



Numerical models of atmospheric composition

M.Sofiev, E.Genikhovich Finnish Meteorological Institute Main Geophysical Observatory

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- Structure of a chemical transport model (CTM)
- Could we cope with atmospheric stochasticity?
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Scales of atmospheric composition

- A specific feature of the atmospheric composition problem is a very wide range of scales, both temporal and spatial combined with very sharp gradients of the species
 - scales are largely dictated by chemical and removal lifetimes
 - > gradients are largely dictated by sources
- Gradients tend to reproduce themselves at every spatial scale, from street-canyon to global
 - consequence: at every resolution the model has to be able to deal with highly irregular field
- Non-linearities in the governing equations make averaging problematic and further complicate the scale interaction problem



Scales of atmospheric composition.2

- Global NO2 in column observed from space (SCHIAMACHY, mean July 2007)
- NO2 column over Europe (SCIAMACHY, mean July 200
- PM 2.5 observed from space Northern Italy (June 2004)
- Global CO, modelled (17 Feb 200
- NO₂ forecast, Europe (10.7.2008)
- Ozone over Central Europe, fore (8.7.2008)
- Ex.2: Primary PM 2.5 from Finnis sources, forecast (8.7.2008)
- Ex.4 NO2 for Lisbon (mean 2001 2002)





Basic equations





Basic equations.2

K-theory:

$$\overline{u_i'\varphi'} \approx -\mu_{ij} \frac{\partial\varphi}{\partial x_j}$$

$$L\varphi \equiv \frac{\partial \varphi}{\partial t} + \frac{\partial}{\partial x_i} (u_i \varphi) - \frac{\partial}{\partial x_i} \mu_i \frac{\partial \varphi}{\partial x_i} + \sigma \varphi = f$$

Boundary & initial conditions:

$$\varphi(t=0) = \varphi_0$$
$$\varphi(\vec{r} \in S) = \varphi_S$$



Basic equations.3

Forward problem:

$$L = \frac{\partial}{\partial t} + \frac{\partial}{\partial x_i} (u_i) - \frac{\partial}{\partial x_i} \mu_i \frac{\partial}{\partial x_i} + \sigma; \quad L\varphi = f; \ M = (p, \varphi)$$

Inverse (adjoint) problem

$$L^{*} = -\frac{\partial}{\partial t} - \frac{\partial}{\partial x_{i}}(u_{i}) - \frac{\partial}{\partial x_{i}}\mu_{i}\frac{\partial}{\partial x_{i}} + \sigma; \quad L^{*}\varphi^{*} = p; \ M = (f,\varphi^{*})$$

advection diffusion chemistry receptor emission
removal sensitivity

Numerical algorithms: split



- Why not to discretise and solve directly ?
- Formal operator split

$$\frac{\partial \varphi}{\partial t} + L\varphi = 0; \quad L = L_1 + L_2$$

• Physical processes -based split

$$\frac{\partial \varphi}{\partial t} + \frac{\partial}{\partial x_i} (u_i \varphi) = 0 \quad transport, \, mass - concervative$$
$$\frac{\partial \varphi}{\partial t} = \frac{\partial}{\partial x_i} \mu_i \frac{\partial \varphi}{\partial x_i} - \sigma \varphi - f \quad diffusion, \, sink, \, emission$$



Numerical algorithms: split.2

- Physical processes split
 - LOCALLY independent, additive processes
 - > Symmetrization of the algorithms within a single time step τ

 τ_2 advection $\rightarrow \tau_2$ diffusion $\rightarrow \tau_2$ diffusion $\rightarrow \tau_2$ advection



Numerical algorithms: discretization

1D case:
$$\frac{\partial \varphi}{\partial t} + u \frac{\partial \varphi}{\partial x} = 0; \quad let \ \varphi_k^j = \varphi(x_k, t_j)$$

• Explicit scheme

$$\frac{\varphi_k^{j+1} - \varphi_k^j}{\tau} + u \frac{\varphi_k^j - \varphi_{k-1}^j}{\Delta x} = 0$$

• Implicit scheme

$$\frac{\varphi_k^{j+1} - \varphi_k^j}{\tau} + u \frac{\varphi_k^{j+1} - \varphi_{k-1}^{j+1}}{\Delta x} = 0$$



Structure of a dispersion model

- Input data pre-processors
 - > emission
 - meteorology
 - > physiography (domain properties)
- Dynamic emission (simulated vs imported)
- Advection scheme
- Diffusion module
- Chemical transformation module
- Aerosol dynamics module
- Dry and wet deposition module
- Diagnostic quantities
- Output post-processing

Input data pre-processors

- Emission
 - > various source types (point, area, stack...)
 - time variation (diurnal, weekly, seasonal)
 - chemical content (time-dependent)
- Meteorology
 - create extra variables (e.g. ABL parameters)
 - interpolation to the model grid
 - time interpolation

Advection scheme



- There is no ideal scheme
- Scheme type depends on particular task
 - Eulerian schemes are the only ones applicable to non-linear case
 - often suffer from numerical viscosity
 - Large point sources are easier to treat by Lagrangian schemes
 - Problem of representativeness of a single Lagrangian particle



