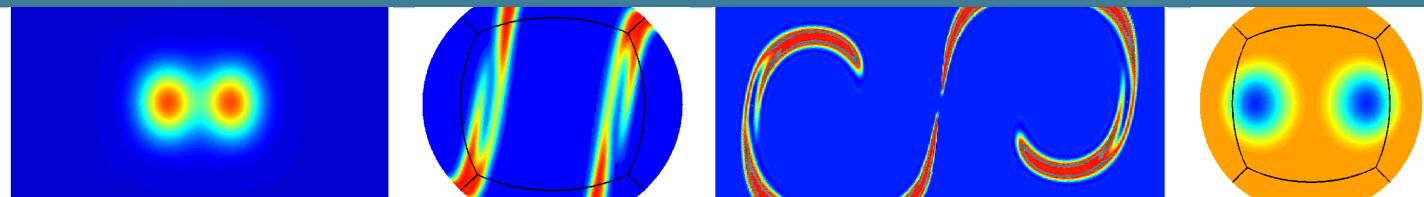
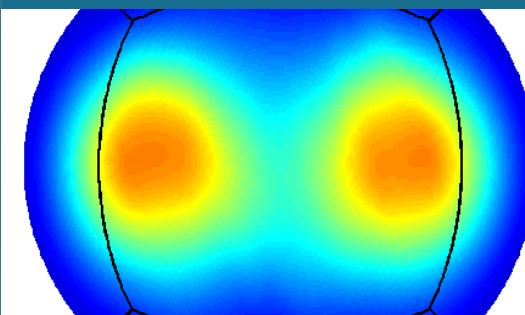




Sandia
National
Laboratories

Optimization-based, property-preserving algorithms for passive tracer transport



Pavel Bochev, Kara Peterson, and Denis Ridzal

10th International Congress on Industrial and Applied Mathematics

August 20-25, Waseda University, Tokyo, Japan.



Office of



Sandia National Laboratories is a multimission laboratory managed and operated by National Technology & Engineering Solutions of Sandia, LLC, a wholly owned subsidiary of Honeywell International Inc., for the U.S. Department of Energy's National Nuclear Security Administration under contract DE-NA0003525.



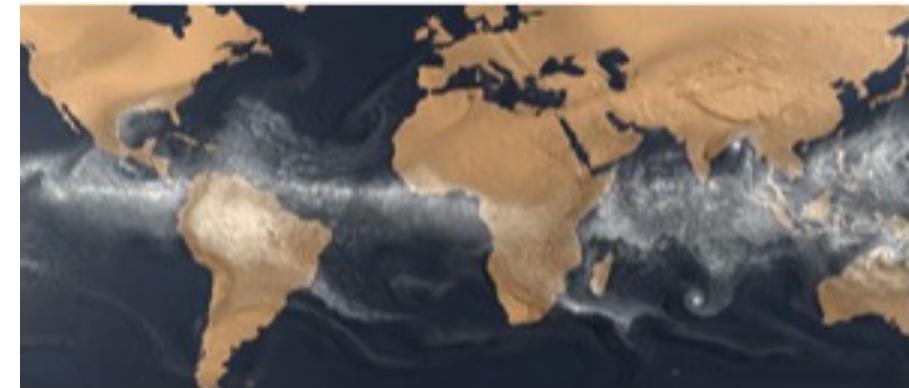
Sandia National Laboratories is a multimission laboratory managed and operated by National Technology & Engineering Solutions of Sandia, LLC, a wholly owned subsidiary of Honeywell International Inc., for the U.S. Department of Energy's National Nuclear Security Administration under contract DE-NA0003525.

Motivation & model problem



Why are transport schemes so important for ESM?

- Atmosphere is the **most expensive** component of Earth System Models
- Tracer advection is the **dominant cost** in atmosphere simulations
- With biogeochemistry **100-1000** tracers are needed



Tracer-density system

$$\left\{ \begin{array}{l} \frac{\partial \rho}{\partial t} + \nabla \cdot \rho \mathbf{v} = 0 \\ \frac{\partial \rho \tau}{\partial t} + \nabla \cdot \rho \tau \mathbf{v} = 0 \end{array} \right. \rightarrow \frac{D\tau}{Dt} \equiv \frac{\partial \tau}{\partial t} + \mathbf{v} \cdot \nabla \tau = 0$$

ρ - Density
 τ - Tracer mixing ratio
 \mathbf{v} - Velocity

Objectives: A numerical transport algorithm for the solution of the tracer transport equations that is

- Accurate
- Efficient
- Works on unstructured grids
- Property preserving**

$$\left\{ \begin{array}{l} M = \int_{\Omega} \rho dx \quad Q = \int_{\Omega} \rho \tau dx \quad \leftarrow \text{Conservation of mass and total tracer} \\ \rho^{\min} \leq \rho \leq \rho^{\max} \quad \tau^{\min} \leq \tau \leq \tau^{\max} \quad \leftarrow \text{Preservation of local bounds} \\ \zeta = a\tau + b \quad \leftarrow \text{Preservation of tracer correlations} \end{array} \right.$$

Background



To develop our method we start with a **generic semi-Lagrangian scheme** for tracer transport.

- Semi-Lagrangian schemes are popular in the **geophysical modeling community** because they allow **much larger time steps** than the CFL- restricted time steps in explicit Eulerian methods.

Nomenclature

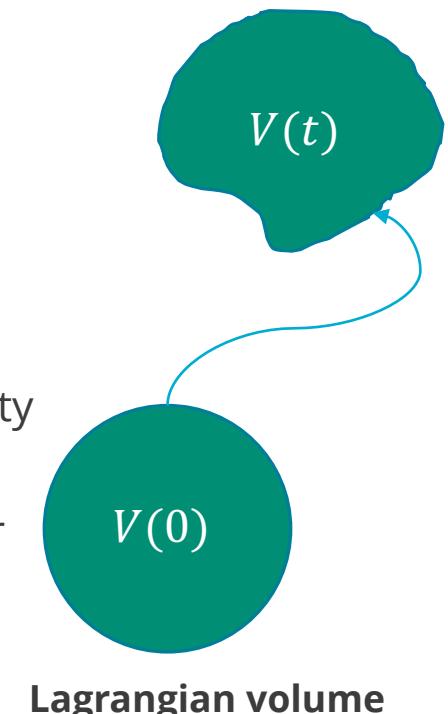
$$\mu_{V(t)} = \int_{V(t)} dV \quad \leftarrow \text{Measure of } V(t)$$

$$m_{V(t)} = \int_{V(t)} \rho(\mathbf{x}, t) dV \quad \leftarrow \text{Mass contained in } V(t)$$

$$q_{V(t)} = \int_{V(t)} \rho(\mathbf{x}, t) \tau(\mathbf{x}, t) dV \quad \leftarrow \text{Total tracer in } V(t)$$

$$\bar{\rho}_{V(t)} = \frac{m_{V(t)}}{\mu_{V(t)}} \quad \leftarrow \text{Volume averaged density}$$

$$\bar{\tau}_{V(t)} = \frac{q_{V(t)}}{m_{V(t)}} \quad \leftarrow \text{Density weighted tracer}$$



Mathematical basis for semi-Lagrangian schemes

$$\frac{d}{dt} m_{V(t)} = 0 \longrightarrow \bar{\rho}_{V(t_{n+1})} = \frac{m_{V(t_{n+1})}}{\mu_{V(t_{n+1})}} = \frac{m_{V(t_n)}}{\mu_{V(t_{n+1})}}$$

Mass and total tracer mass are preserved in Lagrangian volumes

$$\frac{d}{dt} q_{V(t)} = 0 \longrightarrow \bar{\tau}_{V(t_{n+1})} = \frac{q_{V(t_{n+1})}}{m_{V(t_{n+1})}} = \frac{q_{V(t_n)}}{m_{V(t_n)}}$$

Lagrangian update formulas for the averaged quantities

A generic semi-Lagrangian scheme for tracer transport

We consider a scheme based on **backward incremental remapping**

$C(\Omega)$ - Fixed (Eulerian) grid with cells c_i

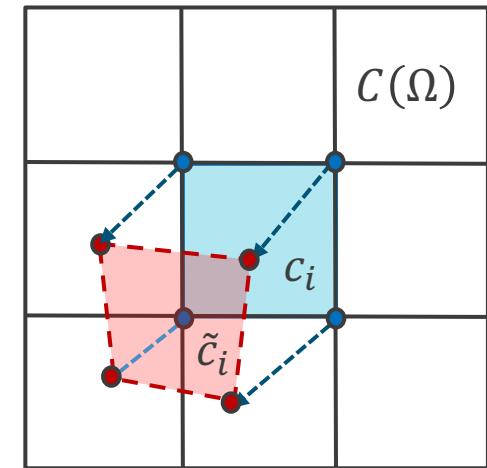
$$V(t_k) := c_i \rightarrow \mu_{i,k}, m_{i,k}, q_{i,k}, \bar{\rho}_{i,k}, \bar{\tau}_{i,k}$$

