



Boundary layer high order concentration statistics

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evaluated by a Lagrangian model
(Luhar et al., 2000)

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- ✓ extending the single-particle model to the movement of the centroid;
- ✓ parametrizing the dispersion around the centre of the cloud.

Lagrangian model for the barycenter

Langevin
equation

$$dx_m = U(z_m)dt$$

$$dw_m = a_m(t, w_m, z_m)dt + b_m(t, z_m)dW(t)$$

$$dz_m = w_m dt$$

Fokker-Planck

$$\frac{\partial P_m}{\partial t} + w \frac{\partial P_m}{\partial z_m} = - \frac{\partial [a(t, w_m, z_m) P_m]}{\partial w_m} + \frac{\sigma_m^2}{T_m} \frac{\partial^2 P_m}{\partial w_m^2}$$

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$$b_m = \sqrt{\frac{2\langle w_m^2 \rangle}{T_m}}$$

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single-particle
Lagrangian Time Scale



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Quadratic
assumption

$$a(w_m, z_m, t) = \alpha_m(z_m, t)w_m^2 + \beta_m(z_m, t)w_m + \gamma_m(z_m, t)$$

(Franzese, 2003)

Lagrangian model for the barycenter

$$\frac{\partial}{\partial t} \langle w^m \rangle + \alpha_z \langle w^{m+1} \rangle + \beta_z \langle w^m \rangle + \gamma_z \langle w^{m-1} \rangle = \frac{1}{m} \frac{\partial \langle w^{m+1} \rangle}{\partial z} - 2(m-1) \frac{\sigma_m^2}{T_m} \langle w^{m-2} \rangle$$

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Centroid Vertical Acceleration Parameters:

$$\alpha(z_m) = \frac{(1/3)(\partial \langle w_m^3 \rangle / \partial t + \langle w_m^4 \rangle / \partial z_m)}{\langle w_m^4 \rangle - \langle w_m^3 \rangle^2 / \langle w_m^2 \rangle - \langle w_m^2 \rangle^2} - \frac{\langle w_m^3 \rangle / (2 \langle w_m^2 \rangle) [\partial \langle w_m^3 \rangle / \partial z_m - 2 \langle w_m^2 \rangle / T_m] + \langle w_m^2 \rangle \partial \langle w_m^2 \rangle / \partial z_m}{\langle w_m^4 \rangle - \langle w_m^3 \rangle^2 / \langle w_m^2 \rangle - \langle w_m^2 \rangle^2}$$

$$\beta_m(z_m) = \frac{1}{2 \langle w_m^2 \rangle} \left[\frac{\partial \langle w_m^2 \rangle}{\partial t} + \frac{\partial \langle w_m^3 \rangle}{\partial z_m} - 2 \langle w_m^3 \rangle \alpha_m(z_m) \right] - \frac{1}{T_m}$$


$$\gamma_m(z_m) = \frac{\partial \langle w_m^2 \rangle}{\partial z_m} - \langle w^2 \rangle \alpha_m(z_m)$$

Energy Filter and Partition of Energy

$$\langle w^2 \rangle = \langle w_m^2 \rangle + \langle w_r^2 \rangle$$

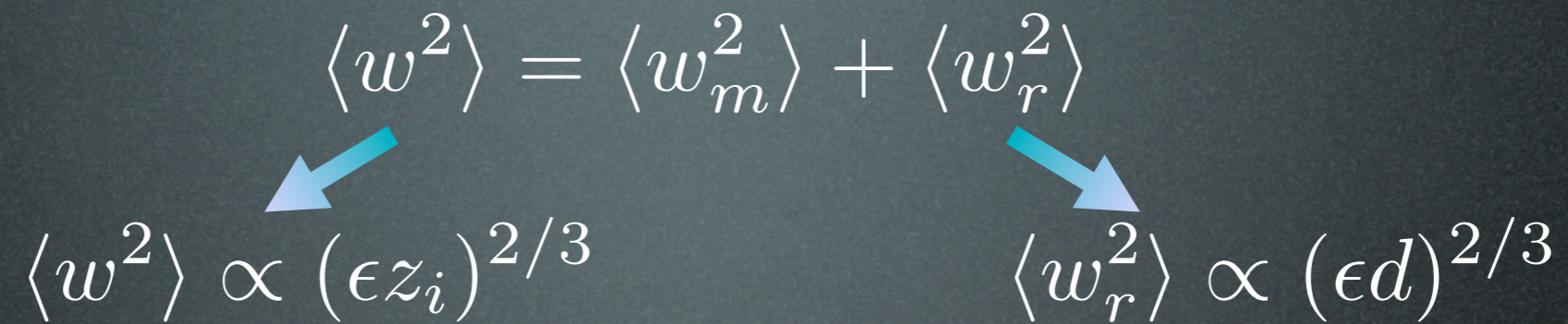
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Filter Function

$$\langle w_m^n \rangle = \langle w^n \rangle \left[1 - \left(\frac{d^2}{d^2 + z_i^2} \right)^{\frac{1}{3}} \right]^{n/2}$$

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$$t_s = \left[\sigma_0^2 / (g \epsilon) \right]^{1/3}$$

Relative Concentration PDF

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Parametrization of the relative concentration

$$p_{cr}(c|x, z, z_m) = \frac{\lambda^\lambda}{\langle c_r \rangle \Gamma(\lambda)} \left(\frac{c}{\langle c_r \rangle} \right)^{\lambda-1} e^{-\frac{\lambda c}{\langle c_r \rangle}}$$

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$\lambda = 1/i_{cr}^2$

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(Luhar et al., 2000)

Relative vertical position PDF

Gaussian Parametrisation

$$p_{zr}(x, z, z_m) = \frac{1}{\sqrt{2\pi}\sigma_{zr}} \sum_{n=-N}^N \left[e^{-\frac{(z - z_m + 2nz_i)^2}{2\sigma_{zr}^2}} + e^{-\frac{(-z - z_m + 2nz_i)^2}{2\sigma_{zr}^2}} \right]$$

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Skewed Parametrisation

$$p_{zr}(x, z, z_m) = \sum_{j=1}^2 \sum_{n=-N}^N \frac{a_j}{\sqrt{2\pi}\sigma_j} \left[e^{-\frac{(z - z_m + 2nz_i - \bar{z}_j)^2}{2\sigma_j^2}} + e^{-\frac{(-z - z_m + 2nz_i - \bar{z}_j)^2}{2\sigma_j^2}} \right]$$

Luhar et al. (2000)

Dosio and De Arellano (2006)

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$$\left. + e^{-\frac{(-z - z_m + 2nz_i - \bar{z}_j)^2}{2\sigma_j^2}} \right]$$

$$S_{zr} = \frac{\langle (z - \langle z \rangle)^3 \rangle - \langle (z_m - \langle z_m \rangle)^3 \rangle}{\sigma_{zr}^3}$$

Case studies:

CBL

Water tank dispersion experiments:
Willis and Deardorff (1976, 1978,
1981).

The turbulence used as input of the
Lagrangian stochastic model is described
in Franzese et al. (1999) and Franzese
(2003).

