

# Scalar advection with adaptive moving meshes

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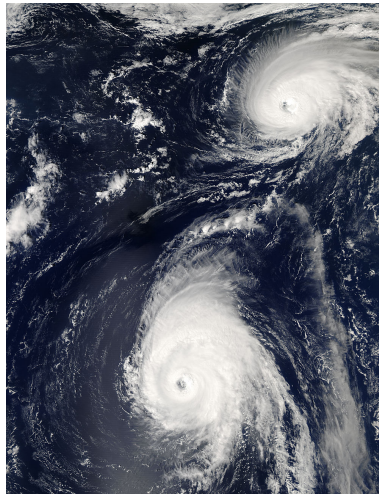
MetStröm

Deutsche  
Forschungsgemeinschaft

**DFG**

# Motivation

- Atmospheric flows: Interaction of a large range of different scales.
- Adequate resolution of the various scales important for the dynamics requires extreme computational effort.
- Use of dynamical adaptive resolution technique could provide one possibility to reduce computational cost in order to achieve a certain resolution of the overall flow.



- DFG-MetStröm priority programme.
- Apply solution-adaptive resolution technique in EULAG for simulating atmospheric flows.
- Extend recent work of Smolarkiewicz and Prusa (2003) by using *a posteriori* criteria and mesh differential equations for adaption in EULAG.
- ▶ Here, presentation of the methodology by means of application to 2D linear advection problem with solver MPDATA.

- 1 Introduction
- 2 Formulation of the linear advection model based on design of EULAG
- 3 Dynamic grid adaption
- 4 Example application: Rotating cone

Continuous mapping approach: Adaptive curvilinear grid in physical domain  $S_p$  specified by time-variable mapping

$$\mathbf{x}(\bar{\mathbf{x}}, \bar{t}) : S_t \rightarrow S_p$$

where  $S_t$  transformed domain with reference computational grid.

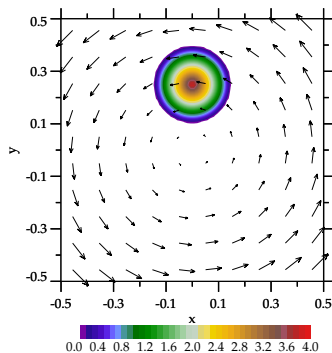
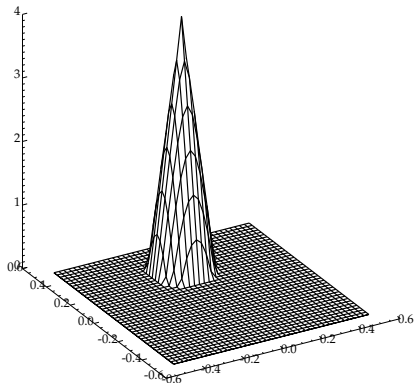
- Mapping may be specified either analytically or numerically.
- Redistribution, no insertion/deletion of grid points  
⇒ conserved data structures.
- Physical problem is formulated in transformed domain  $S_t$ .

Linear scalar advection in two dimensions

$$\frac{\partial \psi}{\partial t} + u \frac{\partial \psi}{\partial x} + v \frac{\partial \psi}{\partial y} = 0$$

- $\psi(x, y, t)$ : Passive scalar function of Cartesian coordinates  $x, y$
- $\mathbf{v} = (u, v)^T$ : Spatially-variable solenoidal velocity field
- Exact solution  $\psi(x, y, t) = \psi(x - ut, y - vt, 0)$ .

## Benchmark: Rotating cone



General time-dependent coordinate mapping of 3D solver EULAG:

$$(\bar{x}, \bar{y}, \bar{z}, \bar{t}) = (\bar{x}(x, y, t), \bar{y}(x, y, t), \bar{z}(x, y, z, t), t) : \mathbf{S}_p \rightarrow \mathbf{S}_t$$



General time-dependent coordinate mapping of 3D solver EULAG:

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General coordinate mapping in the 2D model:

$$(\bar{x}, \bar{y}, \bar{t}) = (\bar{x}(x, y, t), \bar{y}(x, y, t), t) : \mathbf{S}_p \rightarrow \mathbf{S}_t$$

Generalised Eulerian conservation form:

$$\frac{\partial(\bar{S}\psi)}{\partial\bar{t}} + \frac{\partial(\bar{S}\bar{u}\psi)}{\partial\bar{x}} + \frac{\partial(\bar{S}\bar{v}\psi)}{\partial\bar{y}} = 0$$

