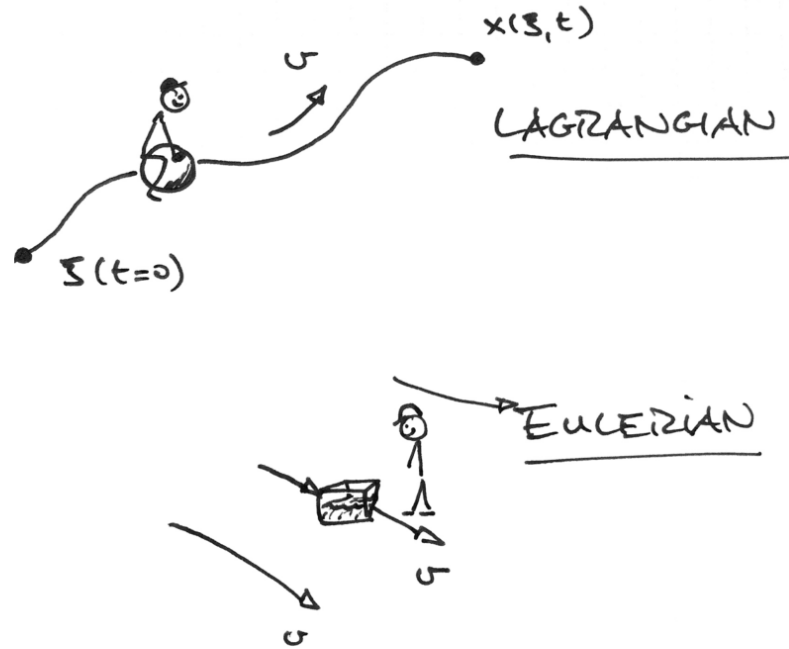


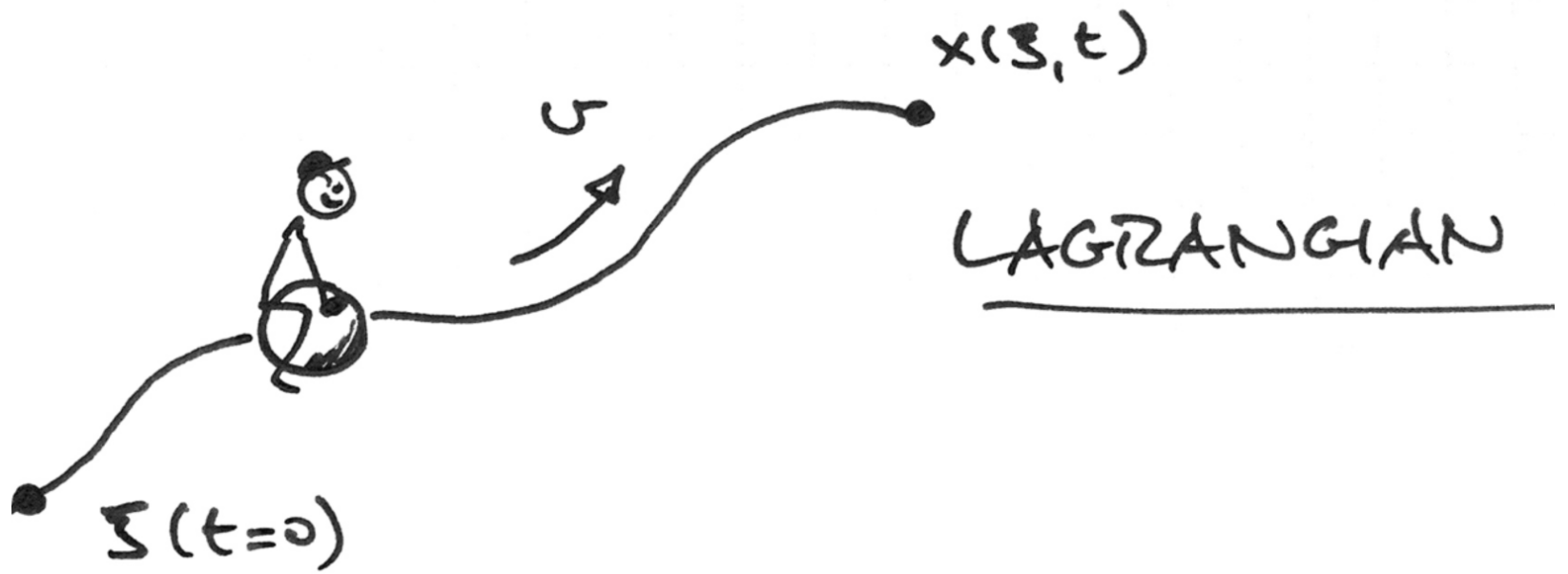
Differential Equations II



Numerical Solution of the Transport Equation:
Lagrangian and Semi-Lagrangian Methods

Lagrangian Perspective





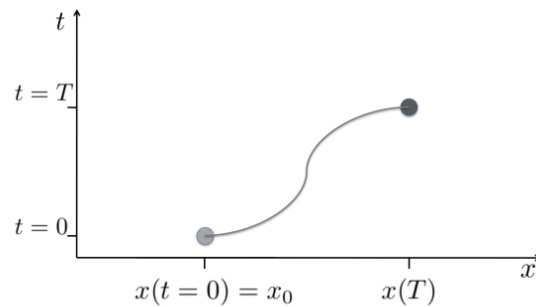
Formalization

- Position: $x = x(t)$.
- Velocity: $v = v(x, t)$.