$\tilde{C}(\Omega)$ - Deformed (Lagrangian) grid with cells \tilde{c}_i

$$V(t_k) := \tilde{c}_i \rightarrow \tilde{\mu}_{i,k}, \tilde{m}_{i,k}, \tilde{q}_{i,k}, \tilde{\bar{\rho}}_{i,k}, \tilde{\bar{\tau}}_{i,k}$$

$\tilde{C}(\Omega)$ = **Backward Lagrangian increment**: the nodes x_i of $C(\Omega)$ are moved backward in time to positions $x(t - \Delta t)$ by solving for each node

$$\begin{cases} \dot{x}(s) = \mathbf{v}(x, s) \\ x(t) = x_j \end{cases}$$



Define the following quantities at the current time t_n :

$$M_n := \sum_{i=1}^{N_C} m_{i,n} = \sum_{i=1}^{N_C} \bar{\rho}_{i,n} \mu_{i,n}$$

$$\bar{\rho}_{i,n}^{\min} = \begin{cases} \min_{j \in N(c_i)} \{\bar{\rho}_{j,n}\} & \text{if } c_i \cap \partial\Omega = \emptyset \\ \min \left\{ \min_{j \in N(c_i)} \{\bar{\rho}_{j,n}\}, \min_{\mathbf{x} \in N(c_i) \cap \partial\Omega} \rho(\mathbf{x}, t_n) \right\} & \text{if } c_i \cap \partial\Omega \neq \emptyset \end{cases}$$

$$Q_n := \sum_{i=1}^{N_C} q_{i,n} = \sum_{i=1}^{N_C} \bar{\tau}_{i,n} m_{i,n}$$

$$\bar{\rho}_{i,n}^{\max} = \begin{cases} \max_{j \in N(c_i)} \{\bar{\rho}_{j,n}\} & \text{if } c_i \cap \partial\Omega = \emptyset \\ \max \left\{ \max_{j \in N(c_i)} \{\bar{\rho}_{j,n}\}, \max_{\mathbf{x} \in N(c_i) \cap \partial\Omega} \rho(\mathbf{x}, t_n) \right\} & \text{if } c_i \cap \partial\Omega \neq \emptyset \end{cases}$$

Total mass and total tracer mass

Physically motivated local density bounds (similar for τ)

A generic semi-Lagrangian scheme for tracer transport



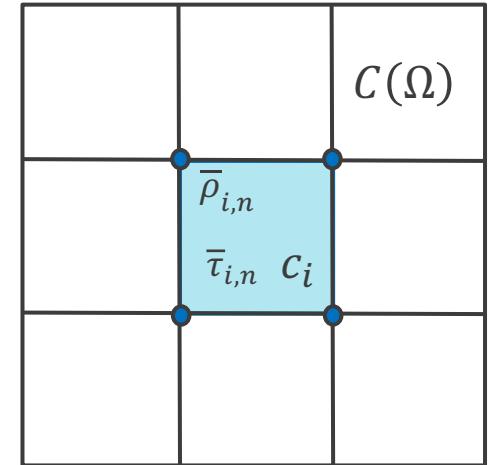
Statement of the monotone, conservative and compatible tracer transport problem

Given: $\bar{\rho}_{i,n}$ and $\bar{\tau}_{i,n}$ on the **fixed grid** $C(\Omega)$ at **current** time t_n

Find: $\bar{\rho}_{i,n+1}$ and $\bar{\tau}_{i,n+1}$ on the **fixed grid** $C(\Omega)$ at **future** time t_{n+1} , satisfying

P1. Monotonicity and compatibility: for all $c_i \in C(\Omega)$

$$\alpha \bar{\rho}_{i,n}^{\min} \leq \bar{\rho}_{i,n+1} \leq \alpha \bar{\rho}_{i,n}^{\max} \quad \text{and} \quad \bar{\tau}_{i,n}^{\min} \leq \bar{\tau}_{i,n+1} \leq \bar{\tau}_{i,n}^{\max} \quad \text{where} \quad \alpha = \frac{\tilde{\mu}_{i,n}}{\mu_{i,n+1}}$$



Note that τ is **constant** along the characteristics but **in general ρ is not**. Thus, the local bounds for $\bar{\tau}_{i,n+1}$ are **unchanged** from those at the current time t_n but **the local bounds for $\bar{\rho}_{i,n+1}$ are different!**

P2. Conservation of mass and total tracer mass

$$M_{n+1} := \sum_{i=1}^{N_C} m_{i,n+1} = M_n \quad \text{and} \quad Q_{n+1} := \sum_{i=1}^{N_C} q_{i,n+1} = Q_n$$

A generic semi-Lagrangian scheme for tracer transport



Solution: the Lagrangian update formula implies

$$\bar{\rho}_{V(t_{n+1})} = \frac{m_{V(t_{n+1})}}{\mu_{V(t_{n+1})}} = \frac{m_{V(t_n)}}{\mu_{V(t_{n+1})}} \longrightarrow \bar{\rho}_{i,n+1} = \frac{m_{i,n+1}}{\mu_{i,n+1}} = \frac{\tilde{m}_{i,n}}{\mu_{i,n+1}}$$

$$\tilde{m}_{i,n} \approx \int_{\tilde{c}_i} \rho(x, t_n) dx$$

where

$$\bar{\tau}_{V(t_{n+1})} = \frac{q_{V(t_{n+1})}}{m_{V(t_{n+1})}} = \frac{q_{V(t_n)}}{m_{V(t_n)}} \longrightarrow \bar{\tau}_{i,n+1} = \frac{q_{i,n+1}}{m_{i,n+1}} = \frac{\tilde{q}_{i,n}}{\tilde{m}_{i,n}}$$

$$\tilde{q}_{i,n} \approx \int_{\tilde{c}_i} \rho(x, t_n) \tau(x, t_n) dx$$

- We need the average mass $\tilde{m}_{i,n}$ and tracer $\tilde{q}_{i,n}$ on the cells \tilde{c}_i of the **deformed mesh** $\tilde{\mathcal{C}}(\Omega)$.
- However, the solution at **current** time t_n is given on the cells c_i of the **fixed mesh** $\mathcal{C}(\Omega)$
- Classical schemes use **monotone & conservative reconstructions** of $\rho(x, t_n)$ and $\tau(x, t_n)$ on the cells c_i of the **fixed mesh** $\mathcal{C}(\Omega)$ to compute the integrals on the cells \tilde{c}_i of the **deformed mesh**:

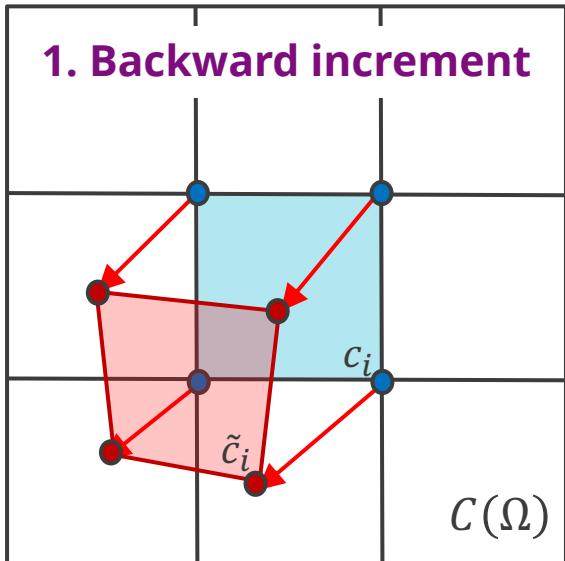
$$\rho_n^h(x) \approx \rho(x, t_n) \longrightarrow \rho_{i,n}^h(\mathbf{x}) = \bar{\rho}_{i,n} + \mathbf{g}_{\rho,i} \cdot (\mathbf{x} - \mathbf{x}_{b_i}), \quad \mathbf{x}_{b_i} = \frac{\int_{c_i} \mathbf{x} dV}{\mu_i}$$

- Monotone
- Conservative
- Mean-preserving

Slope-limited gradient reconstructions

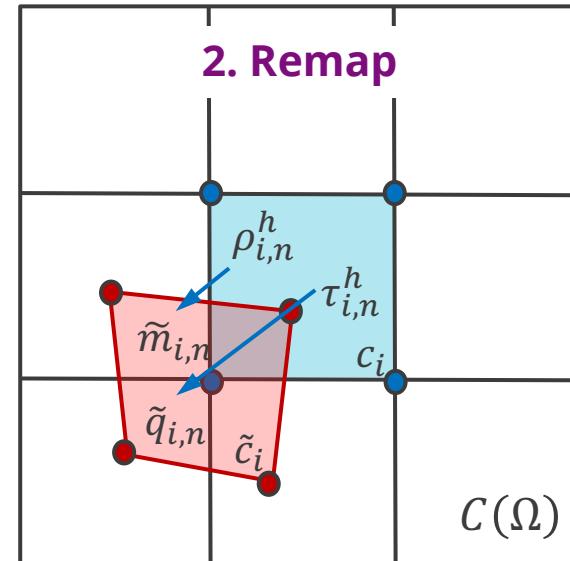
$$\tau_n^h(x) \approx \tau(x, t_n) \longrightarrow \tau_{i,n}^h(\mathbf{x}) = \bar{\tau}_{i,n} + \mathbf{g}_{\tau,i} \cdot (\mathbf{x} - \mathbf{x}_{c_i}), \quad \mathbf{x}_{c_i} = \frac{\int_{c_i} \mathbf{x} \rho_{i,n}^h(\mathbf{x}) dV}{\int_{c_i} \rho_{i,n}^h(\mathbf{x}) dV}$$

