

# Investigation of the dispersion of air pollutants by the RePLaT model



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# Lagrangian dispersion models

## □ “ghost” particles (computational particles)

- point-like particles
- with artificial, **time-dependent** mass (e.g.  $m_p = 1$  kg) → mass decreases exponentially due to deposition:

$$\Delta m_p / \Delta t = -C \cdot m_p$$

- trajectory is determined by the atmospheric flows + in some models **mean** settling/terminal velocity
- e.g.: FLEXPART, HYSPLIT

## □ “real” particles

- particles with **fixed, realistic** size and density (e.g.  $r = 1 \mu\text{m}$ ,  $\rho_p = 2000 \text{ kg/m}^3$ )
- trajectory is determined by the atmospheric flows + terminal velocity of **individual** particles
- e.g. for fast prediction of volcanic ash dispersion
- no wet deposition
- e.g.: PUFF, VAFTAD

# RePLaT model<sup>1</sup>

## (Real Particle Lagrangian Trajectory model)

- a Lagrangian model tracking **“real” aerosol particles**
- the particles have fixed, realistic size (e.g.  $r = 1\text{--}10\ \mu\text{m}$ ) and density (e.g.  $\rho_p = 2000\ \text{kg/m}^3$ )
- equation of motion  $\leftarrow$  Newton’s equation
  - advection
  - turbulent diffusion

$\mathbf{r}_p$	particle position
$r$	particle radius
$\rho_p, \rho$	density of particle, air
$\nu$	kinematic viscosity of air
$\mathbf{v}_p, \mathbf{v}$	velocity of particle, air
$g$	gravitational acceleration
$\xi$	noise
$\mathbf{K}$	turbulent diffusion
$\mathbf{n}$	unit vector pointing upwards

$$\frac{d\mathbf{r}_p}{dt} = \mathbf{v} + w_{\text{term}}\mathbf{n} + \xi \cdot \mathbf{K}$$

$$w_{\text{term}} = -\frac{2}{9} \frac{\rho_p r^2 g}{\rho \nu} \quad \text{terminal velocity}$$

Stokes law  $\uparrow$   
(aerosol particles,  $r \lesssim 10\ \mu\text{m}$ )

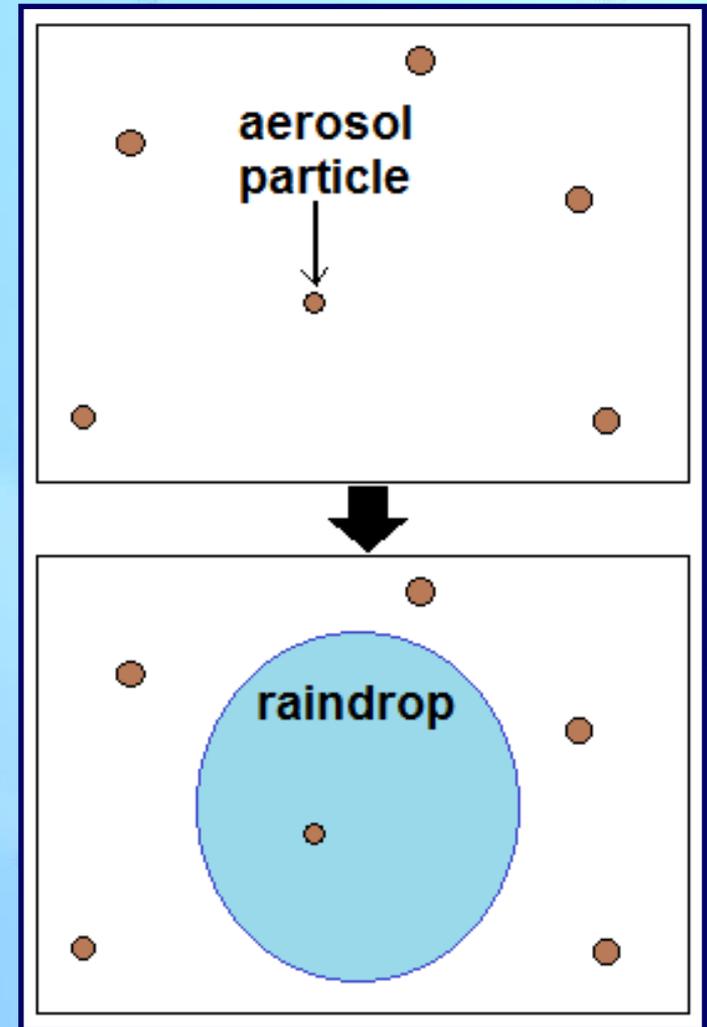
<sup>1</sup> Haszpra, T. and Tél, T. (2013) *Nonlin. Proc. Geophys.* 20(5), 867–881.