Particle position can be computed by

$$\dot{x} = \frac{dx}{dt} = v(x, t)$$

With initial condition $x(t = 0) = x_0$



Lagrangian Transport with Source

The equation is given by

$$\frac{d\rho}{dt} = s(x, t)$$

where $\frac{d}{dt} = \frac{\partial}{\partial t} + \frac{\partial}{\partial x} \cdot \frac{dx}{dt}$

Remark: With $\dot{x} = v(x, t)$ we have that

$$\frac{d}{dt} = \rho_t + v\rho_x = s(x, t).$$

Summary: We need to solve two ODEs:

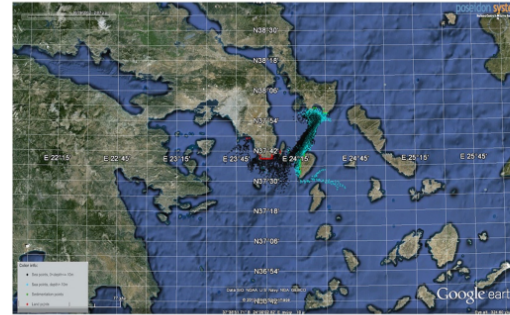
$$\begin{aligned}\dot{x} &= v(x, t), & x(0) &= x_0, \\ \dot{\rho} &= s(\rho, x, t), & \rho(x, 0) &= \rho_0(x).\end{aligned}$$

Remark: Usually many particles are used for real applications!

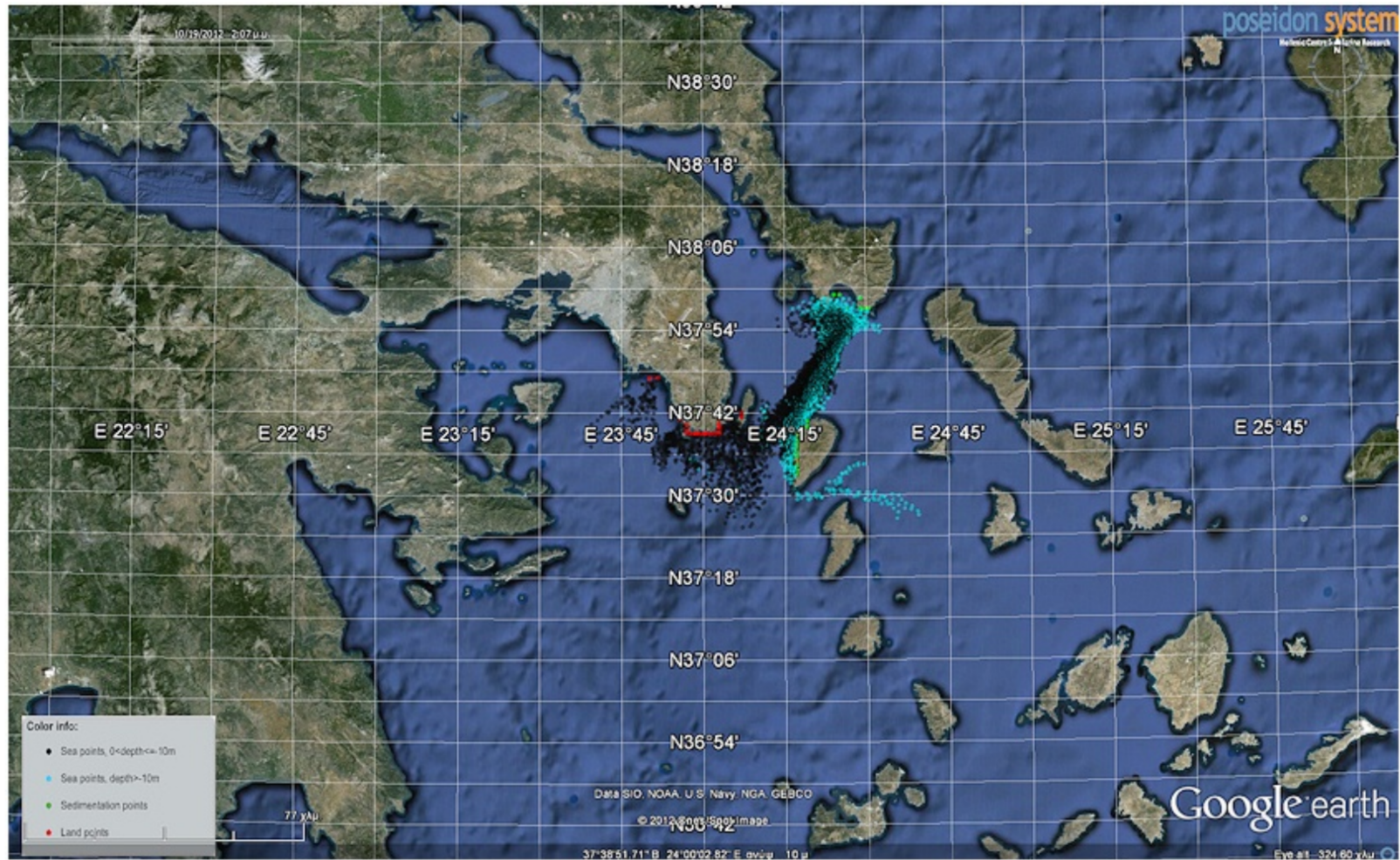
Remark: Homogeneous advection, where $s \equiv 0$, yields:

$$\dot{\rho} = 0 \quad \Rightarrow \quad \rho(\cdot, t) \equiv \text{const.} = \rho_0(\cdot).$$

Since $x = x(t)$, the particle position is implied: $\rho(x, t) = \rho(x(t), t)$.



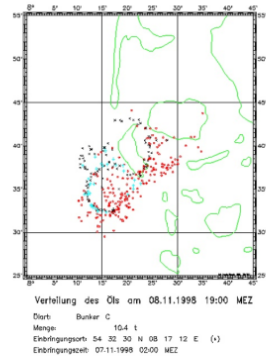
<http://www.medess4ms.eu/oil-spill-models>

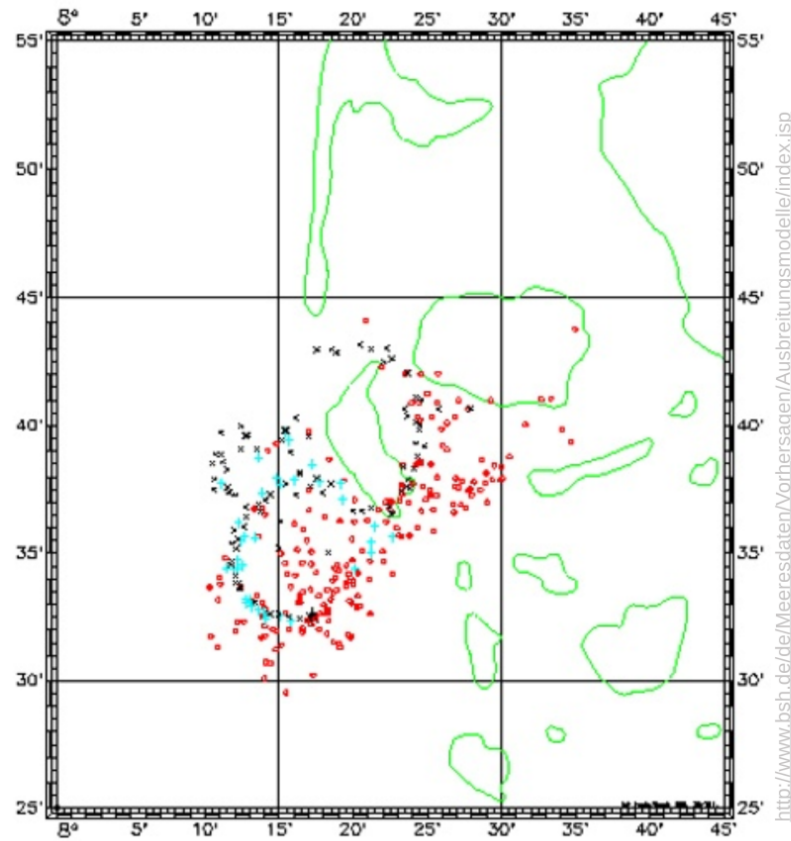


<http://www.medess4ms.eu/oil-spill-models>

Problems of purely Lagrangian Methods

- Distribution of particles will eventually become very irregular.
- Interaction between particles is difficult to simulate (diffusive processes).
- Density distribution fields with spatial coverage hard to reconstruct.





Verteilung des Öls am 08.11.1998 19:00 MEZ

Ölart: Bunker C

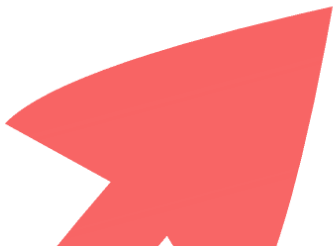
Menge: 10.4 t

Einbringungsort: 54 32 30 N 08 17 12 E (•)