We will modify the last two stages of the generic scheme



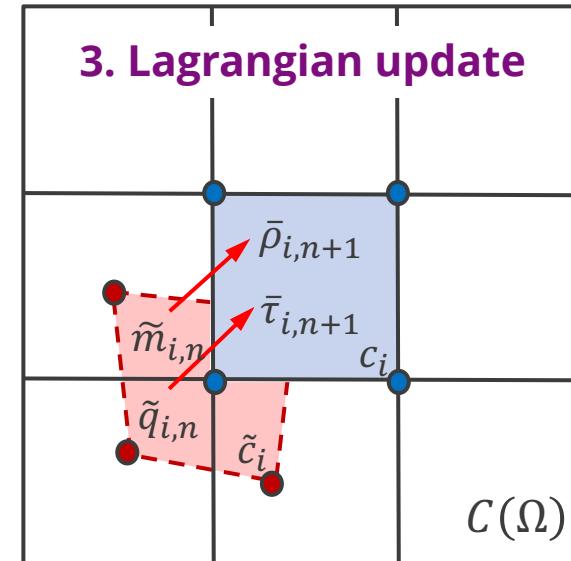
$$\dot{\mathbf{x}}(s) = \mathbf{v}(\mathbf{x}, s)$$

$$\mathbf{x}(t) = \mathbf{x}_j$$



$$\tilde{m}_{i,n} = \int_{\tilde{c}_i} \rho_n^h(\mathbf{x}) dV$$

$$\tilde{q}_{i,n} = \int_{\tilde{c}_i} \rho_n^h(\mathbf{x}) \tau_n^h(\mathbf{x}) dV$$



$$\bar{\rho}_{i,n+1} = \frac{\tilde{m}_{i,n}}{\mu_{i,n+1}}$$

$$\bar{\tau}_{i,n+1} = \frac{\tilde{q}_{i,n}}{\tilde{m}_{i,n}}$$

Monotone reconstruction drawbacks:

- Based on local worst-case scenarios
- Mixes accuracy with preservation of properties
- More difficult to ascertain solution optimality

Our changes separate accuracy from property preservation:

- **Remap stage:** use 2nd order accurate but not monotone and/or conservative reconstruction
- **Lagrangian update stage:** enforce properties by coaching this stage into a constrained optimization problem.

Optimization-based semi-Lagrangian tracer transport



New remap stage

$$\begin{aligned}\tilde{m}_{i,n}^T &= \int_{\tilde{c}_i} \rho_n^{h,LS}(\mathbf{x}) dV & \leftarrow \rho_n^{h,LS} &= \bar{\rho}_{i,n} + \mathbf{g}_{\rho,i}^{LS} \cdot (\mathbf{x} - \mathbf{x}_{b_i}) \\ \tilde{q}_{i,n}^T &= \int_{\tilde{c}_i} \rho_n^{h,LS}(\mathbf{x}) \tau_n^{h,LS}(\mathbf{x}) dV & \leftarrow \tau_n^{h,LS} &= \bar{\tau}_{i,n} + \mathbf{g}_{\tau,i}^{LS} \cdot (\mathbf{x} - \mathbf{x}_{c_i})\end{aligned}$$

Least-squares reconstruction of the gradients on cell c_i

- $\tilde{m}_{i,n}^T$ and $\tilde{q}_{i,n}^T$ are 2nd order accurate but not guaranteed to be monotone and/or conservative

New Lagrangian update stage. Part 1 – definition of targets & local bounds at future time t_{n+1}

$$\bar{\rho}_{i,n+1}^T = \frac{\tilde{m}_{i,n}^T}{\mu_{i,n+1}} \quad \bar{\rho}_{i,n+1}^{\min} = \bar{\rho}_{i,n}^{\min} \frac{\tilde{\mu}_{i,n}}{\mu_{i,n+1}} \quad \bar{\rho}_{i,n+1}^{\max} = \bar{\rho}_{i,n}^{\max} \frac{\tilde{\mu}_{i,n}}{\mu_{i,n+1}}, \quad \alpha = \frac{\tilde{\mu}_{i,n}}{\mu_{i,n+1}}$$

$$\bar{\tau}_{i,n+1}^T = \frac{\tilde{q}_{i,n}^T}{\tilde{m}_{i,n}}$$

Tracer is constant along characteristics: **reuse bounds** from t_n

- $\bar{\rho}_{i,n+1}^T$ and $\bar{\tau}_{i,n+1}^T$ are 2nd order accurate but not guaranteed to be monotone and/or conservative

Optimization-based semi-Lagrangian tracer transport



New **Lagrangian update** stage. Part 2 - enforce properties at future time t_{n+1}

Density:
$$\begin{cases} \text{minimize} & \frac{1}{2} \sum_{i=1}^{N_C} \mu_{i,n+1}^2 (\bar{\rho}_{i,n+1} - \bar{\rho}_{i,n+1}^T)^2 \quad \text{subject to} \\ & \sum_{i=1}^{N_C} \bar{\rho}_{i,n+1} \mu_{i,n+1} = M_n \quad \text{and} \quad \bar{\rho}_{i,n+1}^{\min} \leq \bar{\rho}_{i,n+1} \leq \bar{\rho}_{i,n+1}^{\max}; \quad i = 1, \dots, N_C \end{cases} \rightarrow \text{Set } \bar{\rho}_{i,n+1} = \bar{\rho}_{i,n+1}^{\text{OPT}}$$

$$m_{i,n+1}^{\text{OPT}} := \bar{\rho}_{i,n+1}^{\text{OPT}} \mu_{i,n+1}$$

Tracer:
$$\begin{cases} \text{minimize} & \frac{1}{2} \sum_{i=1}^{N_C} (\bar{\tau}_{i,n+1} - \bar{\tau}_{i,n+1}^T)^2 \quad \text{subject to} \\ & \sum_{i=1}^{N_C} \bar{\tau}_{i,n+1} m_{i,n+1}^{\text{OPT}} = Q_n \quad \text{and} \quad \bar{\tau}_{i,n}^{\min} \leq \bar{\tau}_{i,n+1} \leq \bar{\tau}_{i,n}^{\max}; \quad i = 1, \dots, N_C \end{cases} \rightarrow \text{Set } \bar{\tau}_{i,n+1} = \bar{\tau}_{i,n+1}^{\text{OPT}}$$

Theorem 1. Existence & uniqueness

The feasible sets of the optimization problems are non-empty & have unique solutions

Properties



Theorem 2. Global optimality of the optimization-based solution

$$\frac{1}{2} \sum_{i=1}^{N_C} \mu_{i,n+1}^2 (\bar{\rho}_{i,n+1}^{\text{OPT}} - \bar{\rho}_{i,n+1}^{\text{T}})^2 \leq \frac{1}{2} \sum_{i=1}^{N_C} \mu_{i,n+1}^2 (\bar{\rho}_{i,n+1}^{\text{G}} - \bar{\rho}_{i,n+1}^{\text{T}})^2$$

Generic scheme

$$\frac{1}{2} \sum_{i=1}^{N_C} (\bar{\tau}_{i,n+1}^{\text{OPT}} - \bar{\tau}_{i,n+1}^{\text{T}})^2 \leq \frac{1}{2} \sum_{i=1}^{N_C} (\bar{\tau}_{i,n+1}^{\text{G}} - \bar{\tau}_{i,n+1}^{\text{T}})^2$$

The optimization-based solution is the **best possible**, with respect to the **targets**, approximate solution that also **satisfies properties P.1–P.2.**