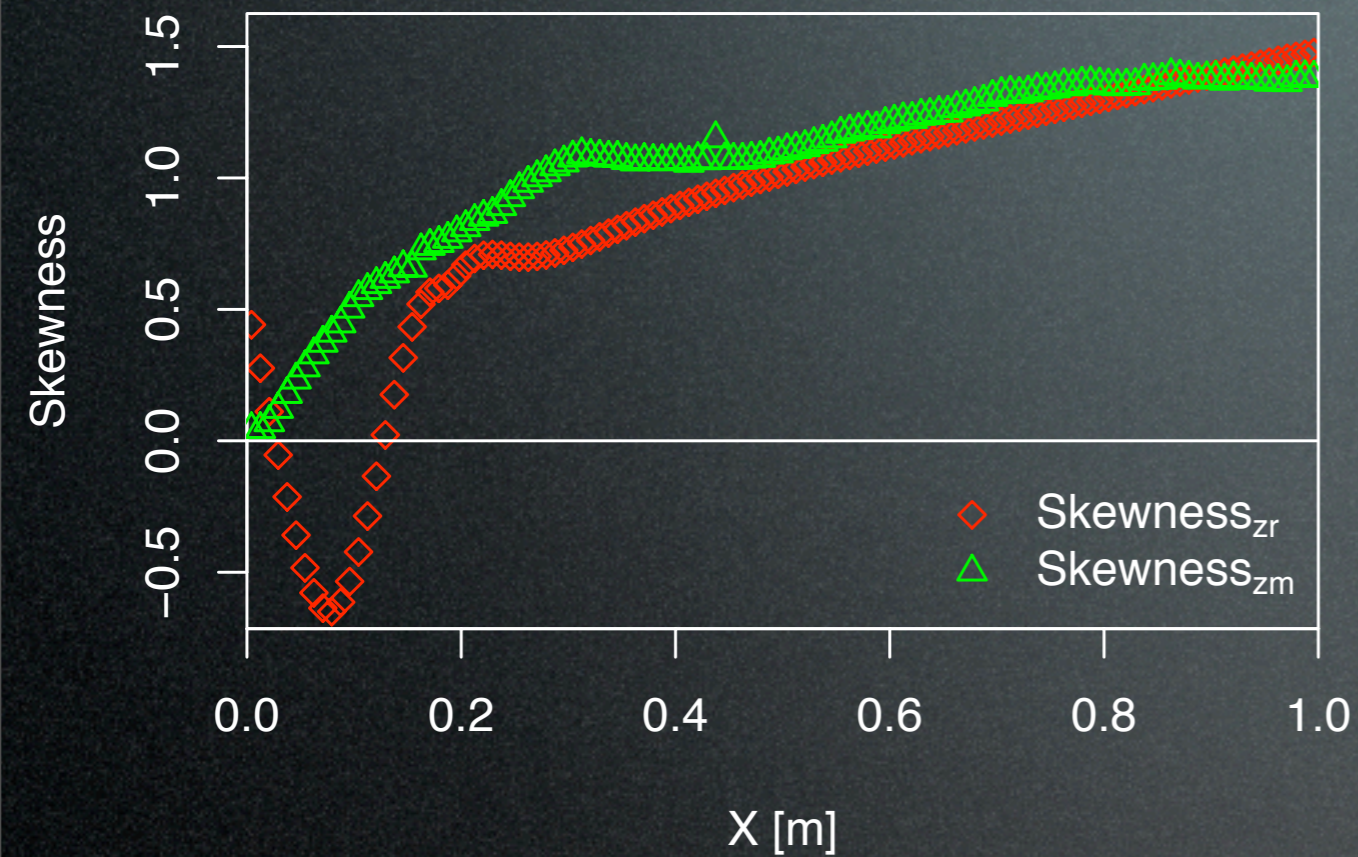
Canopy

Simulated vegetal canopy from wind
tunnel experiments: Raupach et al.
(1986), Legg et al. (1986), Coppin et al.
(1986).

The turbulence used as input for the
Lagrangian stochastic model was
derived by polynomial and spline fit of
the experimental data of Raupach et al.
(1986), Legg et al. (1986), Cassiani et
al. (2007).

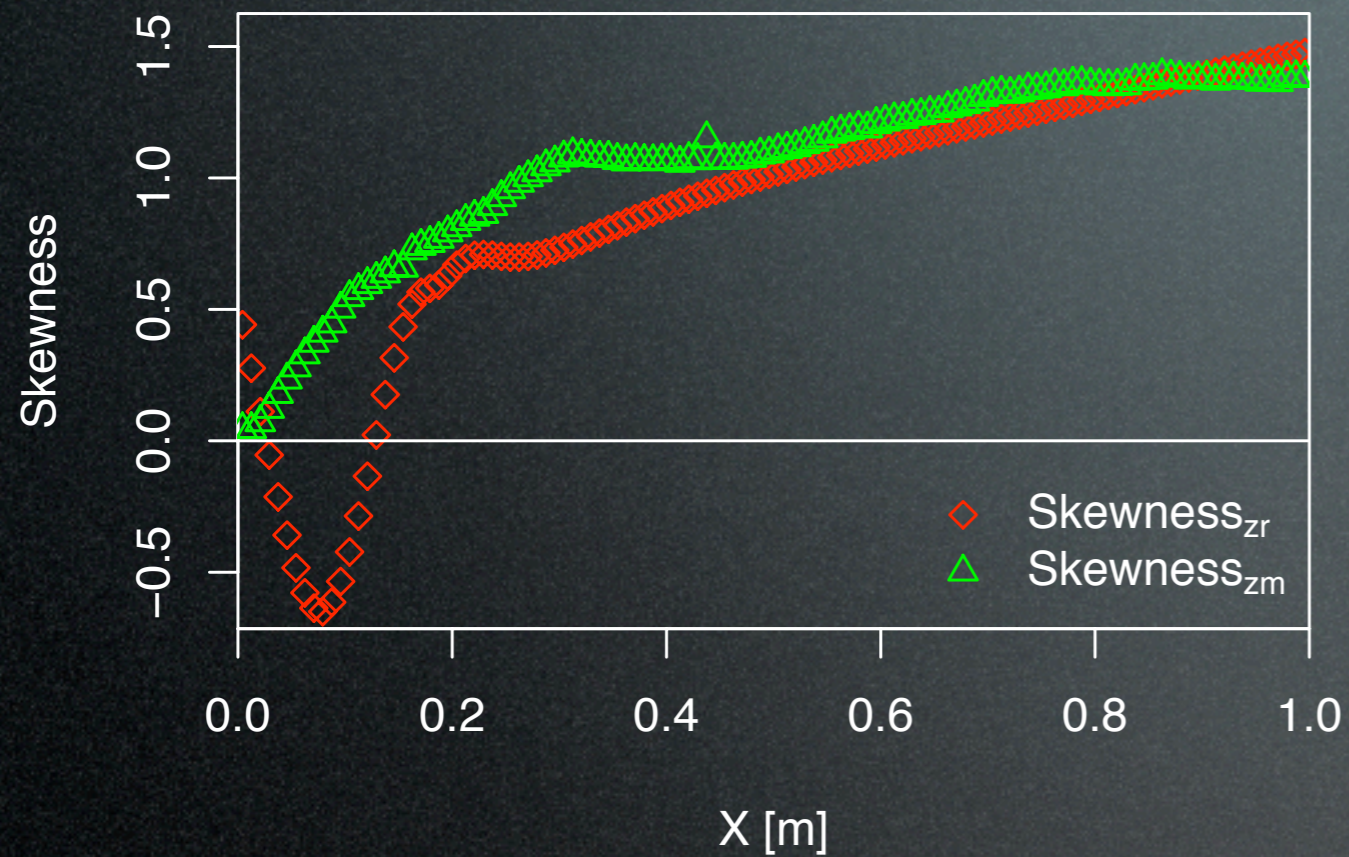
Skewness of the single particle vertical position relative to the barycenter