Advection scheme: numerical viscosity



Mass conservation



- Mass conservation is ensured by the continuity equation
- Meteorological and dispersion models may have different:
 - ➤ grids
 - vertical structures
 - advection schemes
 - time dimension
- Any non-linear grid transformation destroys the continuity equation
 - \Rightarrow interpolated fields are divergent
 - \Rightarrow dispersion model does not conserve mass



Mass conservation (3)



- > local mass conservation: continuity equation
- > global mass conservation: globally integrated continuity equation





Advection scheme: numerical viscosity.2

1D case.
$$\frac{\partial \varphi}{\partial t} + u \frac{\partial \varphi}{\partial x} = K \frac{\partial^2 \varphi}{\partial x^2}; \quad let \quad \varphi_k^j = \varphi(x_k, t_j)$$

Tailor series for $\varphi(x,t)$ near $x=x_k$, $t=t_j$:

$$\varphi(x,t) = \varphi_k^j + (\varphi_t)_k^j (t - t_j) + (\varphi_x)_k^j (x - x_k) + \dots$$

substituting discrete equation, we get:

$$\frac{\partial \varphi}{\partial t} + \frac{\partial}{\partial x} (u_x \varphi) = (K - 0.5u\Delta x) \frac{\partial^2 \varphi}{\partial x^2} \quad at \ x = x_k, \ t = t_j$$

$$u < K / \Delta x - stable \ with \ numerical \ viscosity$$

$$u > K / \Delta x - unstable \ (implicit \ scheme !)$$



SCD advection: some examples

SCD scheme:







Diffusion module



- Straightforward discretization of 1D diffusion equation with e.g. Crank-Nicolson semi-implicit 2d-order accuracy scheme
 - > suitable for both vertical and horizontal diffusion
- 2.5-D with well-mixed boundary layer (somewhat old-fashioned)
 - Additional equation for the mixing height
- Semi-analytic vertical profile of concentrations (2.5-D as well)



Chemical scheme

- One of the most time-consuming modules
- Contains of the most severe non-linearities, also the stiffest subsystem (several orders of magnitude of reaction time scales)
- Chemical kinetics

$$A + B \rightarrow C: \quad \frac{d[C]}{dt} = K[A][B]$$



Aerosol dynamics

- Based on solution of integro-differential equation describing at least
 - Nucleation
 - Condensation
 - Coagulation
- New dimension !!
 - particle size
- The most time-consuming module

Dry and wet deposition



- Dry deposition
 - > linear (well, sometimes)
 - surface process
 - moderate intensity
 - > can be bi-directional (evaporation \Rightarrow re-emission)
 - approached via e.g. resistive analogy (Wessely, 1989)
 - aerodynamic resistance
 - laminar-layer resistance
 - surface resistances: soil, canopy, water surface, …
 - sedimentation
 - detailed landuse needed

- Wet deposition
 - can be non-linear volume process
 - high intensity
 - high complexity and dependence on precipitation and species features =>
 - usually treated via "empirical" 1storder equation:

$$\frac{\partial \varphi}{\partial t} = \Lambda(I, ..., \varphi)\varphi$$

where *I* is a precipitation intensity

Diagnostic and output post-processing



- Computation of diagnostic variables
 - > e.g. optical features of the atmosphere from concentrations
 - proxies for health impact and risk assessment
- Transformation from model-convenient variables to userfriendly ones
 - generation of integrated / averaged variables
- Conversion to convenient file formats
- Grid interpolation (if needed)



Can we cope with the stochasticity of the atmosphere?





An example of the SILAM CTM and its applications

Regional AQ forecasting platform (example)





SILAM modelling system: main parts



Current model functionality



- Dispersion
 - Forward (compute concentrations from given emission sources)
 - inverse (find sources of observed concentrations: source apportionment)
 - Lagrangian and Eulerian advection schemes
- Chemistry and physics
 - SOx-NOx-NHx-O3 chemistry
 - Linear SOx transformations
 - Radioactive decay for up to ~500 nuclides
 - Natural bioaerosols (pollen)
 - Sea salt production and dispersion
 - Inert aerosol (for general particulates of given size distribution)
 - Toxic persistent pollutants (generalised)
 - Passive tracer (probabilistic computations)
- Aerosol representation: sections or modes, arbitrary number and characteristics of size classes

Example of global run

- CO concentrations, February 2001
 - (experimental SILAM application)





Forecast for pm2_5 from forest fires. Last analysis time: 24.4 15.5.2006





Forecast for pm2_5 from forest fires, FRP-based. Last actual FRP map: 2008 7 6 0 0 0.0 UTC



ETEX-1 plume evolution (forward problem)



ETEX-1 inverse problem via adjoint simulation (4

True source:

(2°W, 48.05°N)

Release time: 23.10.1994 16:00 ->

24.10.1994 3:50 (duration ~12 hours)



Summary



- Modelling of distribution of atmospheric trace components is based on solving the turbulent diffusion equation
- Particular realization and corresponding simplifications depend on specific task and available resources
- The technology of computation and corresponding results should be "adequate" to the problem under consideration
- Model quality assurance should cover ALL stages of the model development
- When carrying out CTM, one should be aware of the stochastic nature of the modeled atmospheric processes, expected magnitude of fluctuation of concentration etc
 - > To obtain smart answer one has to ask smart question...



Thank you for your attention