# RePLaT model

(Real Particle Lagrangian Trajectory model)

## □ wet deposition

- **random** process: a particle is captured by a raindrop with a certain  $p_{r,\text{rain}}$  *probability*
- $\rightarrow r' = r_{\text{rain}}, \rho_p' = \rho_{\text{rain}}$
- $r_{\text{rain}}$  and  $p_{r,\text{rain}}$  depend on precipitation intensity



# RePLaT model

## (Real Particle Lagrangian Trajectory model)

### □ wet deposition

based on the Eulerian approach:

$$\frac{dm}{dt} = -k_w m$$

$m$  mass

$k_w$  scavenging coefficient

$$\frac{m(\Delta t)}{m(0)} = \exp(-k_w \Delta t) \leftarrow \text{in}$$

$$1 - \frac{m(\Delta t)}{m(0)} = 1 - \exp(-k_w \Delta t) \leftarrow \text{out}$$

a particle is captured by a raindrop with probability

$$p_{r,\text{rain}} = 1 - \exp(-k_w \Delta t)$$

$k_w, r_{\text{rain}}$ : depend on  $P$  precipitation intensity

$$\rightarrow r' = r_{\text{rain}}, \rho_p' = \rho_{\text{rain}}$$

$$\rightarrow w'_{\text{term}} \gg w_{\text{term}}$$

$$w_{\text{term}} = -\frac{8}{3} \frac{\rho_{\text{rain}} r_{\text{rain}} g}{\rho C_d} \leftarrow \text{quadratic drag force}$$

# RePLaT model

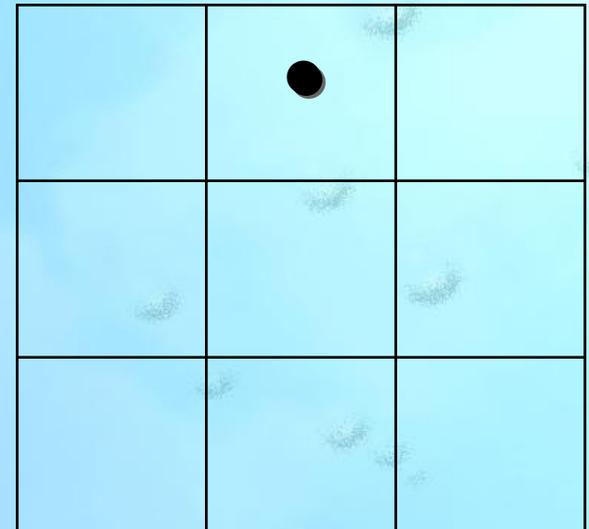
(Real Particle Lagrangian Trajectory model)

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- meteorological data in  $\lambda, \varphi, p$  coordinates  
(e.g. ERA Interim database,  
European Centre for Medium-Range Weather Forecasts)



- equation of motion:  $\lambda_p, \varphi_p, p_p$
- interpolation:
  - bicubic spline in horizontal
  - linear in time and vertical
- numerical solution: Euler method



# Equation of motion

$$\lambda_p(t + \Delta t) = \lambda_p(t) + \frac{u}{R_E \cos \varphi} \Delta t + R \sqrt{24 K_\lambda \Delta t}$$

$$\varphi_p(t + \Delta t) = \varphi_p(t) + \frac{v}{R_E} \Delta t + R \sqrt{24 K_\varphi \Delta t}$$

$$p_p(t + \Delta t) = p_p(t) + (\omega + \omega_{\text{term}}) \Delta t + R \sqrt{24 K_p \Delta t} + \frac{\partial K_p}{\partial p} \Delta t$$

advection

turbulent diffusion

$$K_\lambda = \frac{K_x}{(R_E \cos \varphi_p)^2}$$

$$K_\varphi = \frac{K_y}{R_E^2}$$

$R$  [-0.5; 0.5] uniform distribution random number

$R_E$  Earth's radius

$K_x, K_y$  constant horizontal turb. diff

$K_p \leftarrow K_z$  vertical turb. diff.