Canopy

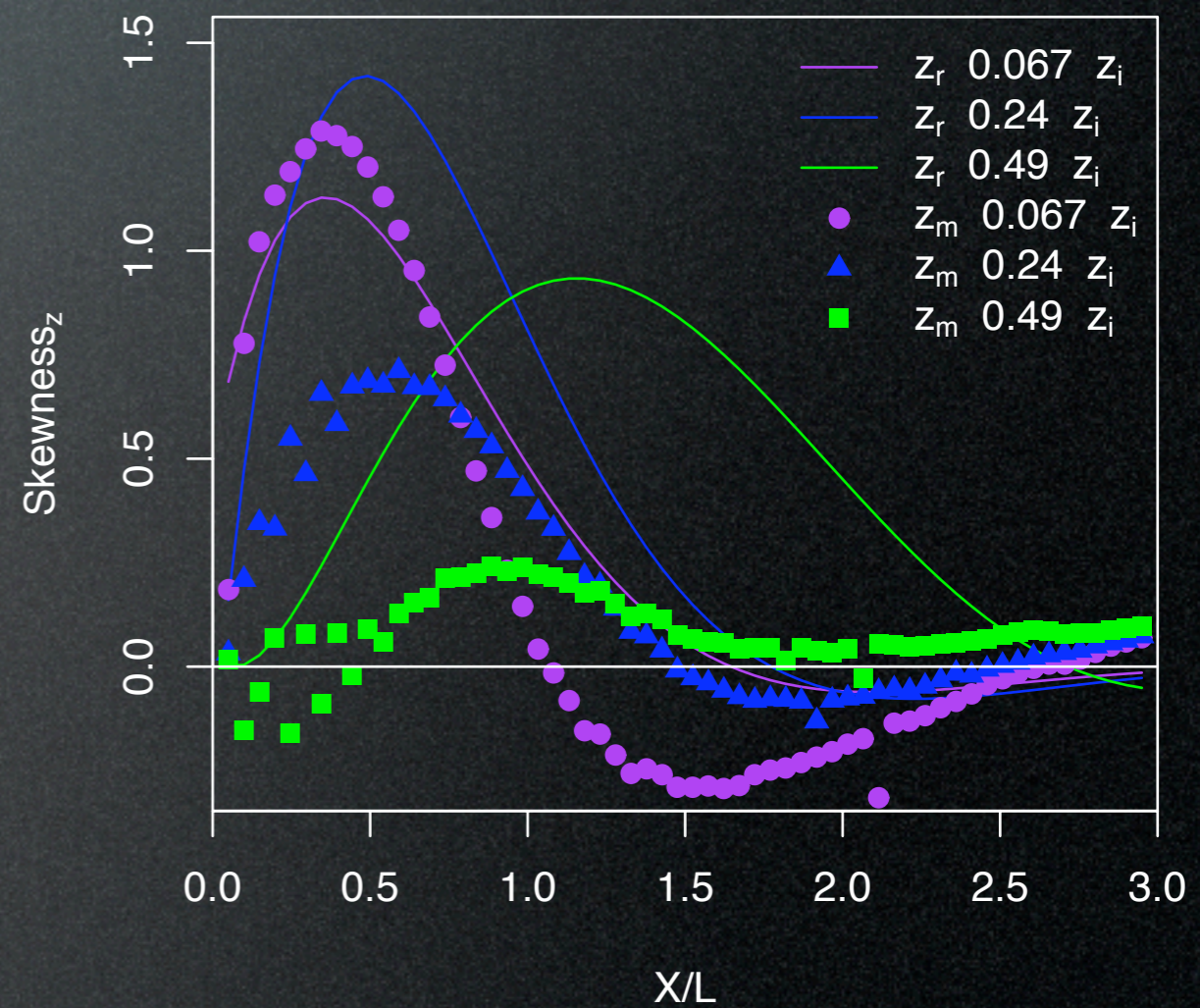


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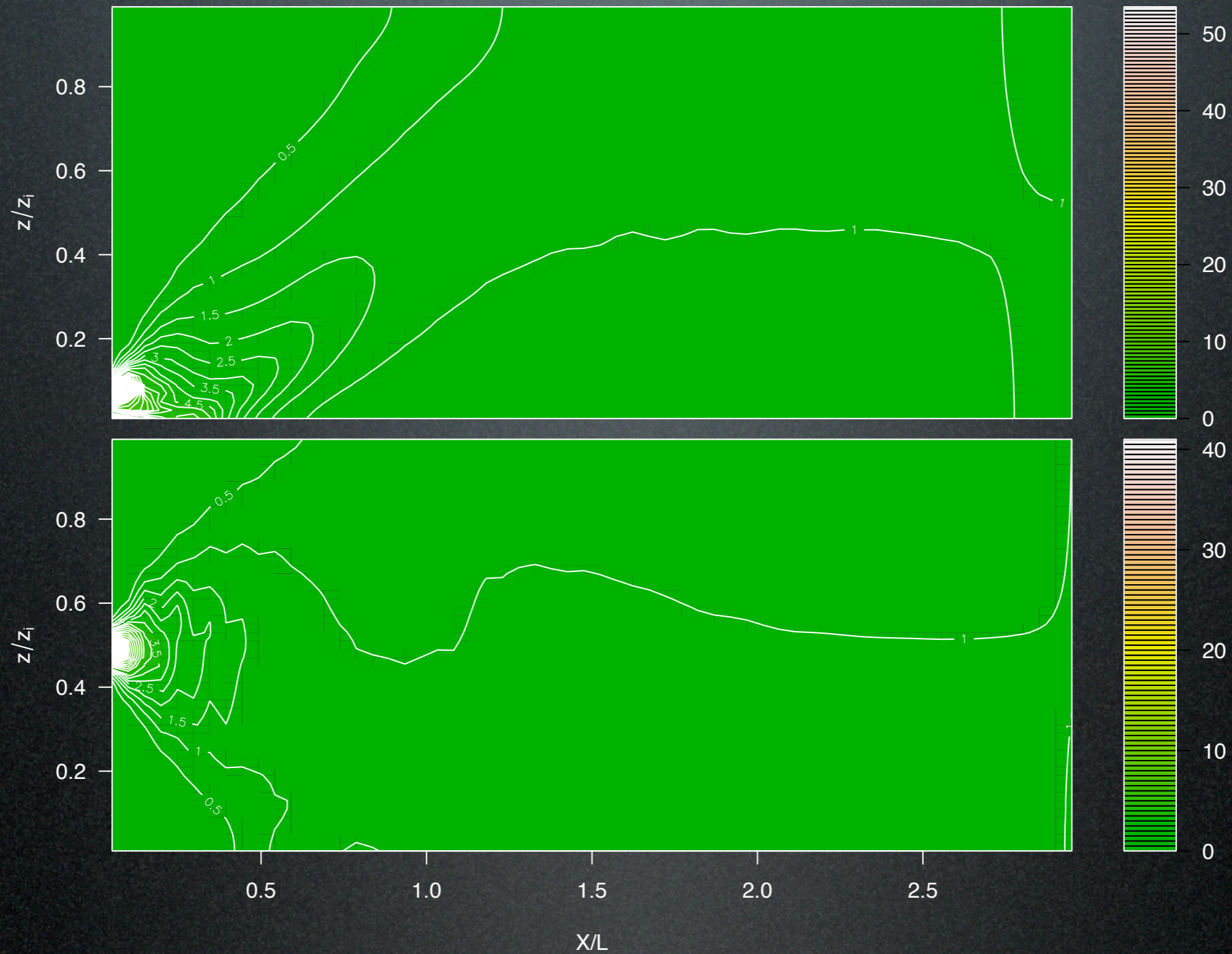