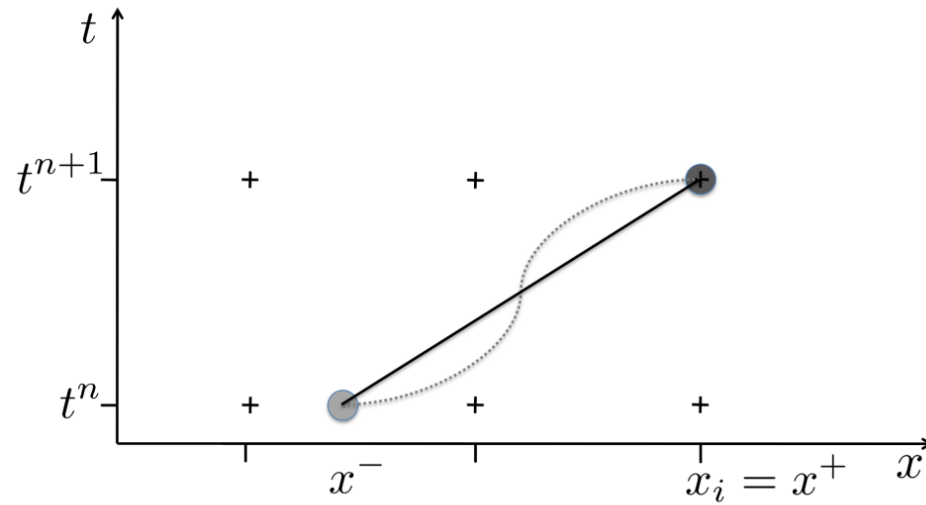
Einbringungszeit: 07-11-1998 02:00 MEZ

Idea

Combine Lagrangian and Eulerian Methods

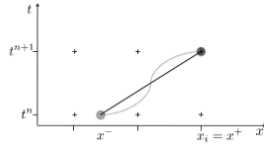


Idea of the Semi-Lagrangian Method



Formalization

Problem: Passive advection ($s \equiv 0$):



$$\begin{aligned}\frac{dx}{dt} &= v(x, t), & x(0) &= x_0, \\ \frac{d\rho}{dt} &= 0, & \rho(x, 0) &= \rho_0(x).\end{aligned}$$

Strategy:

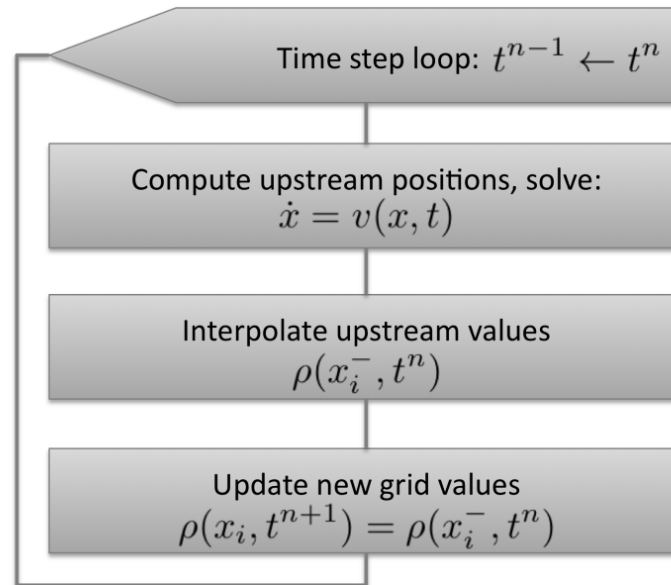
- Solve $\frac{dx}{dt} = v$ by any ODE solver,
- Solve $\frac{d\rho}{dt} = 0$ by finite difference.

$$\frac{d\rho}{dt} \approx \frac{\rho(x_i, t^{n+1}) - \rho(x_i^-, t^n)}{\Delta t} = 0$$

$$\Rightarrow \rho^+ = \rho^-.$$

$x_i, i = 1 : N$ grid points, $t^n, n = 1 : M$ time steps.

Algorithm



Stability and Consistency

Von Neumann Stability Analysis:

Assume linear interpolation of upstream points and exact wind, i.e.

$$\rho_i^{n+1} = (1 - \nu)\rho_{k_i}^n + \nu\rho_{k_i-1}^n$$

- $[x_{k_{i-1}}, x_{k_i}]$ interval containing upstream point x_i^- ,
- $\nu = \frac{v_k - v^-}{\Delta x}$.

Then using $z_n e^{ik(jh)}$ we have:

$$\begin{aligned} z_{n+1} e^{ik(jh)} &= (1 - \nu)z_n e^{ik(k_i h)} + \nu z_n e^{ik(k_i - 1)h}, \\ &= z_n \left[1 - \nu(1 - e^{-ik(h)}) \right] e^{ik(k_i - j)h} e^{ik(jh)}; \\ \Rightarrow \xi &= \left[1 - \nu(1 - e^{-ik(h)}) \right] e^{ik(k_i - j)h}; \\ \Rightarrow |\xi|^2 &= 1 - 2\nu(1 - \nu)(1 - \cos(kh)). \end{aligned}$$

Stability follows for: $0 \leq \nu \leq 1$, i.e. always!

Remark (Order of Consistency):

It can be shown the the semi-Lagrangian advection scheme retains the consistency order of the discretization schemes involved:

$$\left. \begin{array}{l} \dot{x} = v \text{ order } p \\ \dot{\rho} = 0 \text{ order } p \end{array} \right\} \Rightarrow \text{SLM order } p.$$

Stability and Consistency

Von Neumann Stability Analysis:

Assume linear interpolation of upstream points and exact wind, i.e.

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- $[x_{k_i-1}, x_{k_i}]$ interval contain
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• ν wind, i.e.

