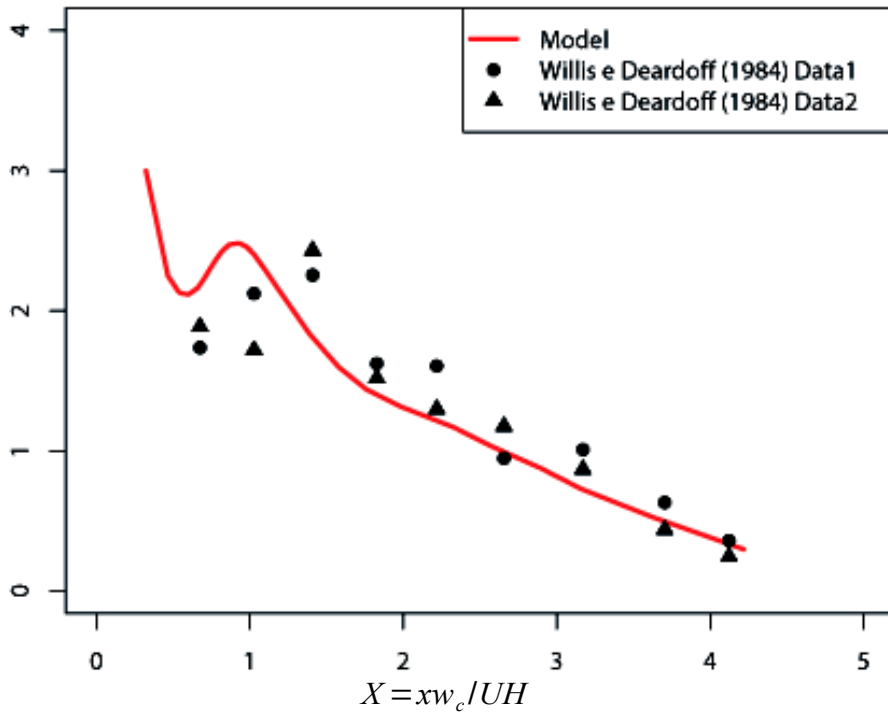
Non-dimensional absolute, relative and centroid vertical spread predicted by the model, along with the water tank data of Hibberd (2000) and Willis and Deardorff (1978) and compared with Franzese (2003) fluctuating plume model



Case study (II): Convective Boundary Layer

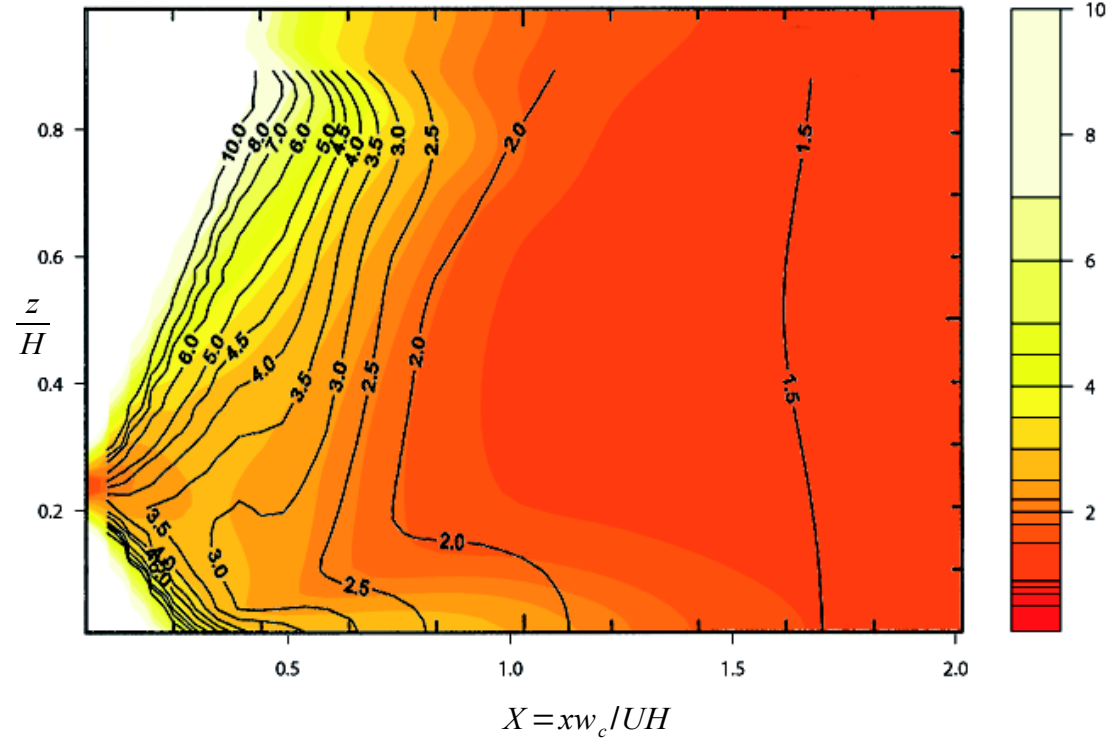
Concentration fluctuation intensity at $z=0.08H$