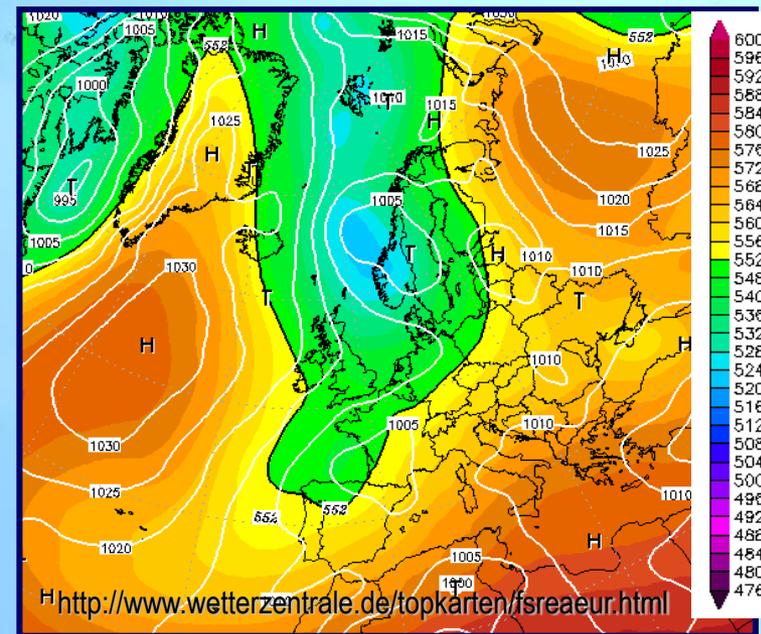
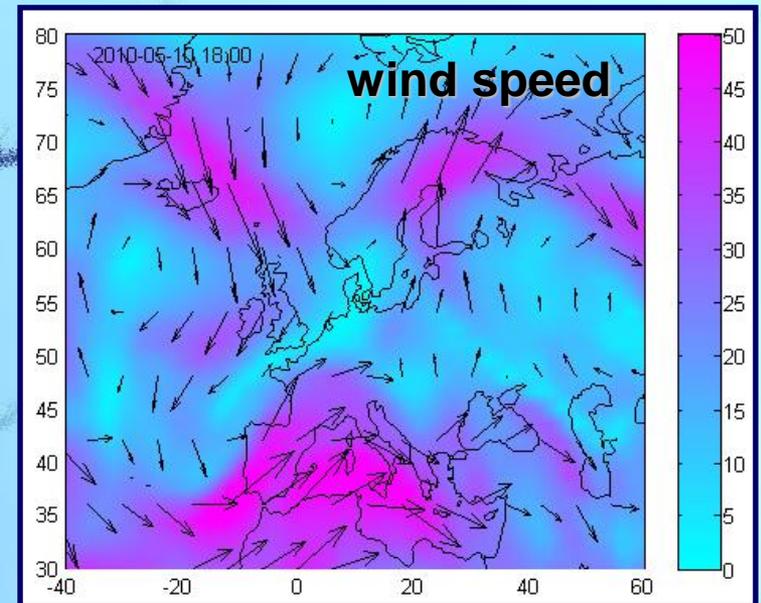
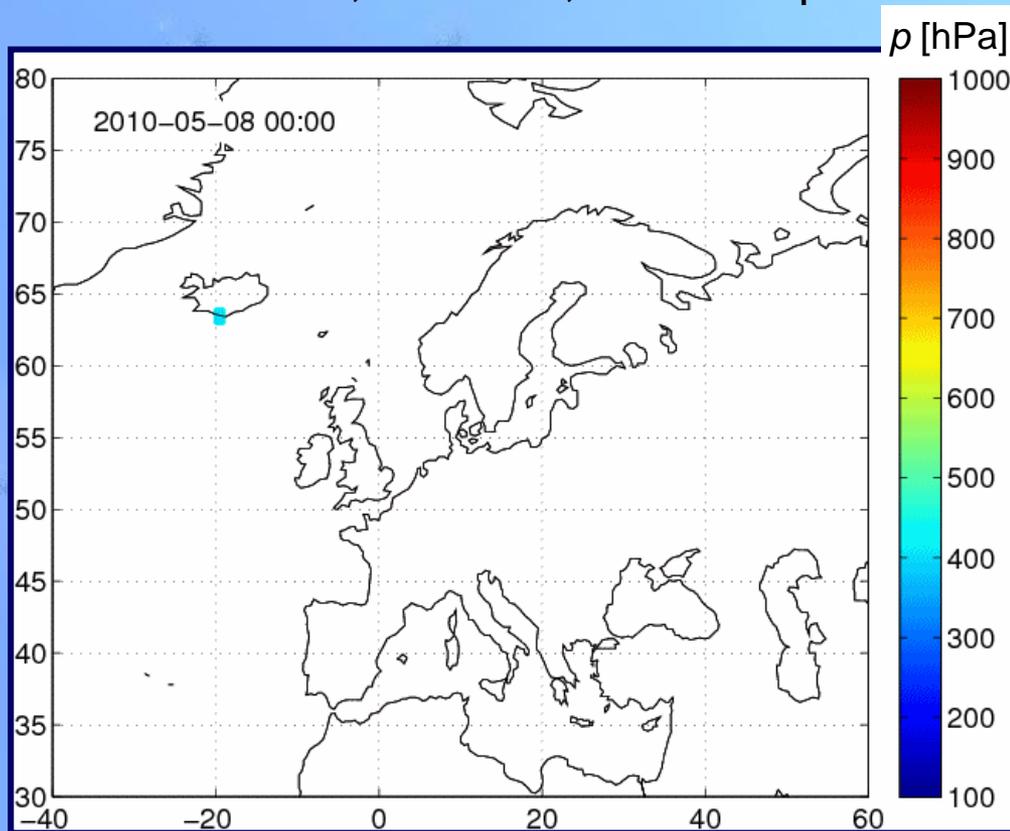
(Monin–Obukhov similarity theory)