- $\bar{u}, \bar{v}$ : contravariant velocity components given as

$$\bar{u} = \frac{d\bar{x}}{d\bar{t}} = \frac{\partial\bar{x}}{\partial t} + u\frac{\partial\bar{x}}{\partial x} + v\frac{\partial\bar{x}}{\partial y} \quad \bar{v} = \frac{d\bar{y}}{d\bar{t}} = \frac{\partial\bar{y}}{\partial t} + u\frac{\partial\bar{y}}{\partial x} + v\frac{\partial\bar{y}}{\partial y}$$

- $\bar{S}$ : Jacobian of the transformation given as

$$\bar{S} = \left( \frac{\partial\bar{x}}{\partial x} \frac{\partial\bar{y}}{\partial y} - \frac{\partial\bar{x}}{\partial y} \frac{\partial\bar{y}}{\partial x} \right)^{-1}$$

$$\frac{\partial}{\partial \bar{x}} \left( \bar{S} \frac{\partial \bar{x}}{\partial x} \right) + \frac{\partial}{\partial \bar{y}} \left( \bar{S} \frac{\partial \bar{y}}{\partial x} \right) = 0$$

$$\frac{\partial}{\partial \bar{x}} \left( \bar{S} \frac{\partial \bar{x}}{\partial y} \right) + \frac{\partial}{\partial \bar{y}} \left( \bar{S} \frac{\partial \bar{y}}{\partial y} \right) = 0$$

$$\frac{\partial \bar{S}}{\partial \bar{t}} + \frac{\partial}{\partial \bar{x}} \left( \bar{S} \frac{\partial \bar{x}}{\partial t} \right) + \frac{\partial}{\partial \bar{y}} \left( \bar{S} \frac{\partial \bar{y}}{\partial t} \right) = 0.$$

Geometric Conservation Law (GCL)

Prusa et al. (2006)

- 1 Diagnostic Jacobian:

$$\bar{S}_d := \left( \frac{\partial \bar{x}}{\partial x} \frac{\partial \bar{y}}{\partial y} - \frac{\partial \bar{x}}{\partial y} \frac{\partial \bar{y}}{\partial x} \right)^{-1}$$

- 2 Time-component of the GCL:

$$\frac{\partial \bar{S}}{\partial t} + \frac{\partial}{\partial \bar{x}} \left( \bar{S} \frac{\partial \bar{x}}{\partial t} \right) + \frac{\partial}{\partial \bar{y}} \left( \bar{S} \frac{\partial \bar{y}}{\partial t} \right) = 0.$$

MPDATA for advection in time-dependent geometries:

$$\psi^{n+1} = \frac{\bar{S}^n}{\bar{S}^{n+1}} \left( \psi^n - \frac{\delta t}{\bar{S}^n} \nabla \cdot (\hat{\mathbf{v}}^{n+1/2} \psi^n) \right)$$

$$\psi_i^{n+1} = \frac{\bar{S}_i^n}{\bar{S}_i^{n+1}} \text{MPDATA}_i(\psi^n, \hat{\mathbf{v}}^{n+1/2}, \bar{S}^n)$$

Smolarkiewicz (1999,2006)

- ▶ Continuously redistribute fixed number grid points according to the evolution of the flow.
- Construct indicators for local amount of adaptivity.
- Distribute available grid points according to these indicators.

# Basic principle of mesh adaption: 1D-perspective

- With mapping:  $x(\bar{x}) : S_t \rightarrow S_p$  we estimate

$$x(\bar{x}_{i+1}) - x(\bar{x}_i) = \frac{\partial x}{\partial \bar{x}} \delta \bar{x} + \mathcal{O}(\delta \bar{x}^2)$$

- Relationship:

$$\text{grid point density} \sim (x_{i+1} - x_i)^{-1} \sim \left( \frac{\partial x}{\partial \bar{x}} \right)^{-1} = \frac{\partial \bar{x}}{\partial x}$$

- Introduce monitor function  $m(x) : S_p \rightarrow \mathbb{R}^+$  chosen that

$$m(x) \sim \frac{\partial \bar{x}}{\partial x}$$

# Basic principle of mesh adaption: 1D-perspective

- ansatz:

$$c \cdot m(x) = \frac{\partial \bar{x}}{\partial x} \quad \text{or} \quad \frac{\partial}{\partial x} \left( m^{-1}(x) \frac{\partial \bar{x}}{\partial x} \right) = 0 + BCs$$

- idea:  $m(x) \rightarrow \bar{x}(x) \rightarrow x(\bar{x})$
- ▶  $x(\bar{x})$  satisfies equidistribution condition with respect to the monitor function  $m(x)$ .