Theorem 3. Preservation of tracer correlations

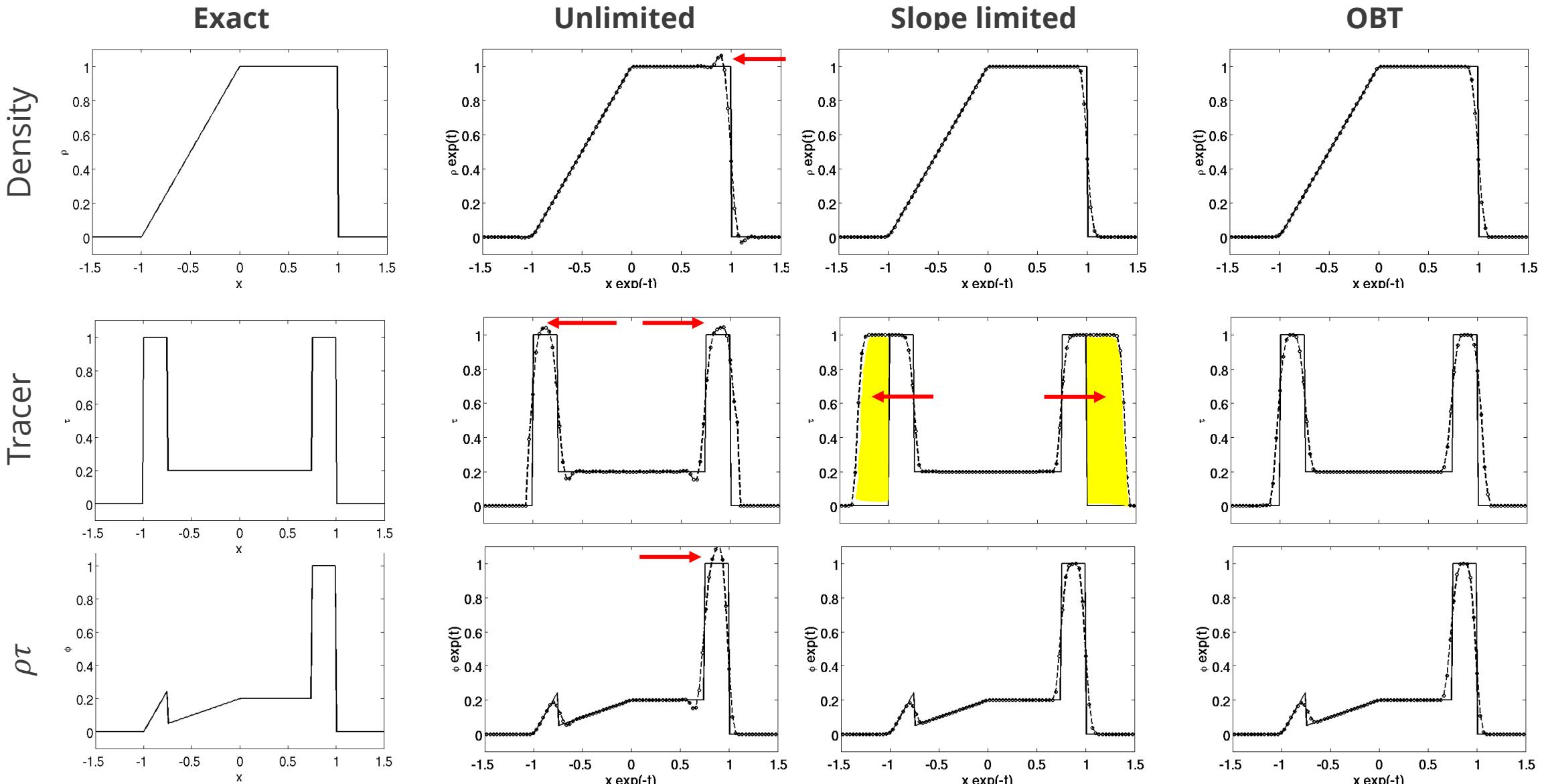
The optimization-based Lagrangian update stage **preserves linear tracer correlations**.

What about the cost?

$$\mathbf{q}_{n+1} = \underset{\mathbf{q}}{\operatorname{argmin}} \mathbf{q}^T \mathbf{M} \mathbf{q} + \mathbf{c}^T \mathbf{q} + \mathbf{c}_0 \quad \text{subject to} \quad \begin{cases} \mathbf{w}^T \mathbf{q} = \mathbf{w}^T \mathbf{q}_n & \leftarrow \text{Conservation} \\ \mathbf{q}^{\min} \leq \mathbf{q} \leq \mathbf{q}^{\max} & \leftarrow \text{Local bounds} \end{cases}$$

- ☞ A “*singly linearly constrained QP with simple bounds*”
- ☞ QP structure admits a **fast $O(N)$** optimization algorithm.

Numerical examples in 1D: Compatibility test



Numerical examples in 2D: Convergence



Tracer profile:

$$\tau = \sin(\pi x)^4 \sin(\pi y)^4 \exp(-\beta(x - x_0)^2 + (y - y_0)^2)$$

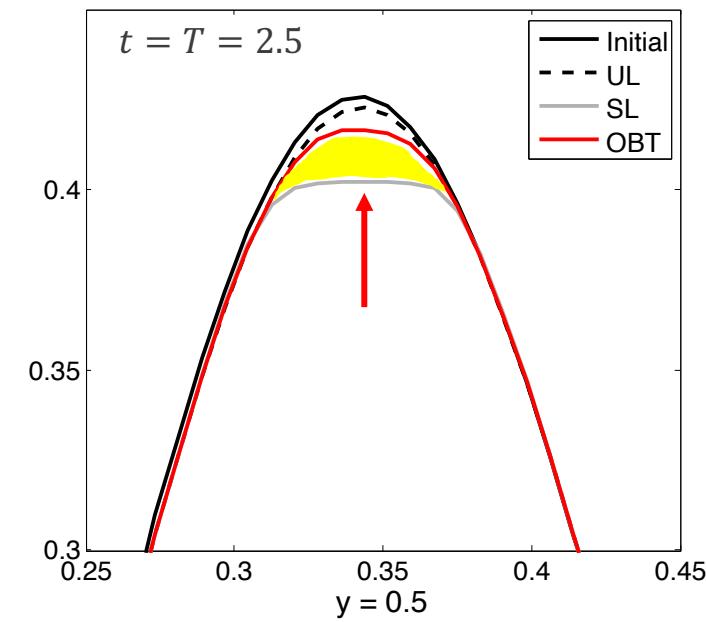
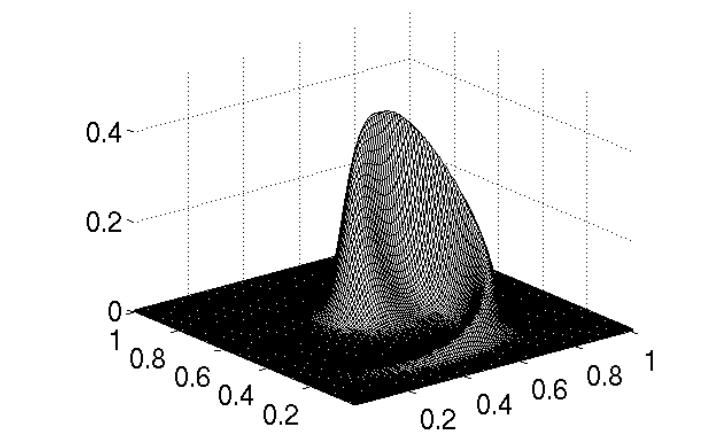
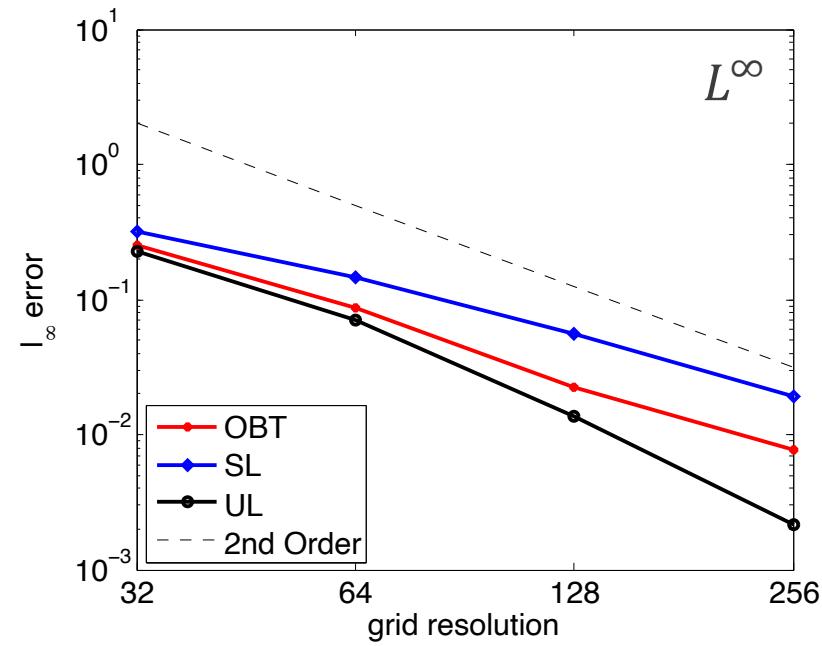
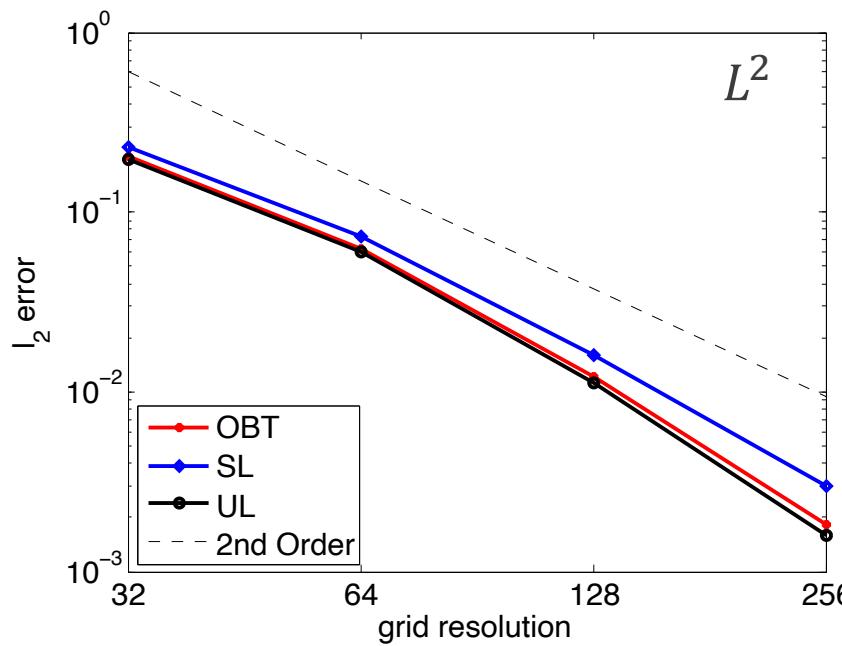
$$(x_0, y_0) = (0.25, 0.25); \beta = 40$$