# Eyjafjallajökull simulation (May 8–19, 2010)

$r = 1 \mu\text{m}$ ,  $\rho_p = 2000 \text{ kg/m}^3$

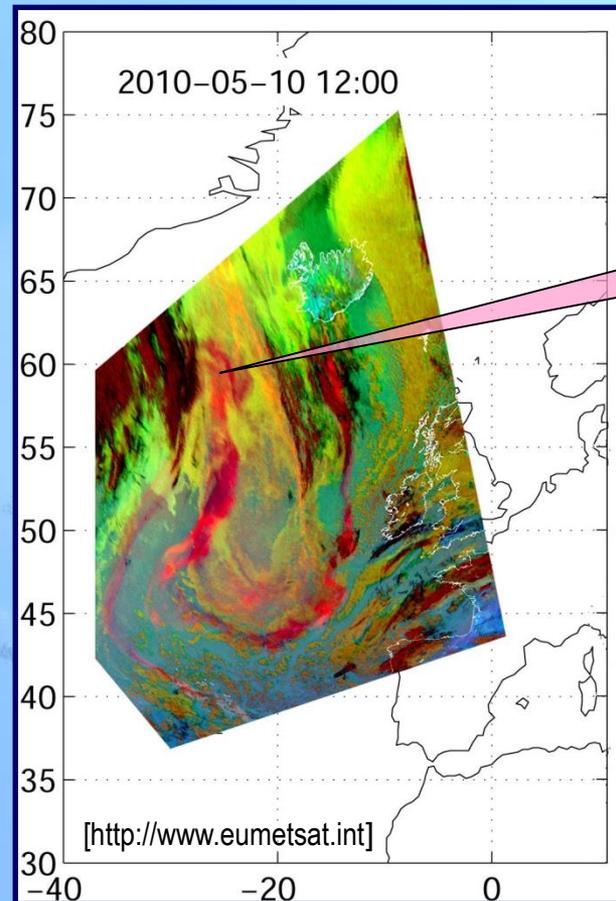
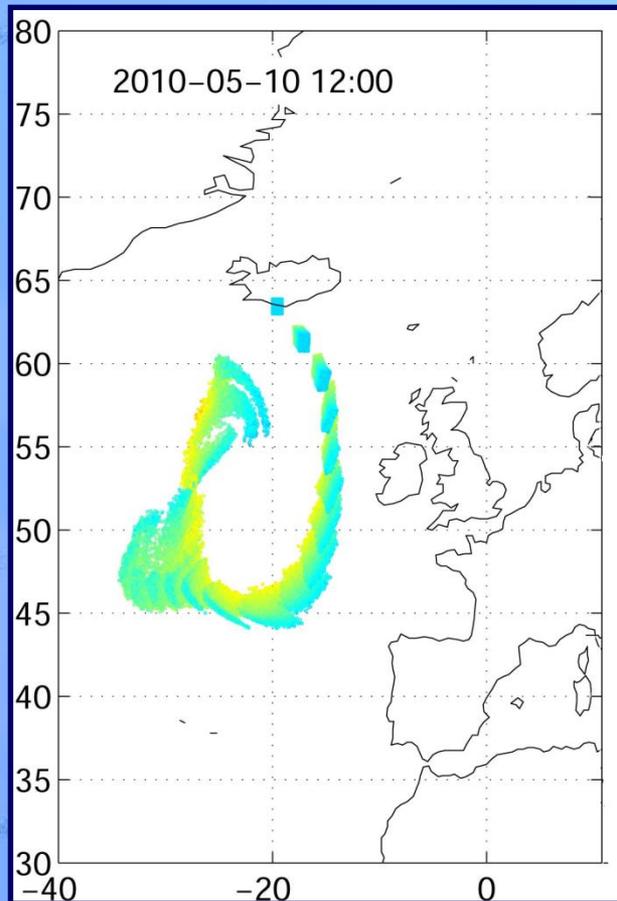
$n = 7 \cdot 10^4$  particles

simulation: adv., turb. diff., no wet dep.

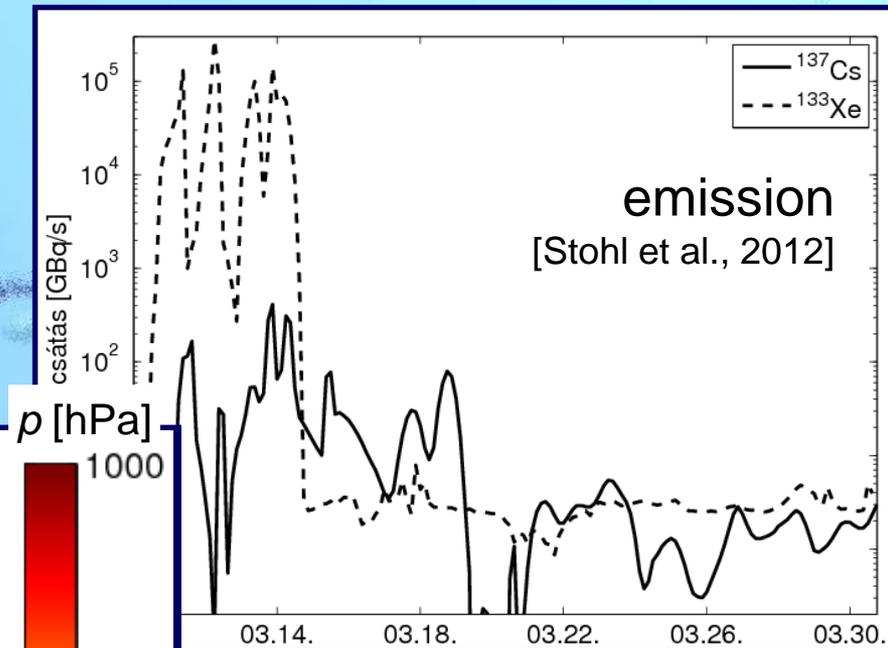
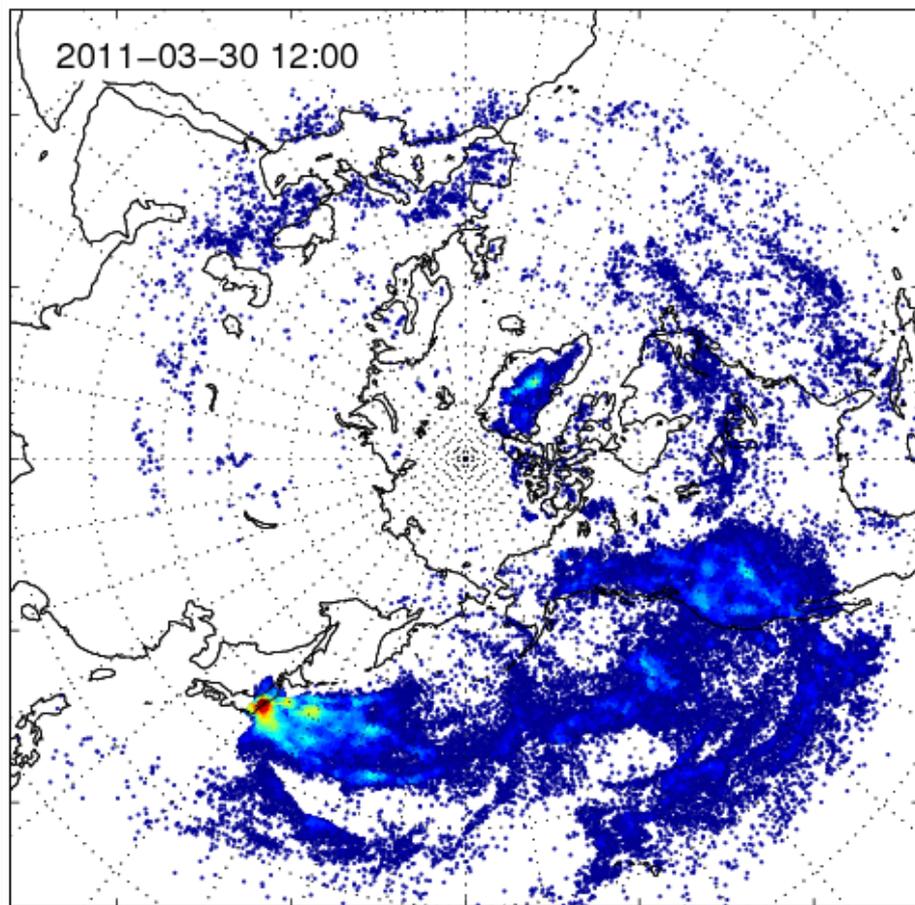