CBL



Mean Concentration Field - CBL

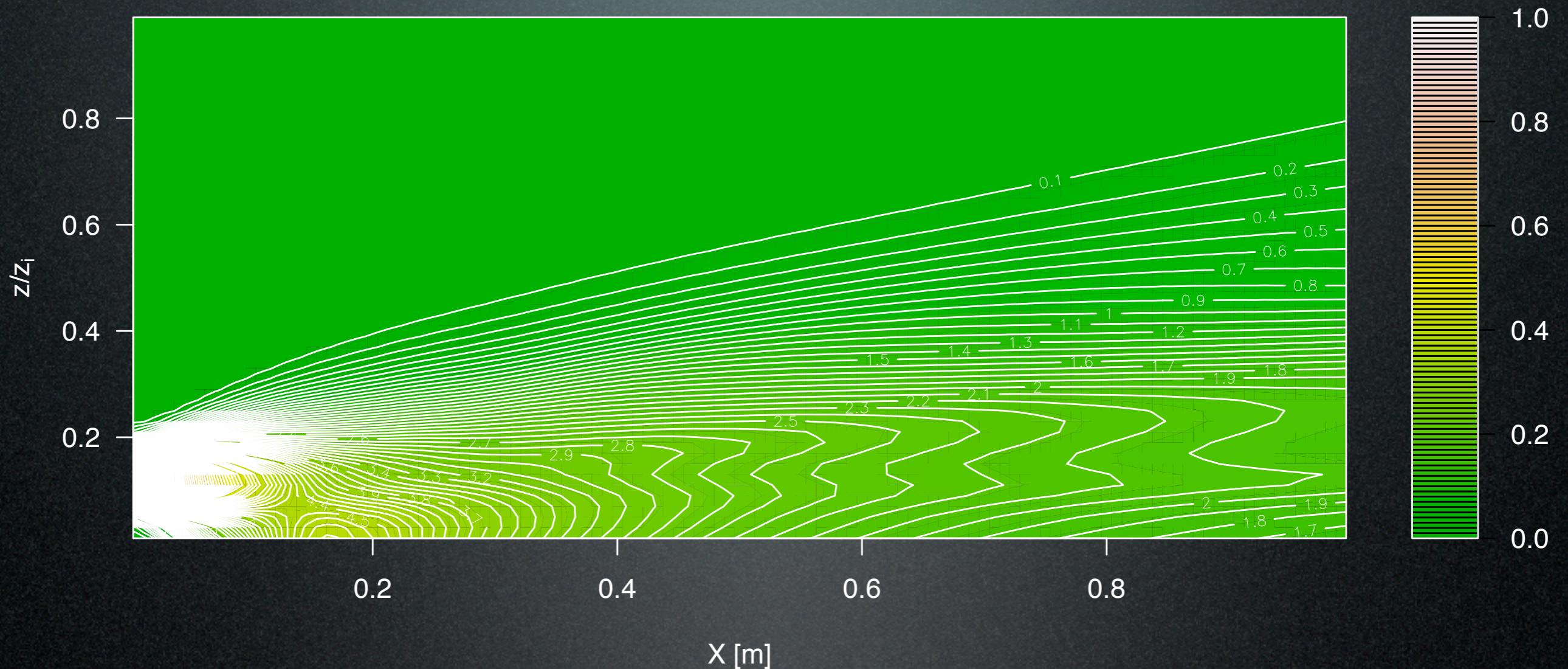
CBL



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Mean Concentration Field - Canopy