- $[x_{k_{i-1}}, x_{k_i}]$ interval containing upstream point x_i^- ,
- $\nu = \frac{x_k - x^-}{\Delta x}$.

Then using $z_n e^{ik(jh)}$ we have:

$$\begin{aligned} z_{n+1} e^{ik(jh)} &= (1 - \nu) z_n e^{ik(k_i h)} + \nu z_n e^{ik(k_i - 1)h}, \\ &= z_n \left[1 - \nu(1 - e^{-ik(h)}) \right] e^{ik(k_i - j)h} e^{ik(jh)}; \\ \Rightarrow \xi &= \left[1 - \nu(1 - e^{-ik(h)}) \right] e^{ik(k_i - j)h}; \\ \Rightarrow |\xi|^2 &= 1 - 2\nu(1 - \nu)(1 - \cos(kh)). \end{aligned}$$

$$\begin{aligned}
z_{n+1} &= (1 - \nu)z_n \\
&= z_n \left[1 - \nu(1 - e^{-\nu}) \right] \\
\Rightarrow \xi &= \left[1 - \nu(1 - e^{-\nu}) \right] \\
\Rightarrow |\xi|^2 &= 1 - 2\nu(1 - \nu)(\dots)
\end{aligned}$$

Stability follows for: $0 \leq \nu \leq 1$, i.e. always!

Remark (Order of Consistency):

It can be shown that the semi-Lagrangian scheme has a consistency order of 2. The consistency order of the discretization is 2.

for: $0 \leq \nu \leq 1$, i.e. always!

Remark (Order of Consistency):

It can be shown that the semi-Lagrangian advection scheme retains the consistency order of the discretization schemes involved:

$$\left. \begin{array}{l} \dot{x} = v \text{ order } p \\ \dot{\rho} = 0 \text{ order } p \end{array} \right\} \Rightarrow \text{SLM order } p.$$

Advection with Source

Recall:

$$\rho_t + v\rho_x = s.$$

Lagrangian:

$$\dot{x} = v; \dot{\rho} = s.$$

For one particle we solve:

$$\begin{aligned}\dot{x} &= v(x, t), & x(0) &= x_0, \\ \dot{\rho} &= s(\rho, x, t), & \rho(x, 0) &= \rho_0(x).\end{aligned}$$

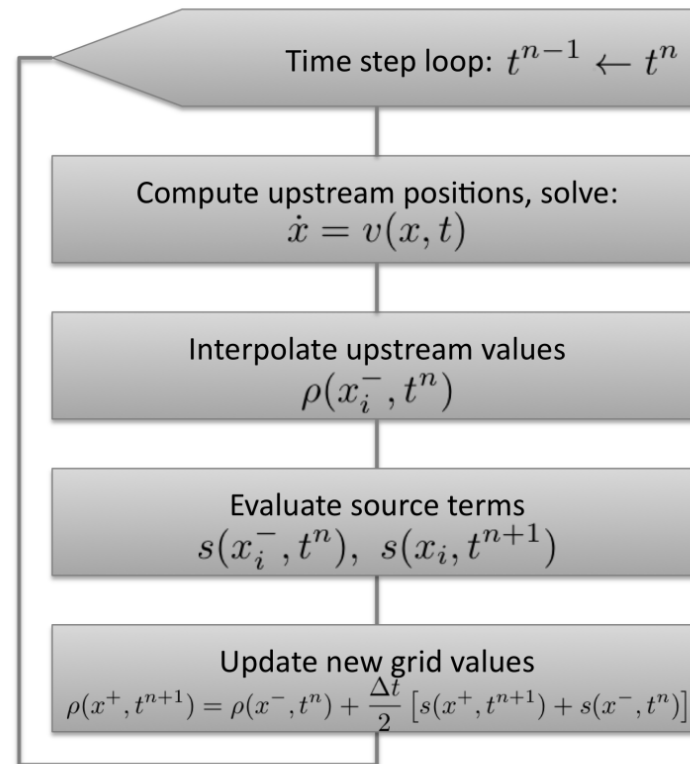
Assume $s = s(x, t)$, use trapezoidal rule:

$$\frac{\rho(x^+, t^{n+1}) - \rho(x^-, t^n)}{\Delta t} = \frac{1}{2} [s(x^+, t^{n+1}) + s(x^-, t^n)]$$

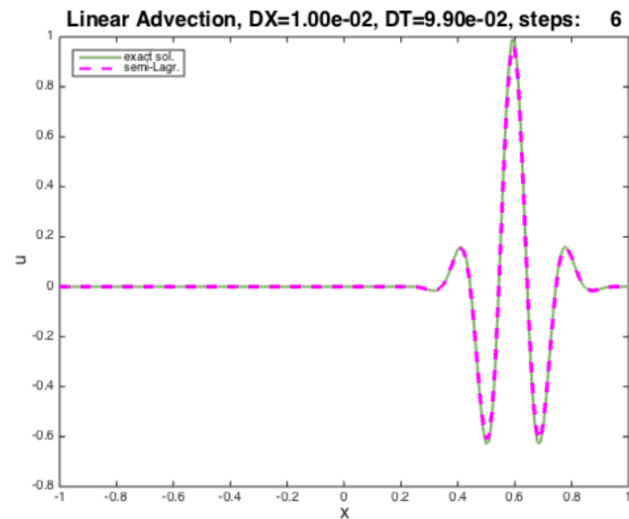
Or even simpler:

$$\frac{\rho(x^+, t^{n+1}) - \rho(x^-, t^n)}{\Delta t} = s(x^0, t^{n+1/2})$$

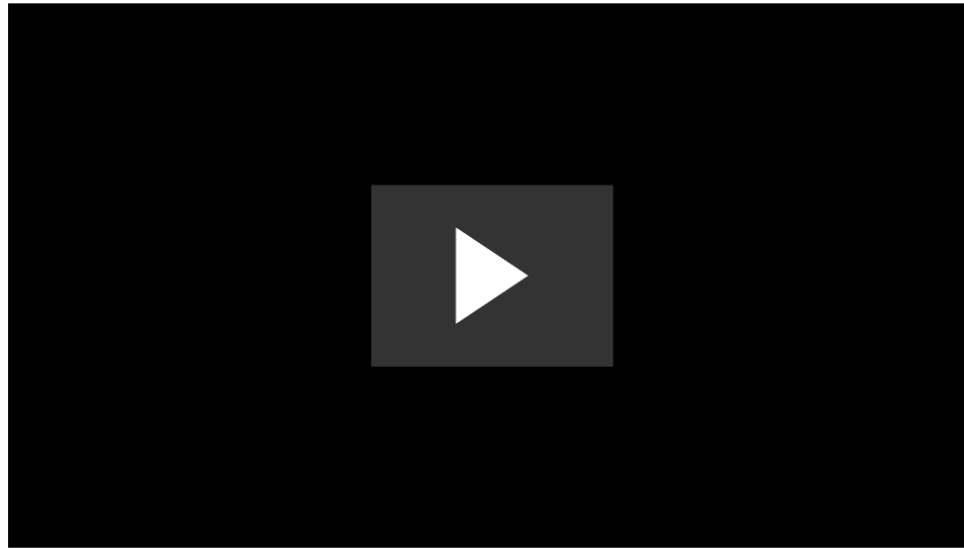
Algorithm with Source



Semi-Lagrangian Algorithm Result



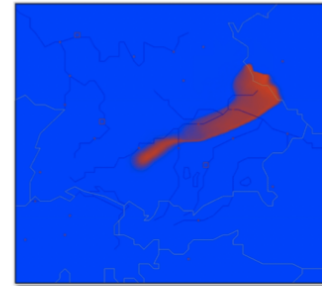
Now 2D



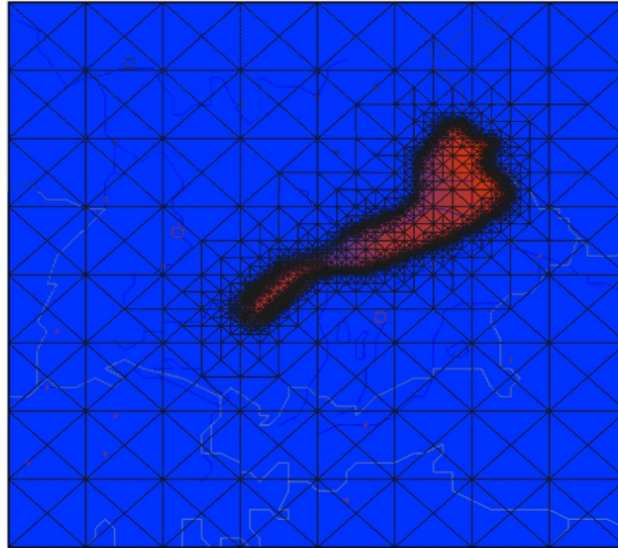
Plume dispersion as passive tracer?

Multi-Scale problem:

Total Extent $\mathcal{O}(10^5 m^2)$
Local concentrations $\mathcal{O}(10^2 m^2)$



Idea: adaptive mesh refinement methods



- refinement only where necessary
- dynamically adaptive during run-time

Advection-Diffusion equation

$$\frac{\partial \rho}{\partial t} + \nabla \cdot (\mathbf{v}\rho) + \nabla \cdot (\mu \nabla \rho) = 0$$

Constituent (possibly multi-component)

$$\rho : \Omega \times T \rightarrow \mathbb{R}^m, \quad \Omega \subset \mathbb{R}^d$$

Given wind field

$$\mathbf{v} : \Omega \times T \rightarrow \mathbb{R}^d$$

Given diffusion coefficient

$$\mu : \Omega \times T \rightarrow \mathbb{R}$$

Lagrangian Form

$$\begin{aligned}\frac{\partial \rho}{\partial t} + \nabla \cdot (\mathbf{v} \rho) &= \frac{\partial \rho}{\partial t} + \mathbf{v} \cdot \nabla \rho + \nabla \cdot \mathbf{v} \rho \\ &= \frac{d\rho}{dt} + \nabla \cdot \mathbf{v} \rho.\end{aligned}$$