# Eyjafjallajökull simulation (May 8–19, 2010)

- *comparison: simulation and satellite measurement*



# Fukushima simulation (March 10–30, 2011)



aerosol-bound  $^{137}\text{Cs}$  isotope

$r = 0.2 \mu\text{m}$ ,  $\rho_p = 1900 \text{ kg/m}^3$   
 $n = 10^6$  particles

simulation: advection  
turb. diff.  
wet deposition

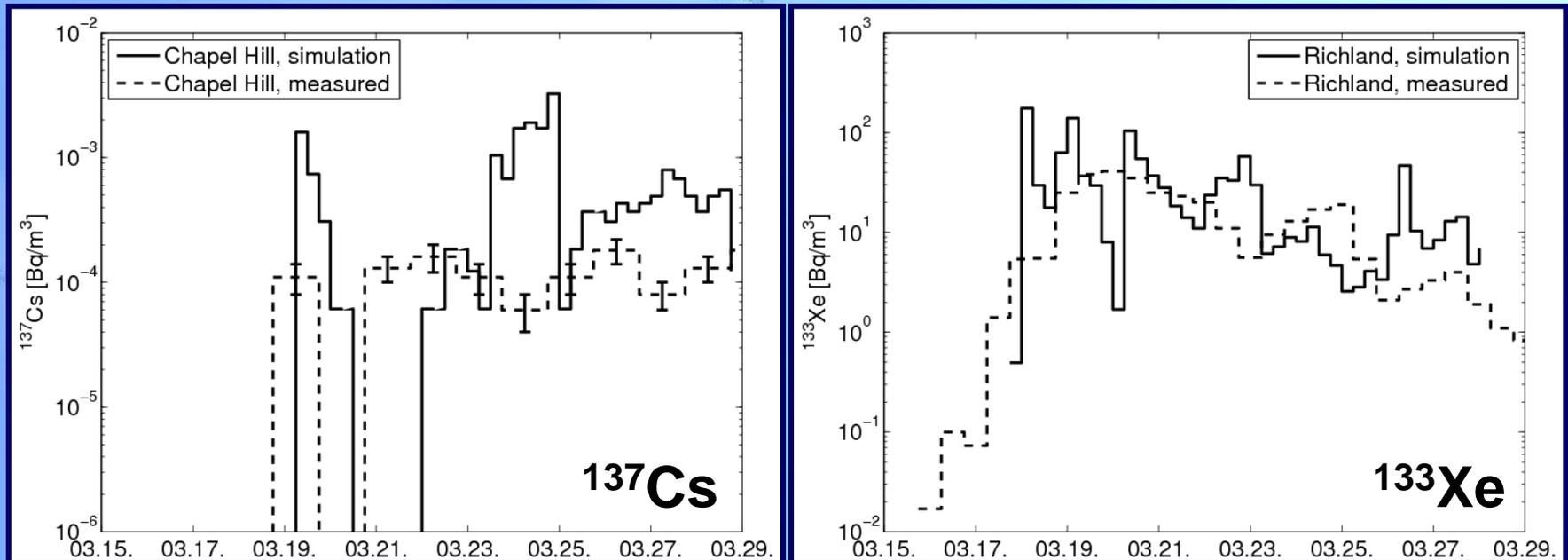
deposition field  
[kBq/m<sup>2</sup>]

# Fukushima simulation (March 10–30, 2011)

- *comparison*: measurement and simulation
- arrival times coincide reasonably well
- $^{133}\text{Xe}$ : simulations were able to reproduce the measured concentrations
- $^{137}\text{Cs}$ : sometimes overestimations

## ***uncertainties:***

- estimated emission data
- coarse resolution of the meteorological data (6h) → heavy precipitation events smoothed out
- parameterizations ...