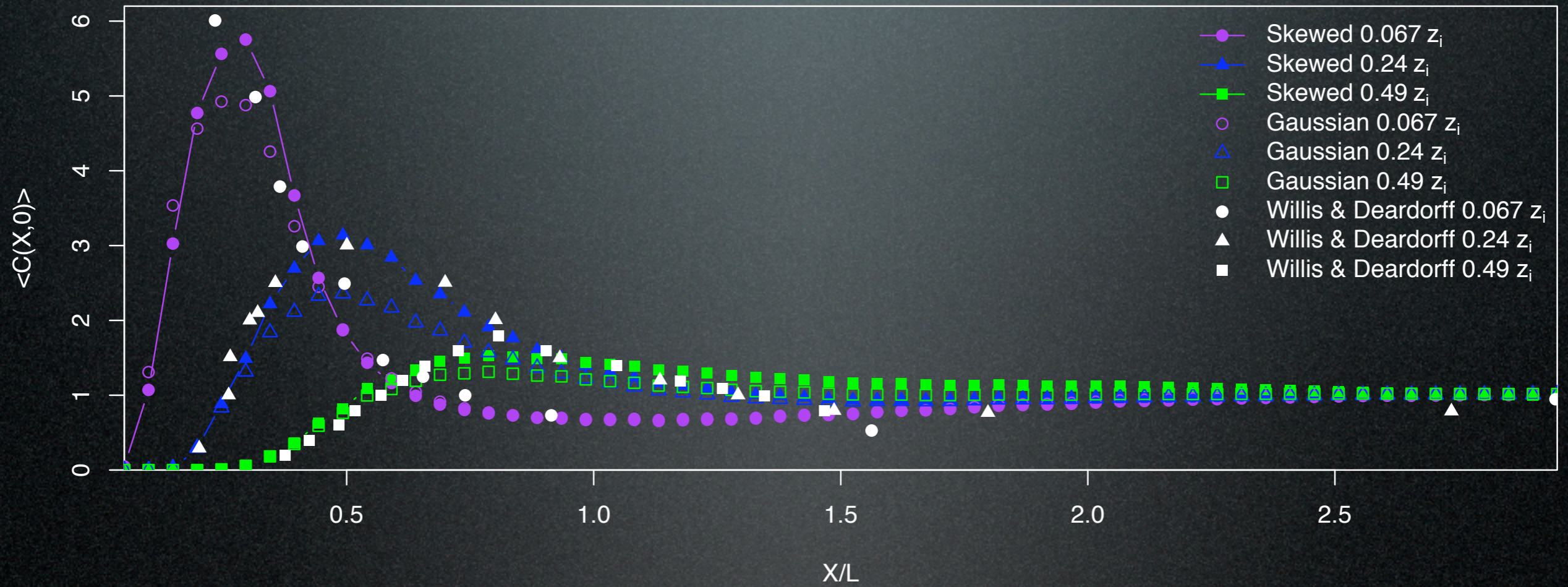
Canopy



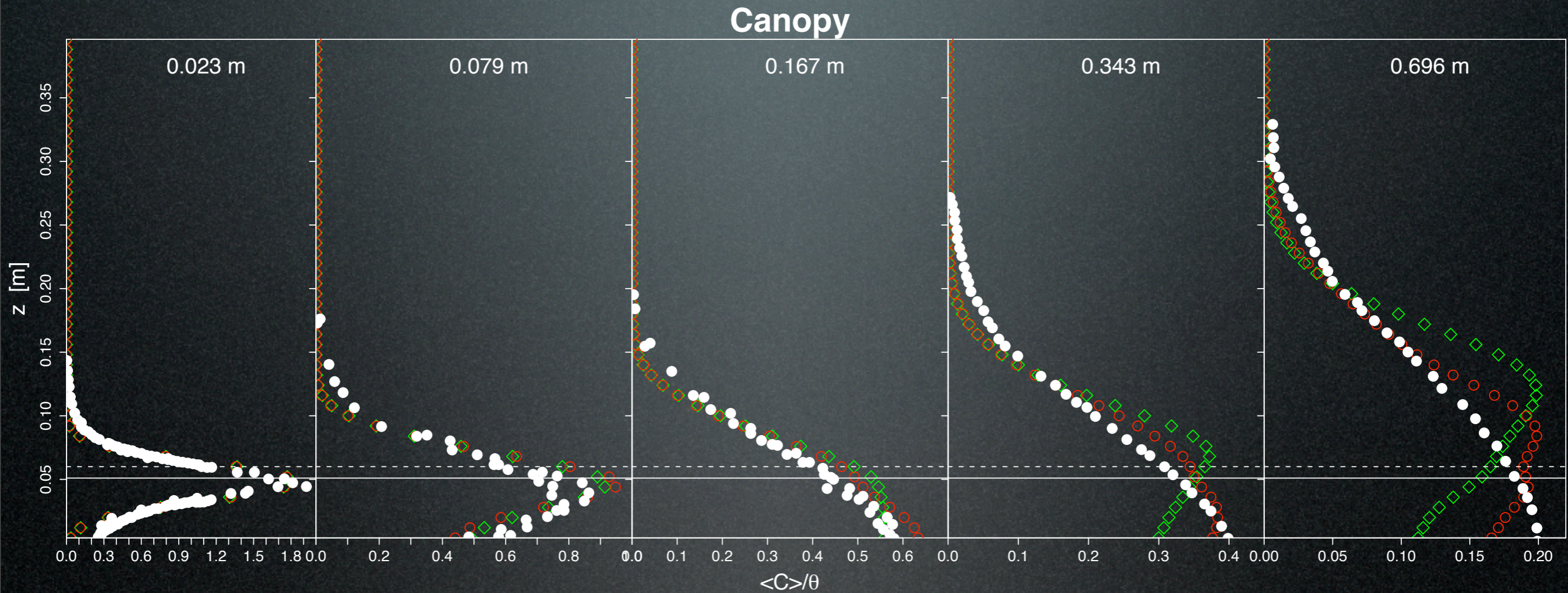
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Ground Level Mean Concentration - CBL

CBL



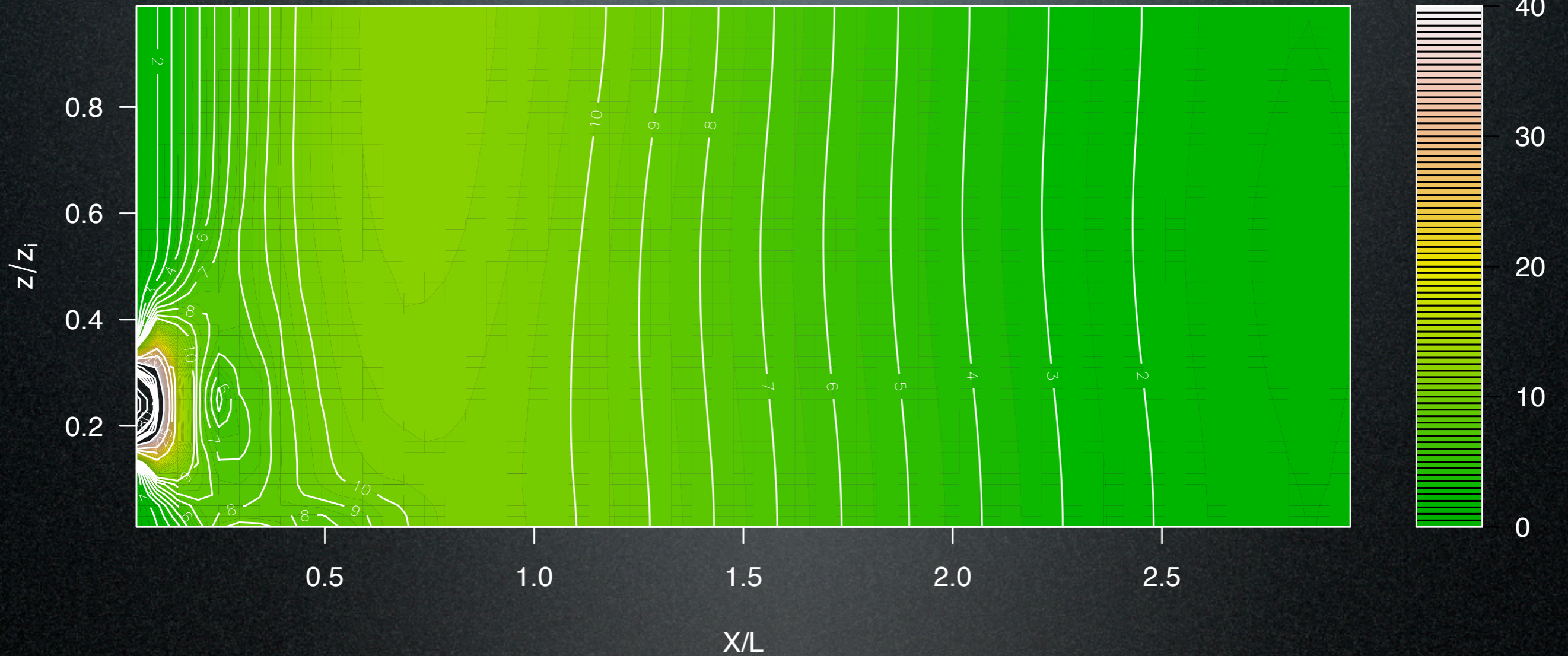
Mean Concentration Profiles- Canopy



- Legg et al. (1986) data
- ◆ Fluctuating Plume Model, Gaussian Formulation
- Fluctuating Plume Model, Skewed Formulation