# Basic principle of mesh adaption: 1D-perspective

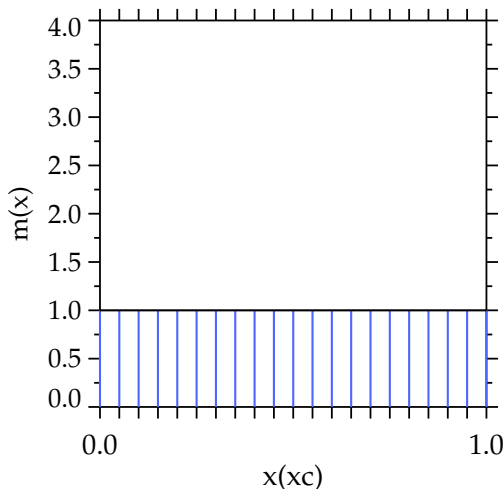
- Example:

$$S_p = [0, 1]$$

$$\bar{x}(0) = 0, \quad \bar{x}(1) = 1$$

$$m(x) = 1$$

Number of grid increments:  $N = 20$



# Basic principle of mesh adaption: 1D-perspective

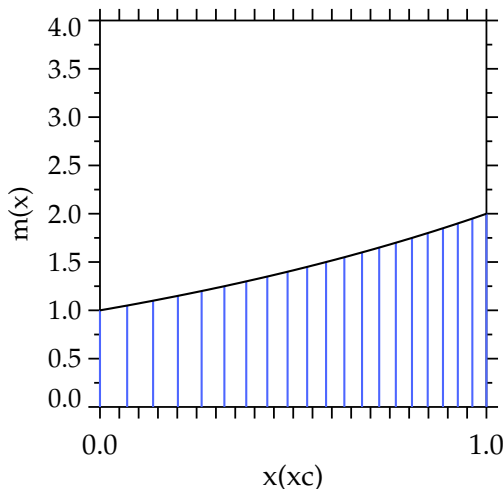
- Example:

$$S_p = [0, 1]$$

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$$m(x) = \exp(x \ln 2)$$

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# Basic principle of mesh adaption: 1D-perspective

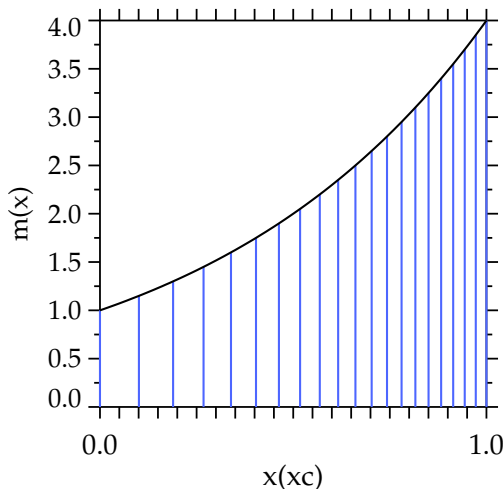
- Example:

$$S_p = [0, 1]$$

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Number of grid increments:  $N = 20$



# Basic principle of mesh adaption: 1D-perspective

- Example:

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$$\bar{x}(0) = 0, \quad \bar{x}(1) = 1$$

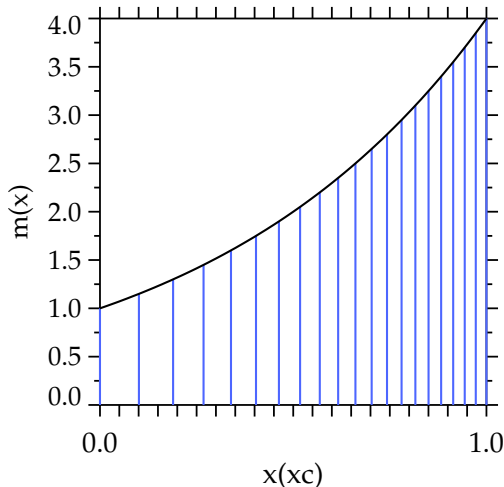
$$m(x) = \exp(x \ln 4)$$

$$\int_{x_{i-1}}^{x_i} m(x) dx = c$$

$$i = 1, \dots, N$$

$$c = \frac{1}{N} \int_0^1 m(x) dx$$

Number of grid increments:  $N = 20$



Mesh adaption functional:

$$I[\bar{\mathbf{x}}] = \frac{1}{2} \int_{\mathbf{S}_p} \sum_{i=1}^2 (\nabla \bar{\mathbf{x}}^i)^T M^{-1} \nabla \bar{\mathbf{x}}^i d\mathbf{x}$$