# Uncertainties in the dispersion forecasts

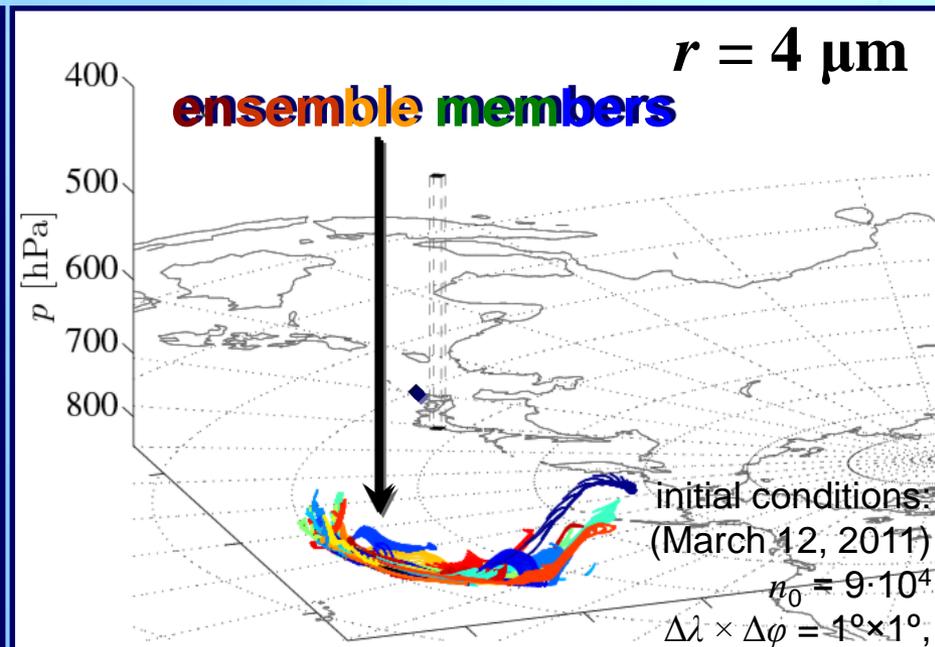
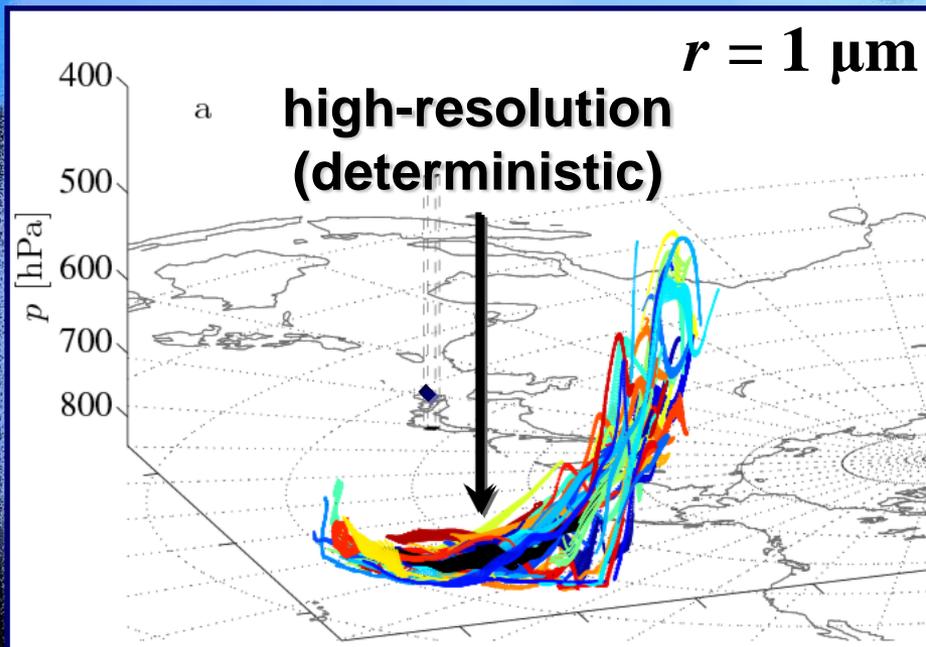
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- **input data** for the dispersion model
  - emission data
  - meteorological data → ensemble forecasts
  
- uncertainties associated to the **dispersion model**
  - processes taken into account, parameterizations
  - numerical approximations
  
- **chaotic advection** of pollutants [Aref, 1984]  
(sensitivity to the initial conditions, irregular motion, complex structures)