Divergence-free flow: $\nabla \cdot \mathbf{v} = 0$

$$\frac{d\rho}{dt} = 0$$

Splitting advection and diffusion

$$\frac{d\rho}{dt} + \nabla \cdot (\mu \nabla \rho) = 0$$

Advection



Diffusion



Time integration over one time step:

$$\int_t^{t+\Delta t} \frac{d\rho}{dt} + \nabla \cdot (\mu \nabla \rho) dt = 0$$

Advection Term (exact integration):

$$\int_t^{t+\Delta t} \frac{d\rho}{dt} dt = \rho(x, t + \Delta t) - \rho(x - \alpha, t)$$

Diffusion Term (trapezoidal rule):

$$\int_t^{t+\Delta t} \nabla \cdot (\mu \nabla \rho) dt \approx \frac{1}{2} [\nabla \cdot (\mu^+ \nabla \rho^+) + \nabla \cdot (\mu^- \nabla \rho^-)] \cdot \Delta t$$

Discretization of Advection-Diffusion eq.

$$\rho^+ = \rho^- - \frac{\Delta t}{2} [\nabla \cdot (\mu^+ \nabla \rho^+) + \nabla \cdot (\mu^- \nabla \rho^-)] = 0$$

Problem with non-uniform Meshes

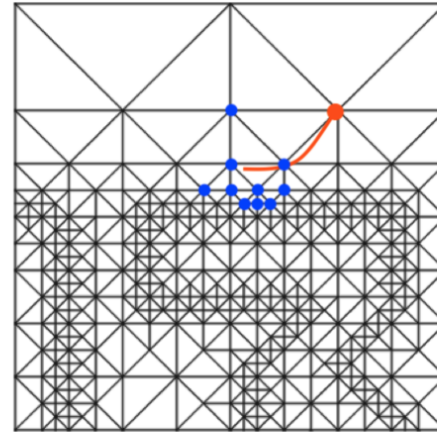
Semi-Lagrangian Method:

Lagrange Form: $\frac{dc}{dt} = 0$

Difference along trajectory:

$$\frac{dc}{dt} \approx \frac{c(\vec{x}, t + \Delta t) - c(\vec{x} - 2\vec{\alpha}, t - \Delta t)}{2\Delta t}$$

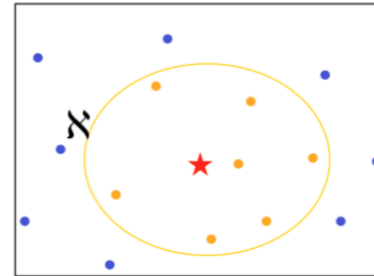
$$\Rightarrow c(\vec{x}, t + \Delta t) = c(\vec{x} - 2\vec{\alpha}, t - \Delta t)$$



Interpolation with Radial Basis Functions

Interpolation Problem:

$$s : \mathbb{R}^d \rightarrow \mathbb{R} \quad \text{with} \quad s|_{\mathfrak{N}} = c(\cdot, t)|_{\mathfrak{N}}.$$



\mathfrak{N} set of k neighbors

Interpolating Function:

$$s(x) = \sum_{j=1}^k \lambda_j \Phi(\|x - y_j\|) + p(x), \quad y_j \in \mathfrak{N}, \quad p(x) = \sum_{l=1}^q \mu_l p_l(x).$$

$$\begin{bmatrix} A_{\Phi, \mathfrak{N}} & P \\ P & 0 \end{bmatrix} \begin{bmatrix} \lambda \\ \mu \end{bmatrix} = \begin{bmatrix} c(y_j) \\ 0 \end{bmatrix}$$

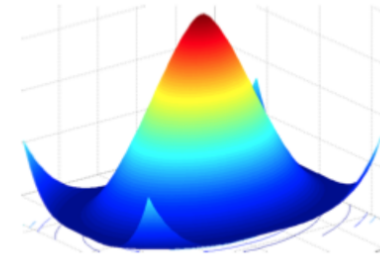
$$A_{\Phi, \mathfrak{N}} = [\Phi(\|x_j - x_l\|)]_{j, l=1:k}, \quad P = [p_n(x_j)]_{n=1:q, j=1:k}.$$

(\mathfrak{N} non-degenerate)

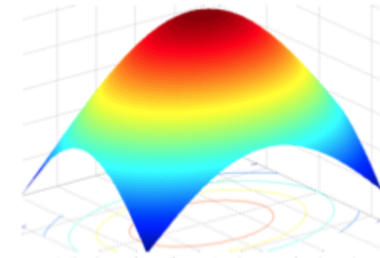
R.L. Hardy (1971), J. Duchon (1976), W.R. Madych/S.A. Nelson (1988)

Examples of Radial Basis Functions

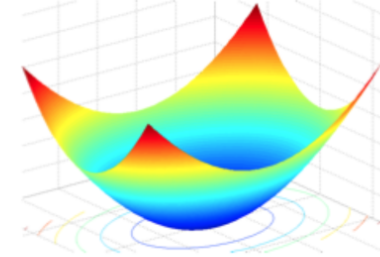
Thin Plate Spline $\Phi(r) = r^2 \log r$