- $M$  denotes  $2 \times 2$  symmetric positive definite monitor matrix
- $M$  constitutes link to physical solution
- $M$  can also be used to influence the geometric properties of the generated mesh

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Euler-Lagrange equations:

$$\nabla \cdot (M^{-1} \nabla \bar{x}^i) = 0 \quad i = 1, 2$$

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In one dimension:

$$\frac{\partial}{\partial x} \left( m^{-1} \frac{\partial \bar{x}}{\partial x} \right) = 0$$

We define

$$M = \begin{pmatrix} q & 0 \\ 0 & q \end{pmatrix}$$

- $q = q(\mathbf{x}, t)$  strictly positive scalar weight function.
- Grid point concentration where  $q$  relatively large.
- Local gradient criteria:

$$q(\mathbf{x}, t) = \alpha + \frac{\beta}{(1 - \beta)} |\nabla \psi(\mathbf{x}, t)|^{\frac{1}{m}} \quad \beta \in (0.0, 0.95)$$

- Low-pass filtering is applied to  $q$ .



# Static equidistant mesh



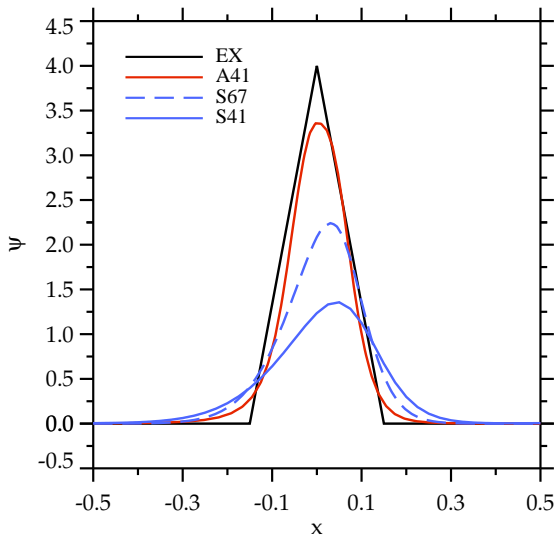
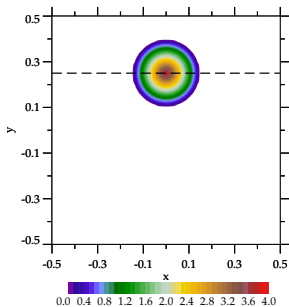
# Dynamic adaption



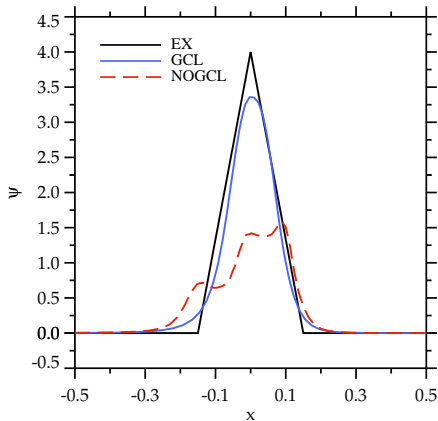
# Dynamic adaption



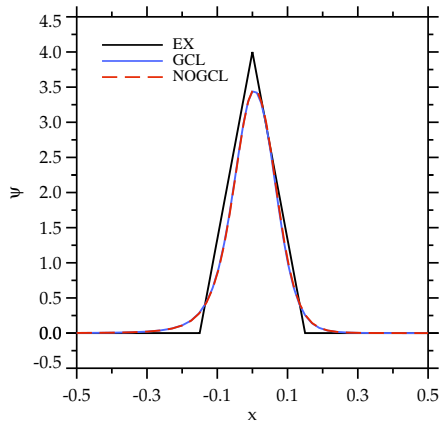
# Comparison dynamic adaptive vs. static equidistant mesh



# Importance of the geometric conservation law



6 filter passes of monitor  
function



20 filter passes of monitor  
function