# Impact of the meteorological data

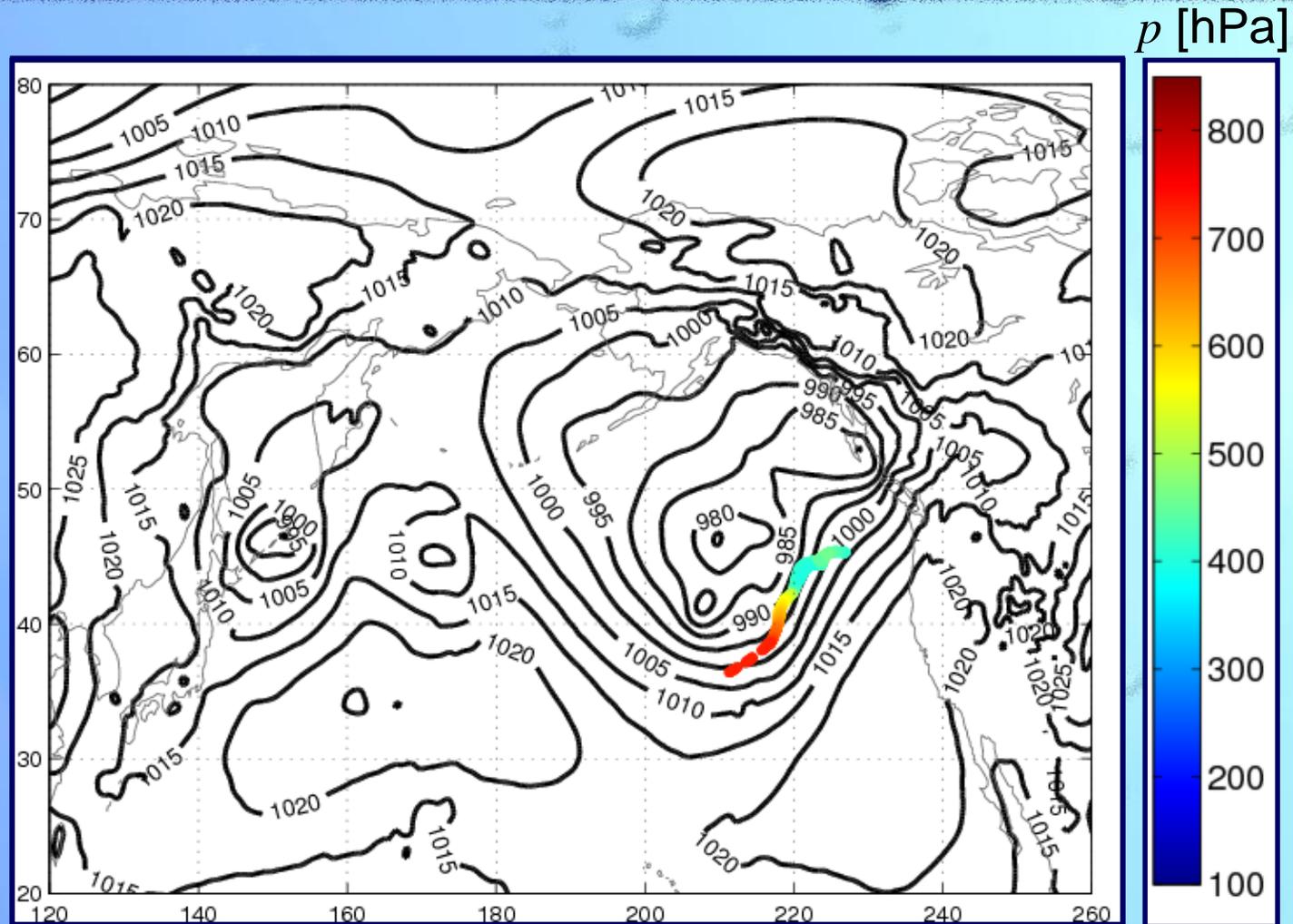
## General overview

- 50 perturbed + 1 unperturbed member (CF) + HRES
- $\rho_p = 2000 \text{ kg/m}^3$ ,  $r = 1, 2, \dots, 10 \text{ }\mu\text{m}$   
**aerosol particles**
- 3D particle distribution after 2.5 days
- **simulations:**  
advection  
no turb. diff.  
no wet dep.



# Impact of the meteorological data

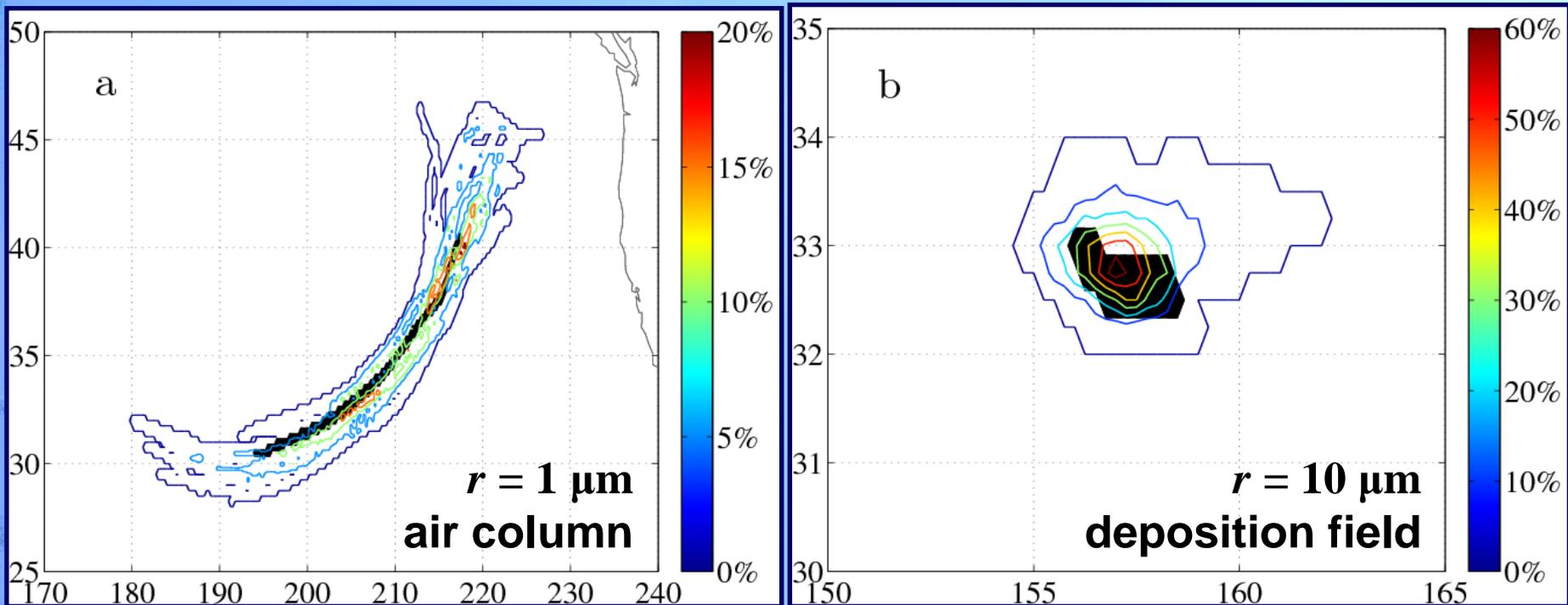
Types of pollutant clouds for  $r = 1 \mu\text{m}$