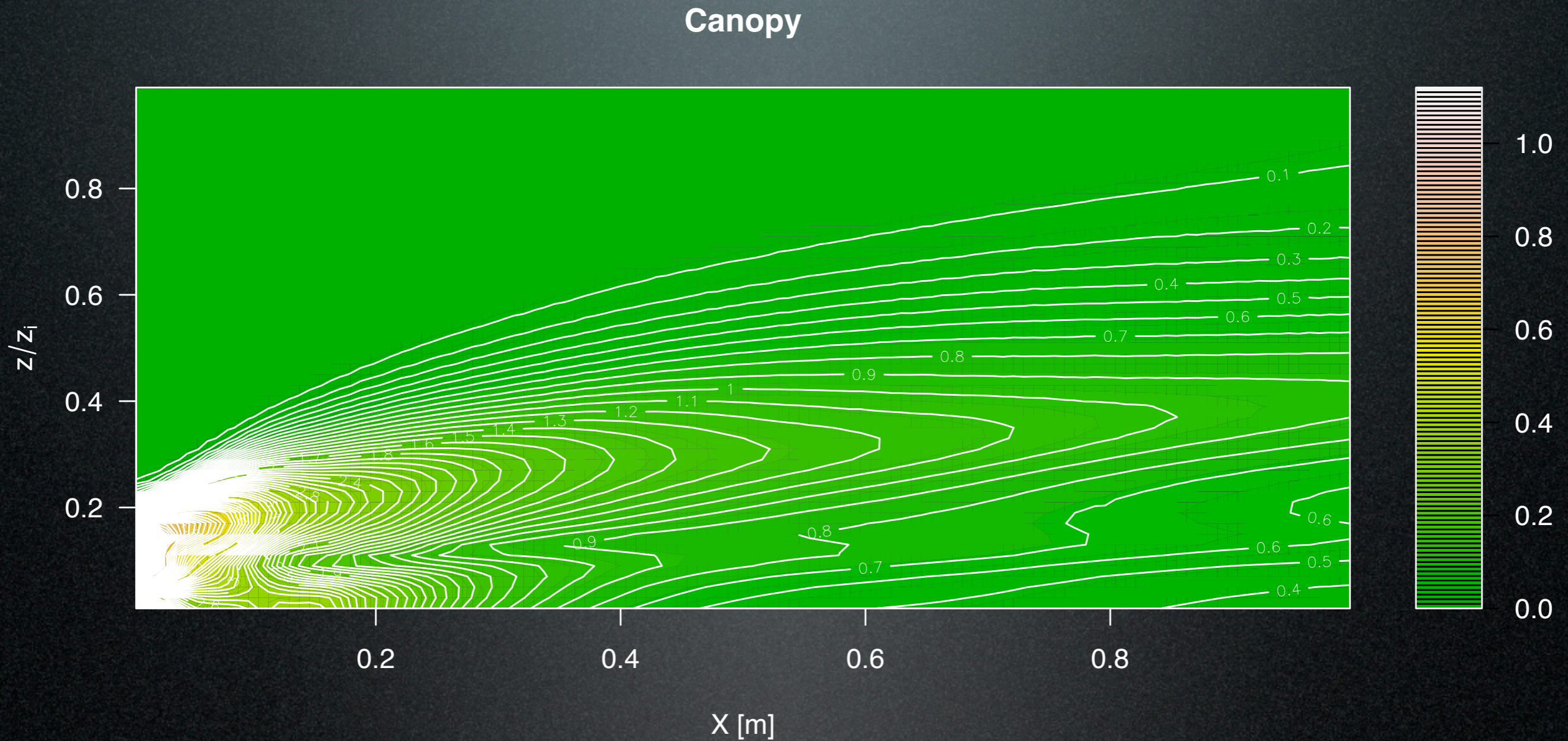
Concentration Fluctuations' Field - CBL

CBL



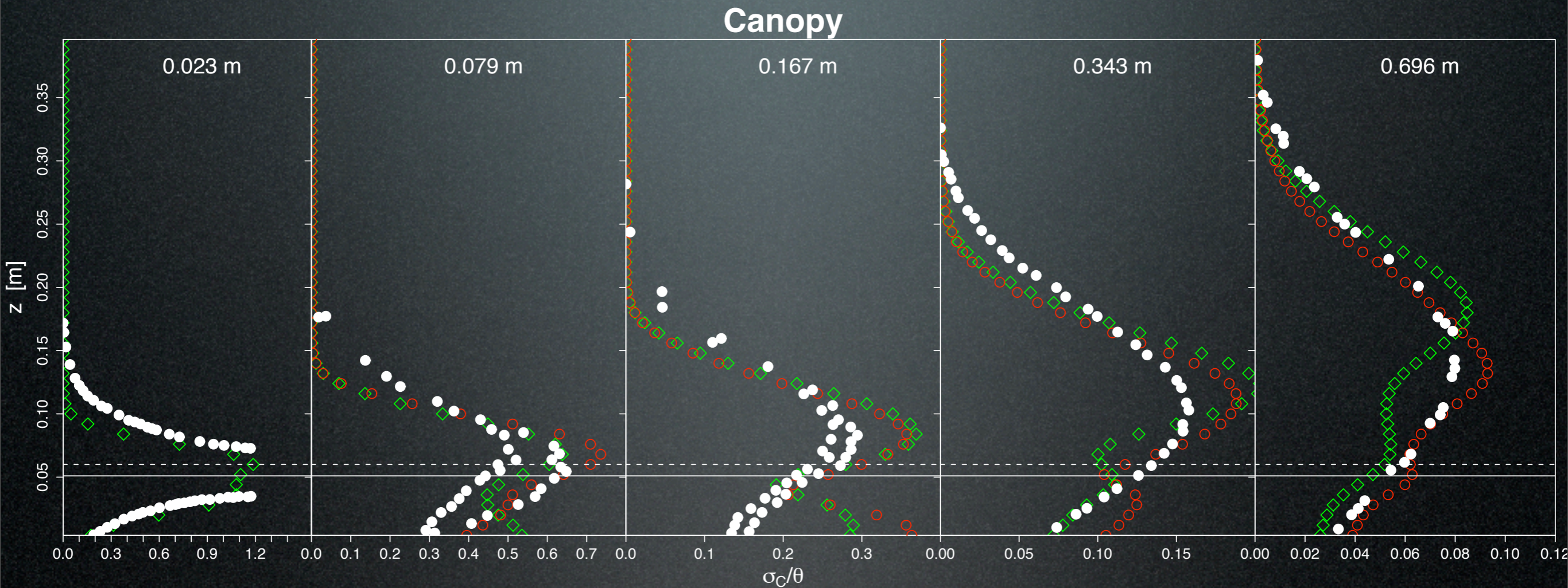
$$\langle c^2(x, z) \rangle = i_{cr}^4 \frac{\Gamma(1/i_{cr}^2 + 2)}{\Gamma(1/i_{cr}^2)} \left(\frac{Q}{U} \right)^2 \int_0^{z_i} p_{zr}^2(x, z, z_m) p_m(x, z_m) dz_m$$

Concentration Fluctuations' Field - Canopy



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Concentration Fluctuations' Profiles - Canopy