Gaussians $\Phi(r) = e^{-r^2}$

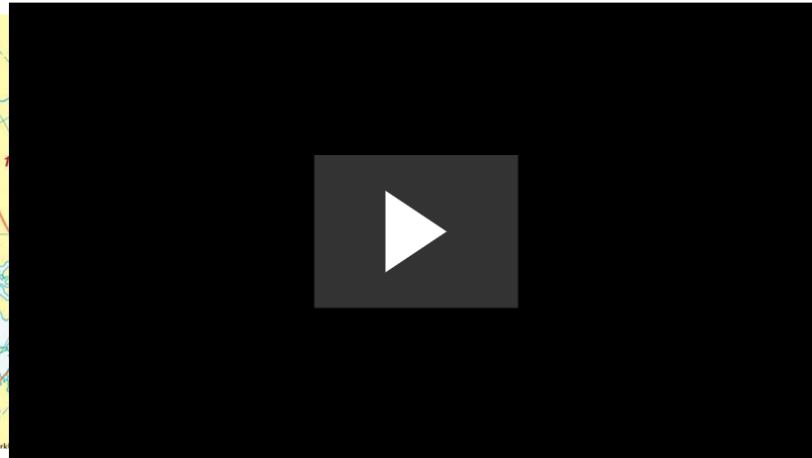


Multiquadrics $\Phi(r) = \sqrt{r^2 + 1}$



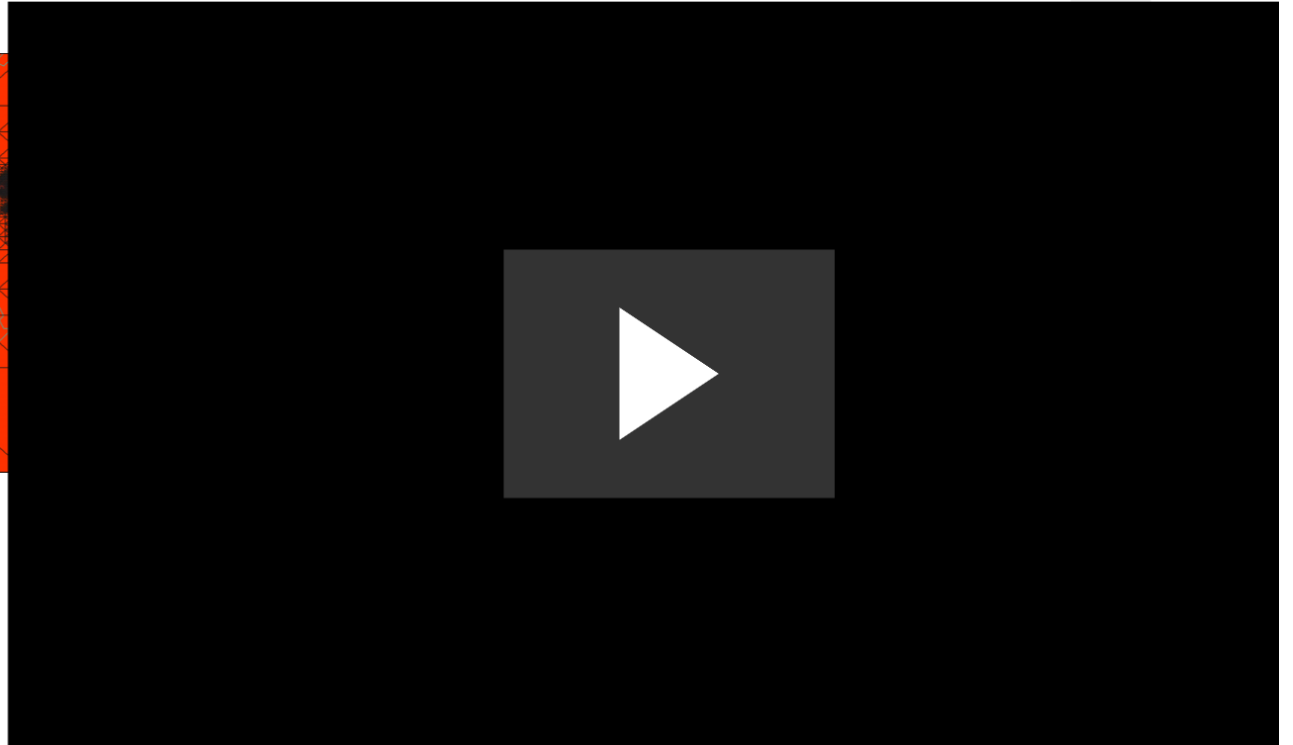
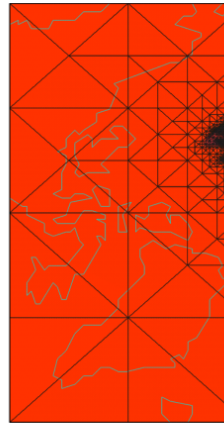
Application: Tracer Transport

Given: Wind Data



<http://www.lib.utexas.edu/maps>

Tracer Transport (Adaptive Mesh)



Tracer Transport (Mesh-free)