Velocity field*

$$u = \sin^2(\pi x) \sin(2\pi y) \cos(\pi t/T)$$

$$v = -\sin^2(\pi y) \sin(2\pi x) \cos(\pi t/T)$$

$$T = 2.5$$



Numerical examples in 2D: Solid body rotation



Tracer profile:

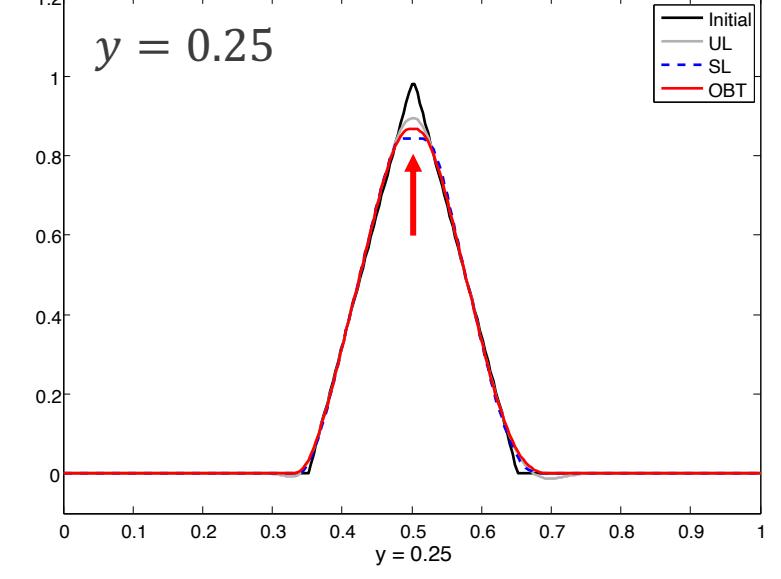
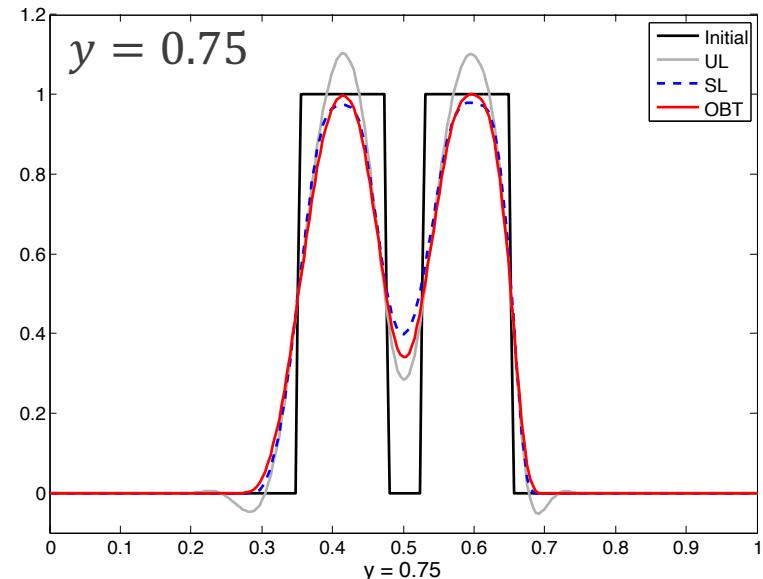
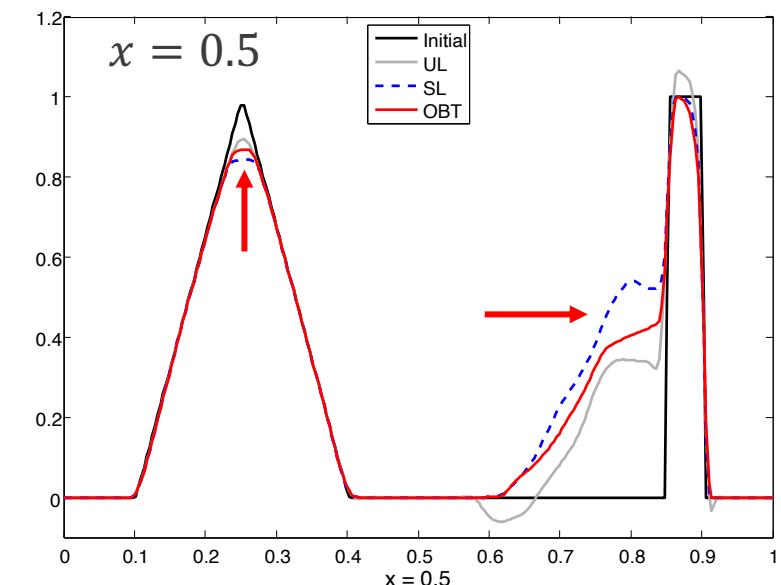
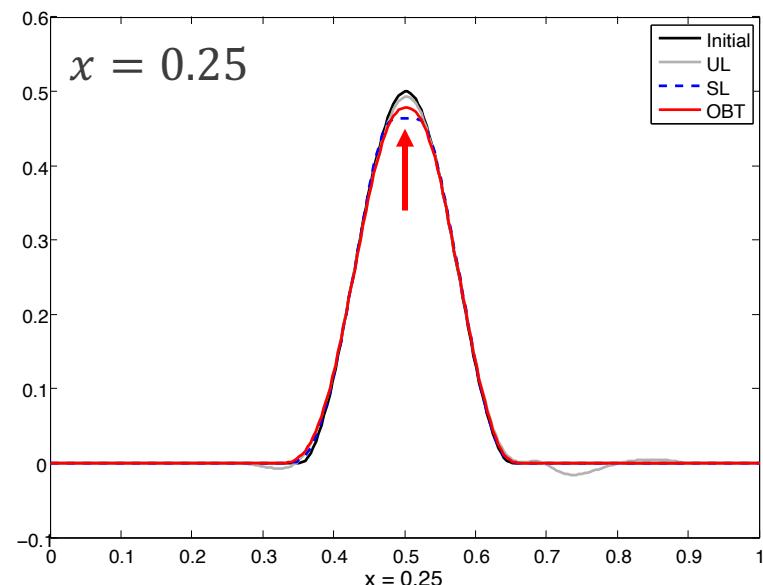
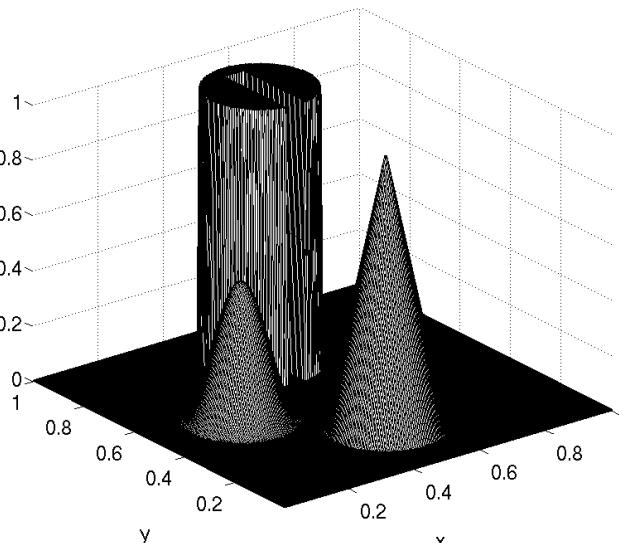
- Zalesak cylinder +
- cone +
- Gaussian hump

Velocity field*

$$u = \sin^2(\pi x) \sin(2\pi y) \cos(\pi t/T)$$

$$v = -\sin^2(\pi y) \sin(2\pi x) \cos(\pi t/T)$$

$$T = 2.5$$



(*)R. J. LeVeque, High-resolution conservative algorithms for advection in incompressible flow, SIAM Journal on Numerical Analysis 33 (1996) 627–665.