# Impact of the meteorological data

## Horizontal distribution for $r = 1$ and $10 \mu\text{m}$

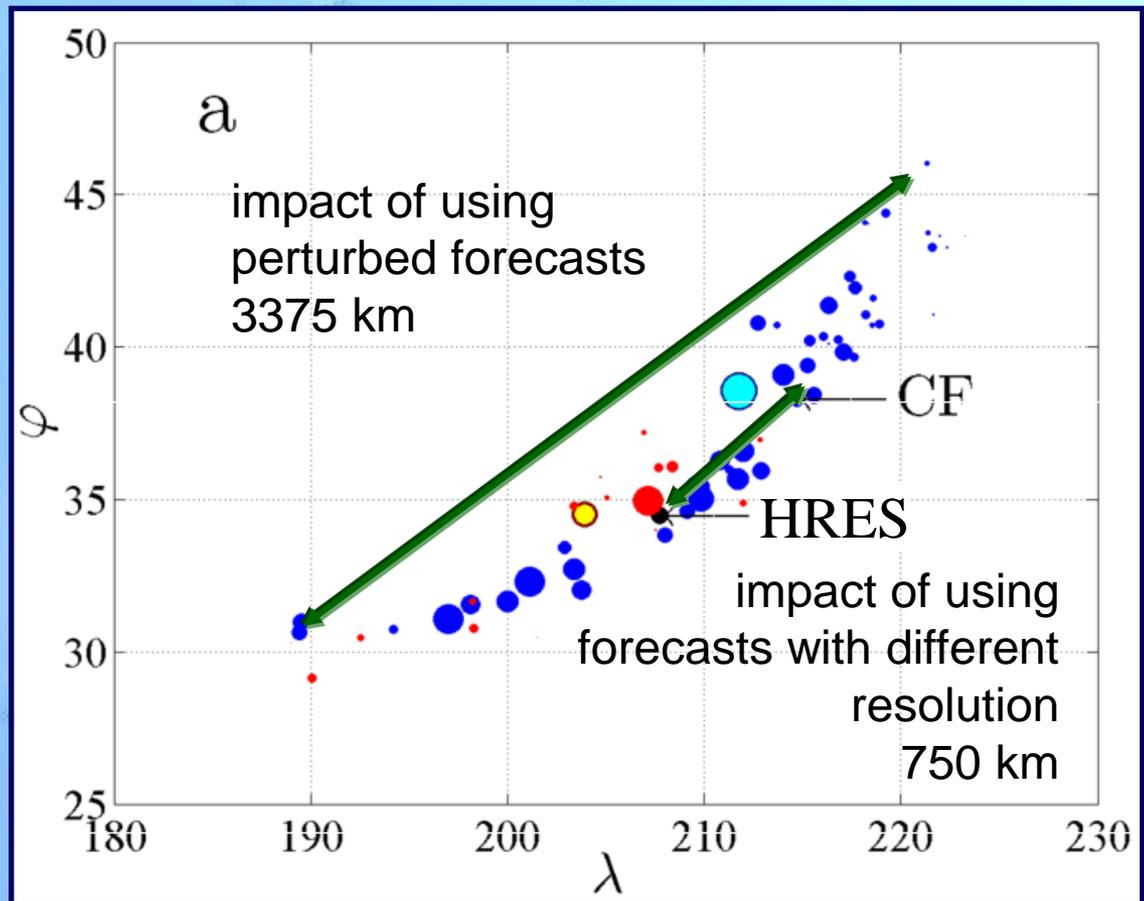
- **Colored contours** indicate the percentage of the ensemble dispersion simulations that predict a concentration above a threshold
- **Black:** the same by using the HRES forecast
- ensemble pollutant clouds expand to a 5–10 times larger area than that of the HRES forecast



# Impact of the meteorological data

## Center of mass for $r = 1 \mu\text{m}$

- for particles in ensemble dispersion members that  
**blue:** remain in air  
**red:** deposited
- radius: proportional to the standard deviation of particles around the center of mass  
**35–960 km**



Haszpra, T., Lagzi, I., Tél, T. (2013): Dispersion of aerosol particles in the free atmosphere using ensemble forecasts. *Nonlin. Proc. Geophys.* 20(5) 759–770

Haszpra, T., Horányi, A. (2014): Some aspects of the impact of meteorological forecast uncertainties on environmental dispersion prediction. *Időjárás* (accepted)

# Summary and Outlook

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- ❑ **RePLaT** Lagrangian dispersion model
- ❑ future work: RePLaT should be improved by additional factors (e.g. more detailed description of the wet deposition)
- ❑ the simulations carried out by the RePLaT model agree reasonably well with observations
- ❑ effect of **uncertainties** in the meteorological data on the dispersion calculation, and its dependence on the particle size
- ❑ ensemble pollutant clouds expand to a larger area than that of the HRES forecast
- ❑ **risk assessment** → where and when does the concentration exceed a certain threshold with what probability?
- ❑ Note: it is only one of the error sources! → it would be useful to take into account **other uncertainty sources**

**Thank you for your attention!**