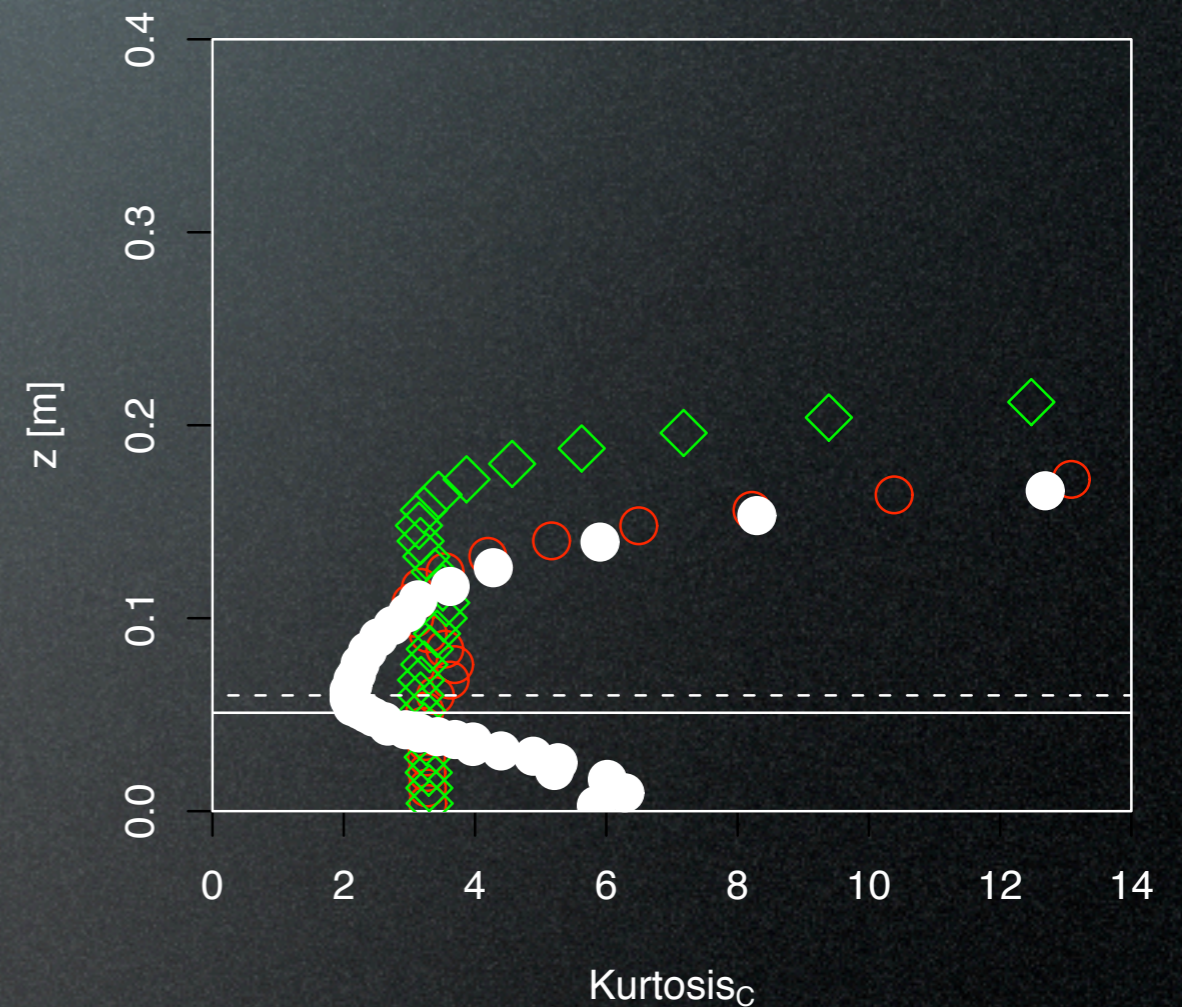
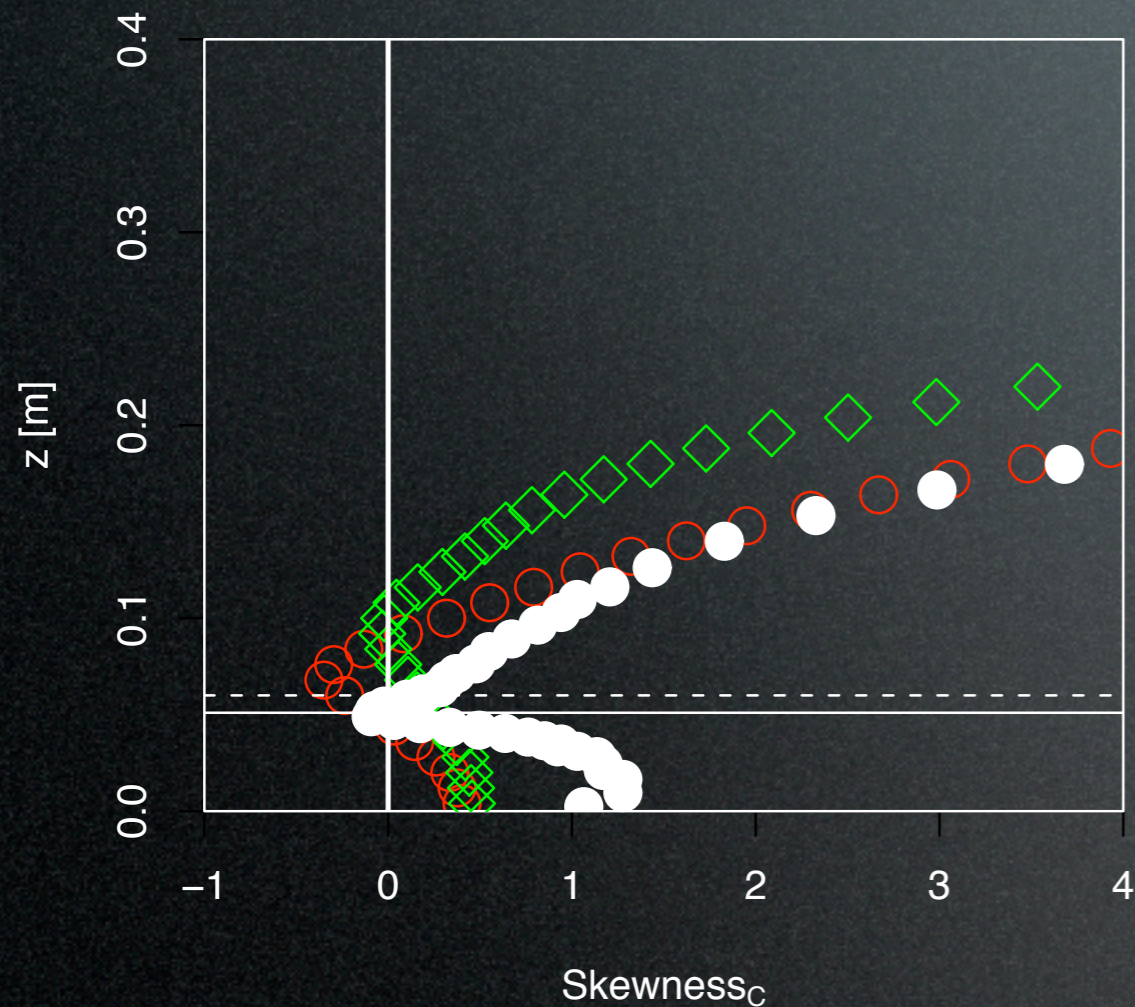
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High Order Concentration Statistics- Canopy