Numerical examples in 2D: Linear tracer correlation

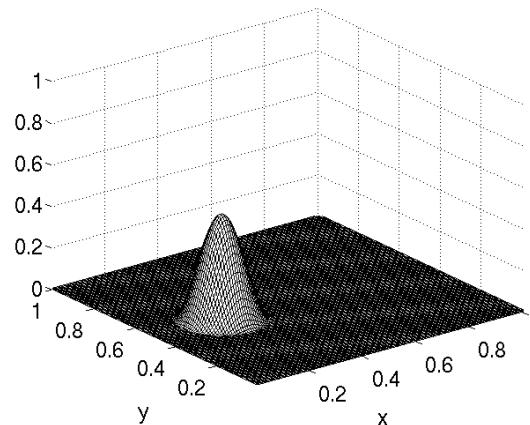


Tracers profile:

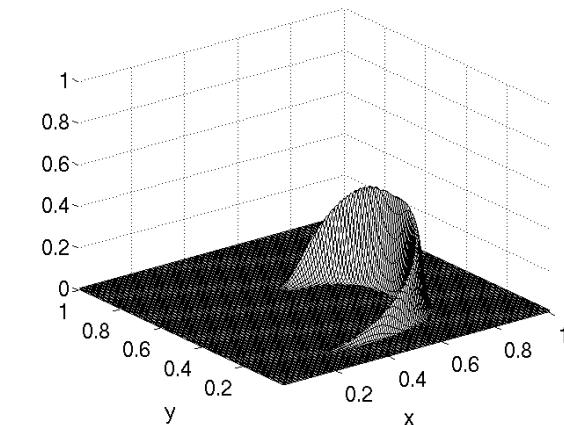
$$\tau = 0.4(1 + \cos(\pi r(x, y)))$$

$$\xi = -1.2\tau + 1$$

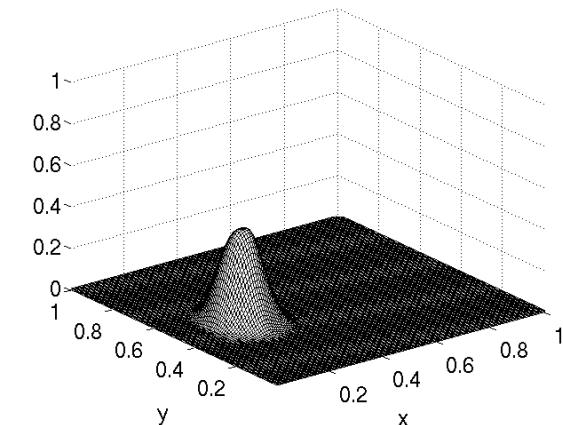
$$r(x, y) = \min\{\|x - x_0\|, r_0\}/r_0$$



$t = 0$



$t = 1.25$



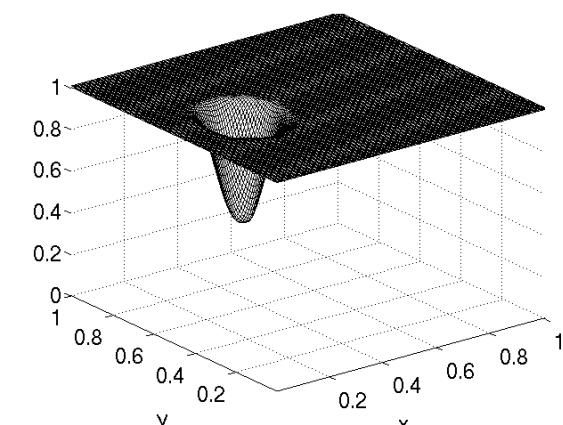
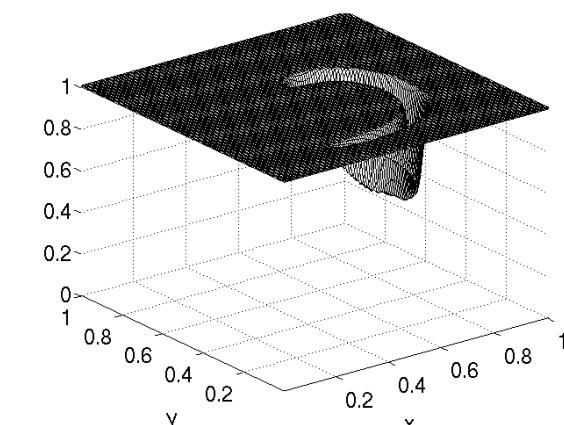
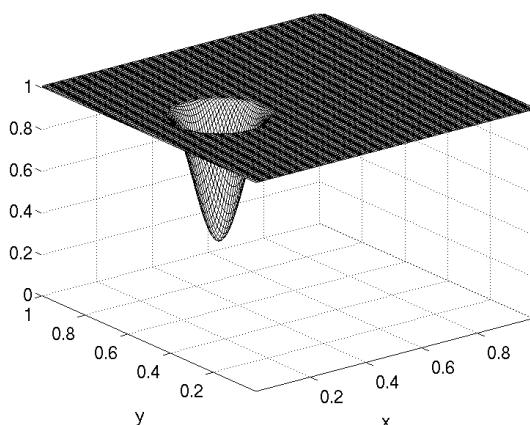
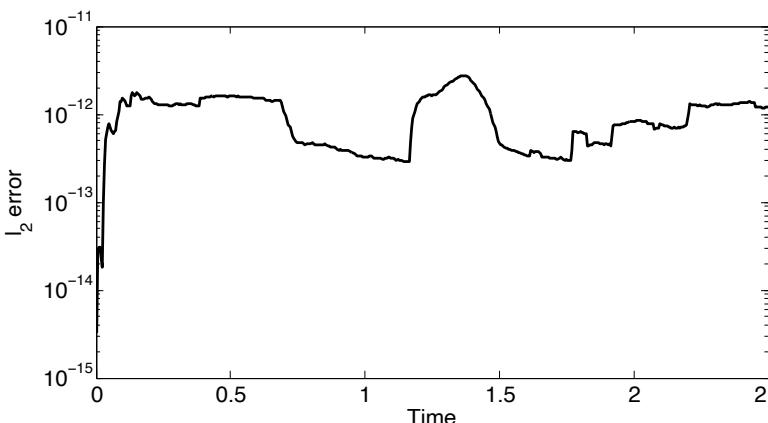
$t = 2.5$

Velocity field*

$$u = \sin^2(\pi x) \sin(2\pi y) \cos(\pi t/T)$$

$$v = -\sin^2(\pi y) \sin(2\pi x) \cos(\pi t/T)$$

$$T = 2.5$$



(*)R. J. LeVeque, High-resolution conservative algorithms for advection in incompressible flow, SIAM Journal on Numerical Analysis 33 (1996) 627–665.