$z_s=0.24$



Contours of the concentration fluctuation intensity $i_c = \frac{\sigma_c}{\langle c \rangle}$

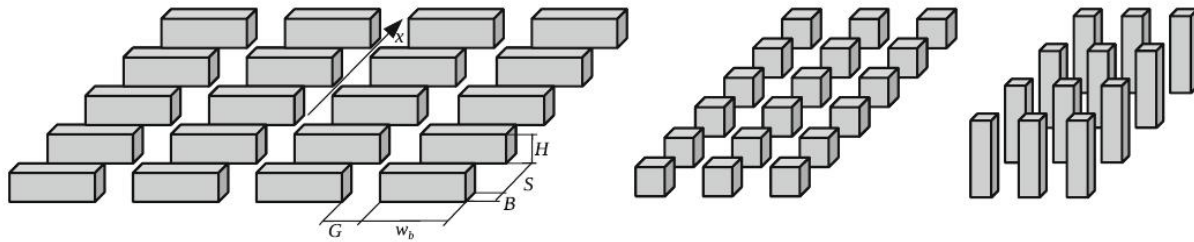
$z_s=0.24$



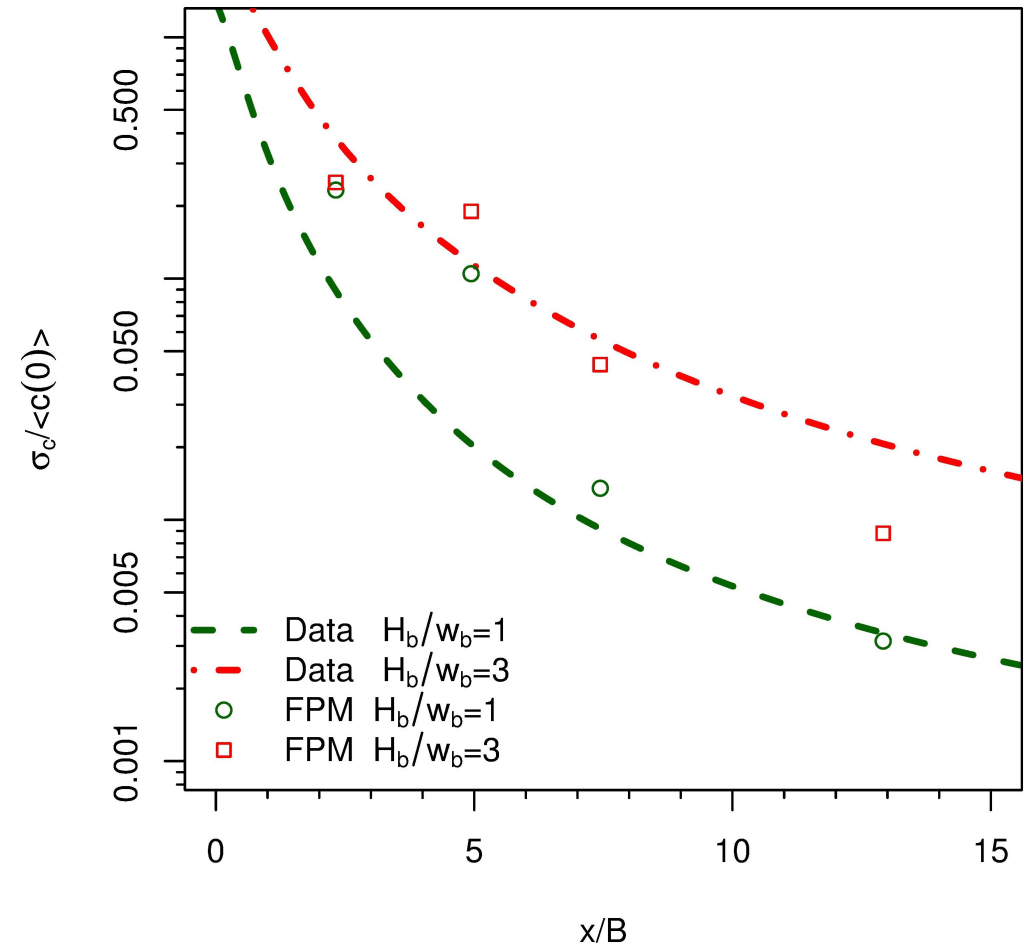
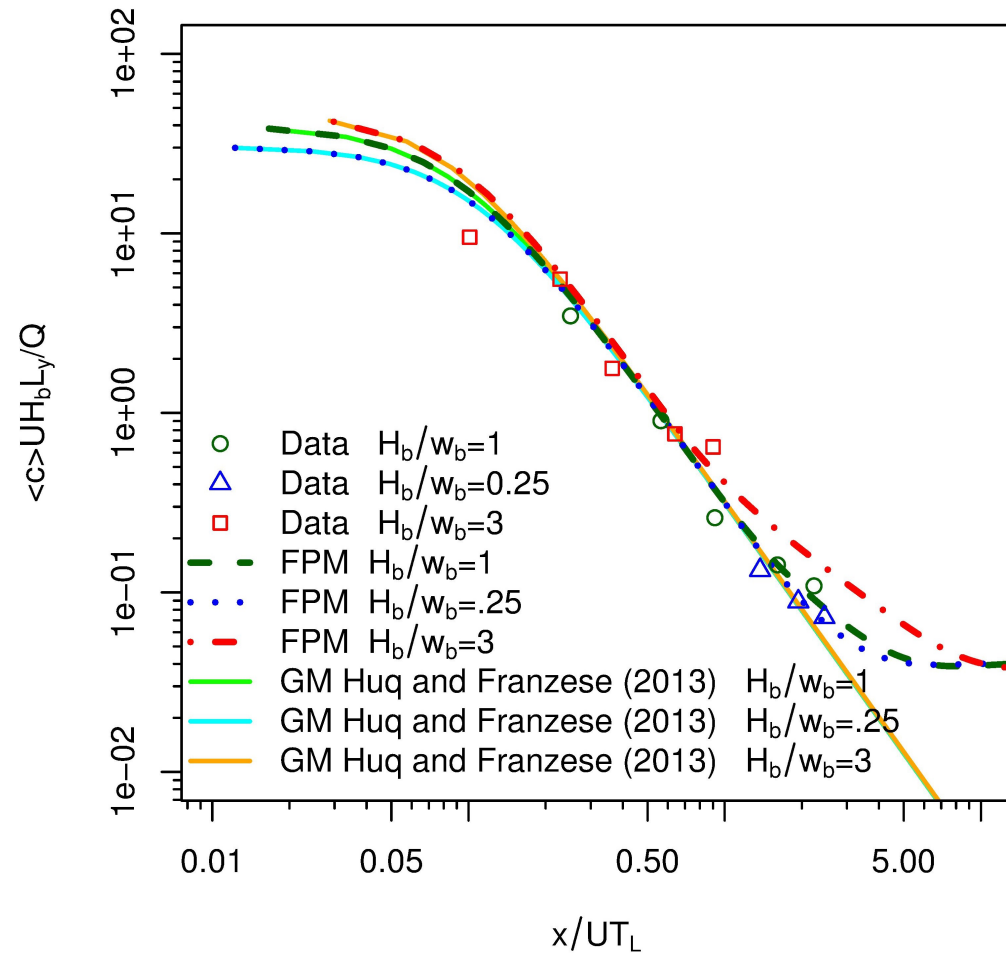
Filled surfaces for the fluctuating plume model presented
Black contours for the Luhar et al. (2000) fluctuating plume model

Case study (III): canopy layer

Comparison between the proposed fluctuating plume model and Huq and Franzese (2013) Gaussian model and water tunnel experimental data.



Schematic of the model urban canopies with $H_b/w_b=0.25$ (left), $H_b/w_b=1$ (centre), and $H_b/w_b=3$ (right). The measurements for $H_b/w_b=0.25$ were previously taken by Macdonald and Ejim (2002) to reproduce the set-up of the MUST experiment (Biltoft 2001).



Conclusions

- The model is able to simulate all the higher order moments of the PDF concentration given the first order one .
- It is independent of the method used to evaluate the requested mean concentration field which can be obtained either from models or from experiments.
- The need for Lagrangian modelling is thus avoided making the computational demand very low.
- The model can be easily adapted to different classes of turbulence modifying the parameterization of the relative-diffusion part and the mean field input data, for which both a single particle model (CBL) and a Gaussian model (canopy) have been considered.
- The overall agreement of the concentration results with the available laboratory data is good.