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0.696 m

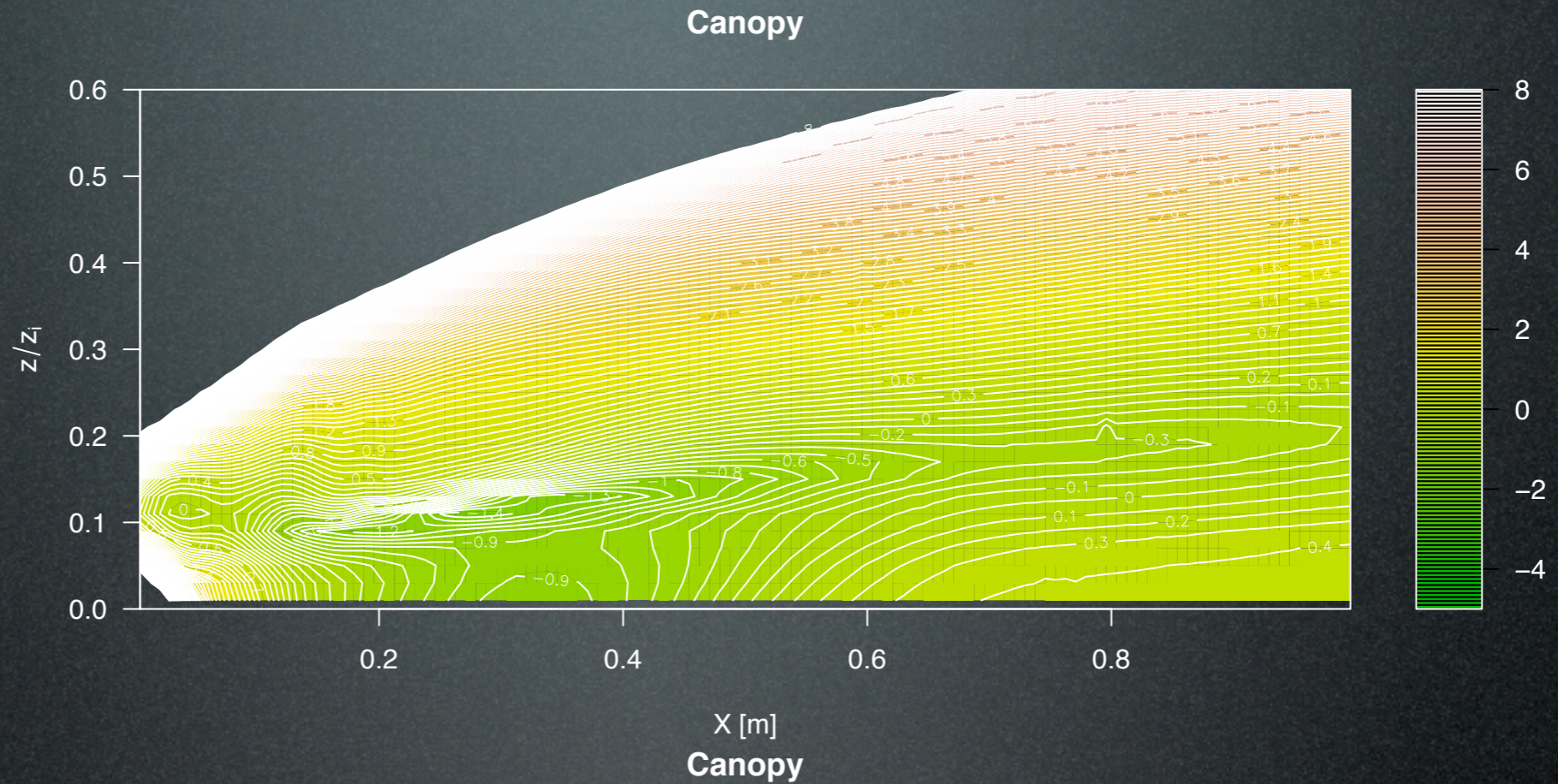
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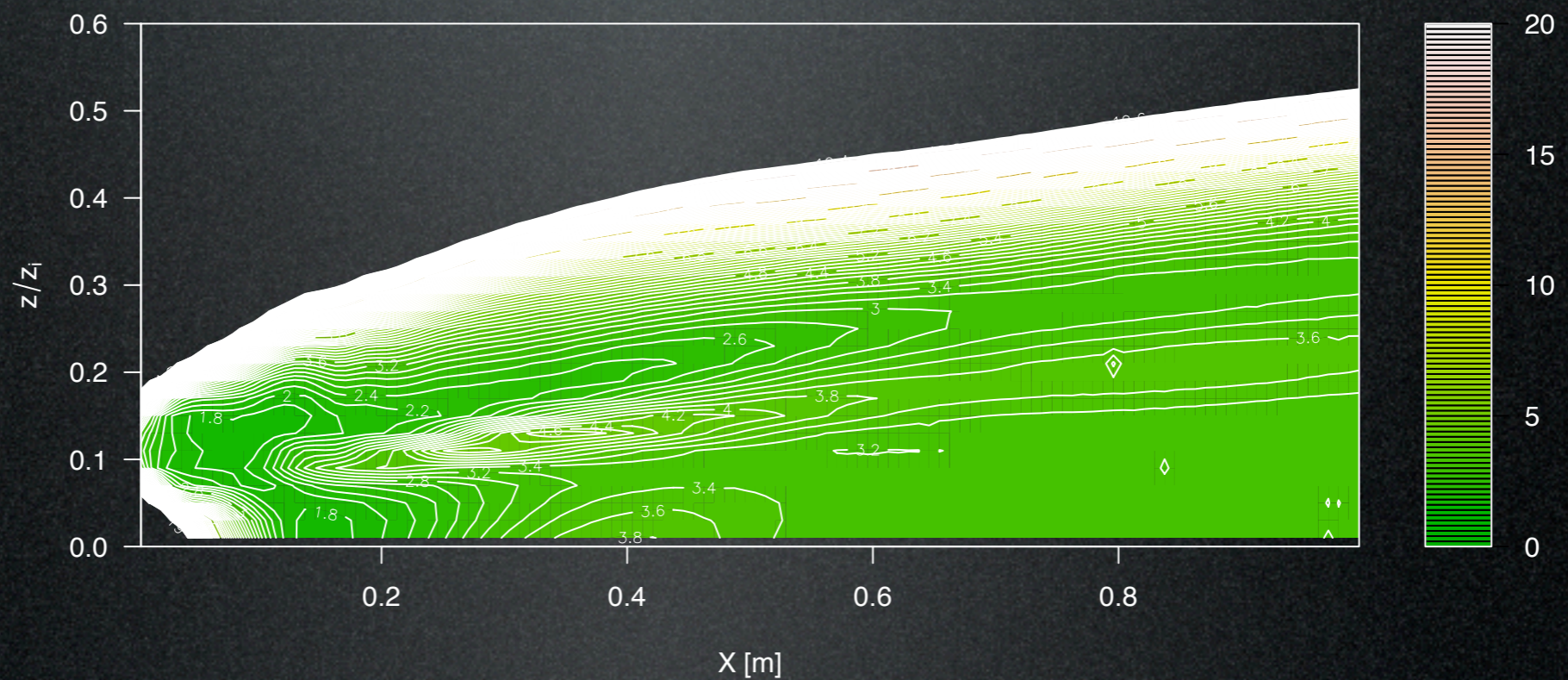
- Coppin et al. (1986) data
- ◇ Fluctuating Plume Model, Gaussian Formulation
- Fluctuating Plume Model, Skewed Formulation

High Order Concentration Statistics- Canopy

Skewness



Kurtosis



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Need of datasets.

Thank you!

luca.mortarini@unipmn.it