Tracer transport on a sphere: setup and convergence

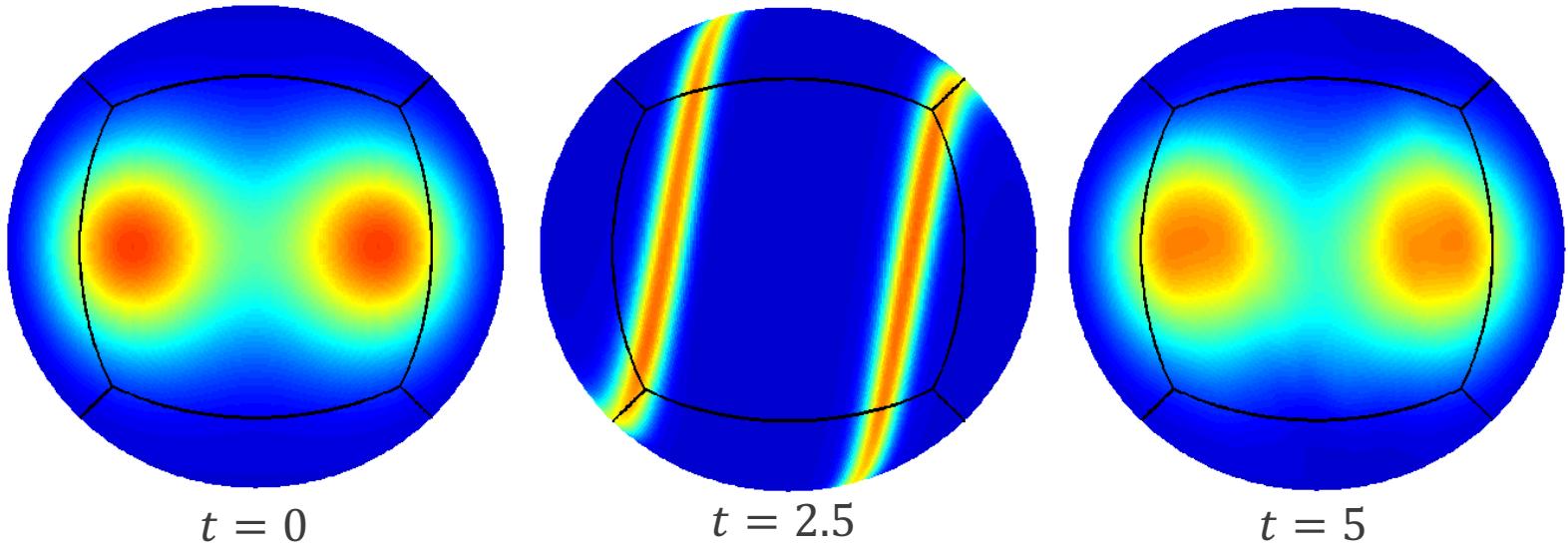
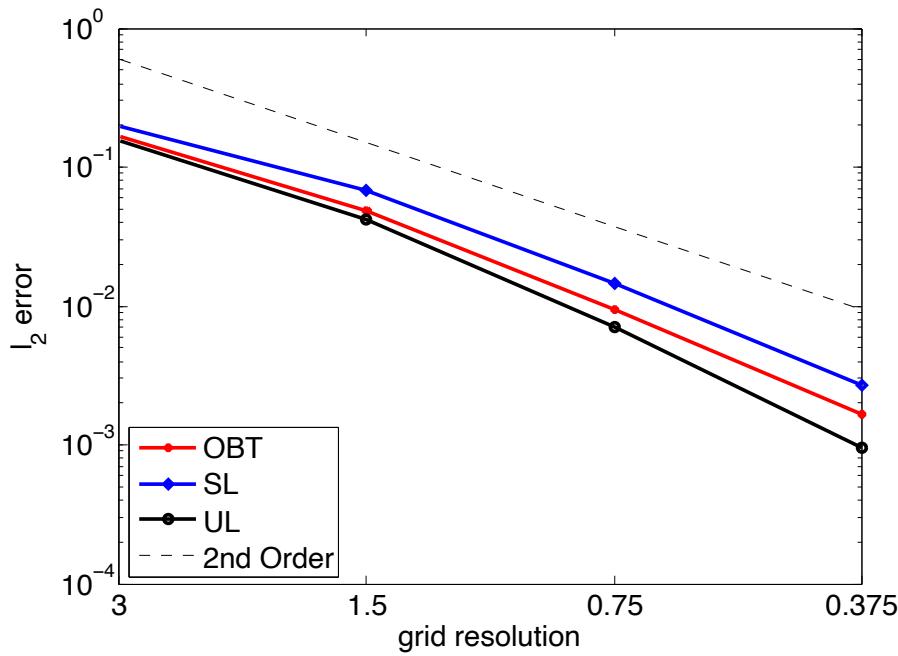
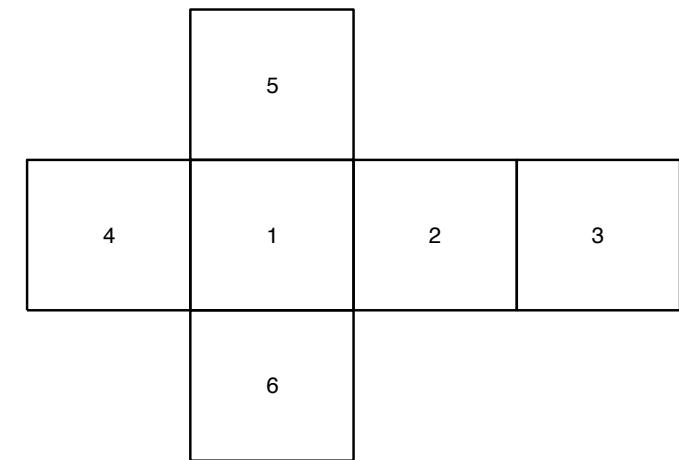
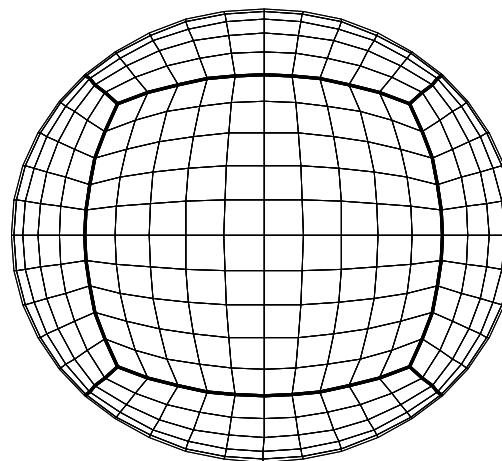


Tracer profile - 2 Gaussian hills

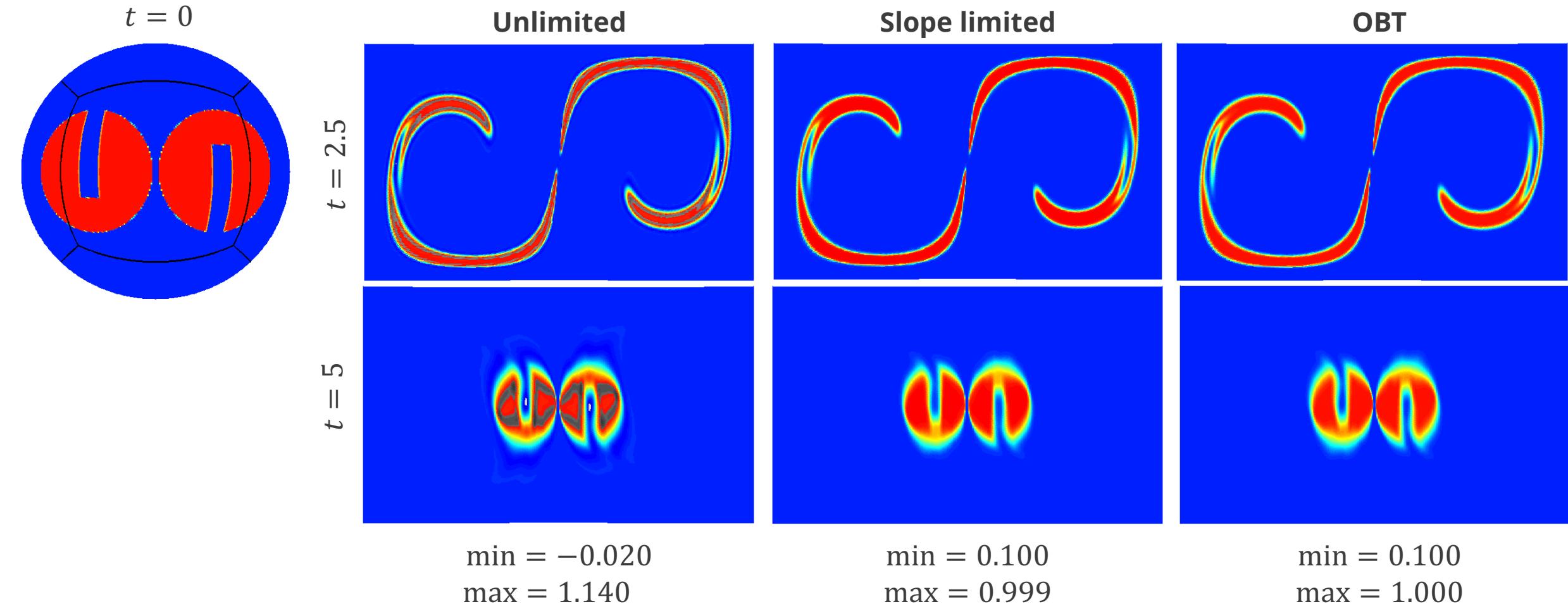
Velocity field* - deformational, div-free field

$$u(\lambda, \theta, t) = 2 \sin^2(\lambda) \sin(2\theta) \cos(\pi t/T)$$

$$v(\lambda, \theta, t) = 2 \sin(2\lambda) \cos(\theta) \cos(\pi t/T)$$



Tracer transport on a sphere: 2 notched cylinders



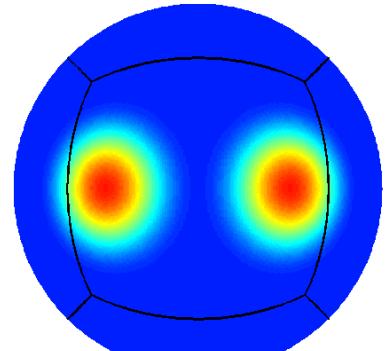
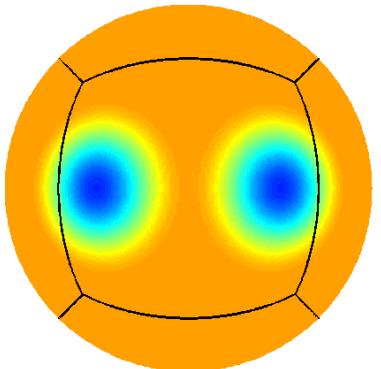
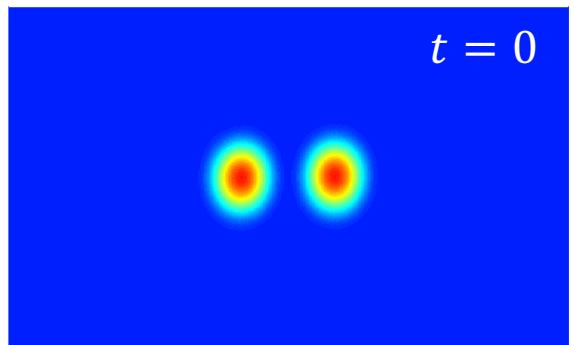
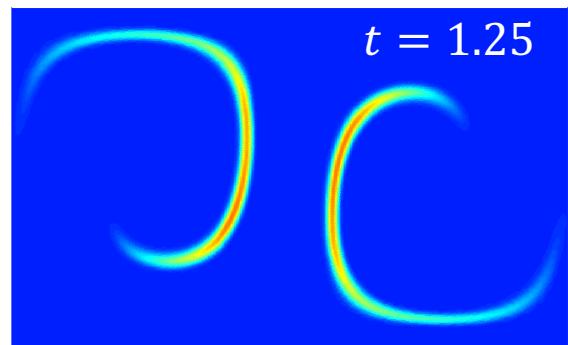
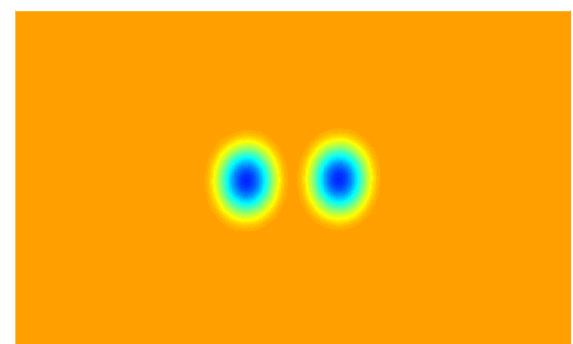
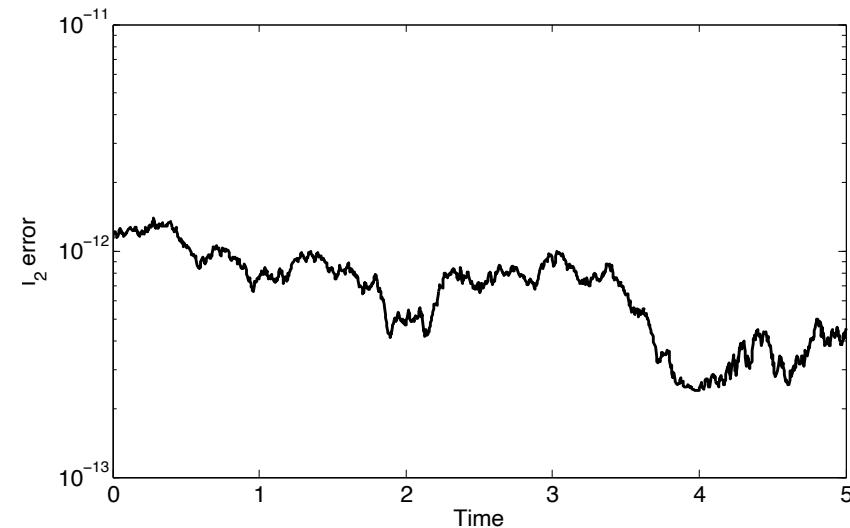
Latitude-longitude projection used for better representation of the results



Tracer transport on a sphere: tracer correlations

Tracers profile: two cosine bells

$$\tau = \begin{cases} 0.5(1 + \cos(\pi r_i/r)) & \text{if } r_i < r, i = 1, 2 \\ 0 & \text{otherwise} \end{cases} \quad \text{and} \quad \xi = -0.8\tau + 0.9$$


 τ

 ξ

 $t = 0$

 $t = 1.25$

 $t = 2.5$


How much does it cost?



Scheme	Mesh	Relative time	ℓ_2 error	Min	Max	Exact max
Unlimited (benchmark)	32×32	1.00	1.958e-01	-0.054	0.340	0.42585
	64×64	1.00	5.954e-02	-0.079	0.402	
	128×128	1.00	1.1132e-02	-0.105	0.423	
	256×256	1.00	1.6020e-03	-0.105	0.426	
Slope Limited	32×32	1.05	2.3030e-01	0	0.281	0.42585
	64×64	1.06	7.3340e-02	0	0.362	
	128×128	1.11	1.5930e-02	0	0.402	
	256×256	1.19	2.9830e-03	0	0.418	
OBT	32×32	1.10	2.0540e-01	0	0.323	0.42585
	64×64	1.08	6.2220e-02	0	0.391	
	128×128	1.07	1.2120e-02	0	0.416	
	256×256	1.14	1.8360e-03	0	0.423	

Red = best result

Extensions: semi-Lagrangian spectral element scheme



Start with a generic SE+SL (SESL) scheme:

1. Determine GL **departure points** $\rightarrow \tilde{\mathbf{p}}_{ij} = \mathbf{x}(t_n)$

2. Determine solution at **arrival points** $\rightarrow \rho_h(\mathbf{p}_{ij}, t_{n+1}) = \rho(t_{n+1})$ and $q_h(\mathbf{p}_{ij}, t_{n+1}) = q(t_{n+1})$

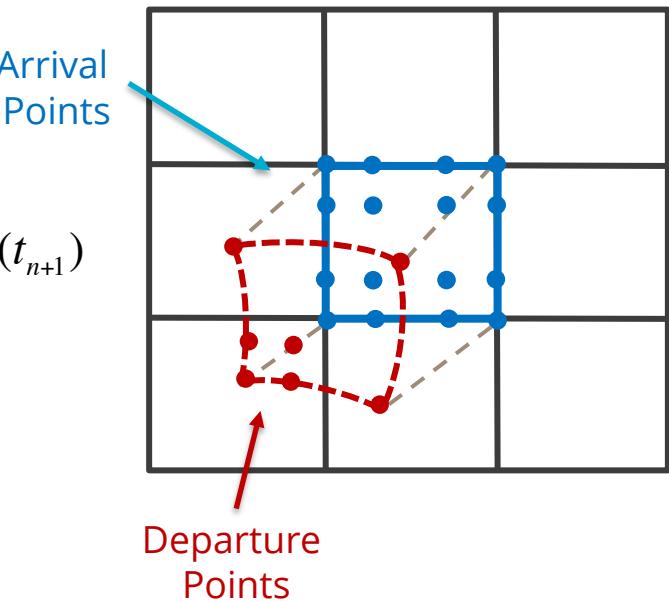
Then proceed as follows to find the tracer at t_{n+1} (density is similar)

3. Set optimization target to SE+SL solution: $\hat{q} := q_h(\mathbf{p}_{ij}, t_{n+1})$

4. Determine local solution bounds: $q_{ij}^{\min} \leq q(\mathbf{p}_{ij}, t_{n+1}) \leq q_{ij}^{\max}$

5. Set solution at the new time step by solving

$$q_{n+1}^* = \underset{q \in Q^r}{\operatorname{argmin}} \|q - \hat{q}\|_0^2 \quad \text{subject to} \quad \begin{cases} \int_{\Omega} q \, dx = \int_{\Omega} q_n \, dx & \leftarrow \text{Conservation} \\ q_{ij}^{\min} \leq q_{ij} \leq q_{ij}^{\max} & \leftarrow \text{Local bounds} \end{cases}$$



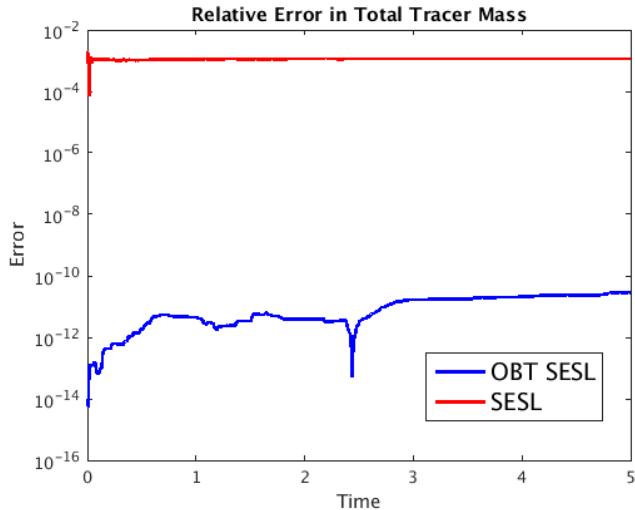
The structure of the optimization problem is identical to the one before!

QP structure admits a fast $O(N)$ optimization algorithm.

Numerical examples



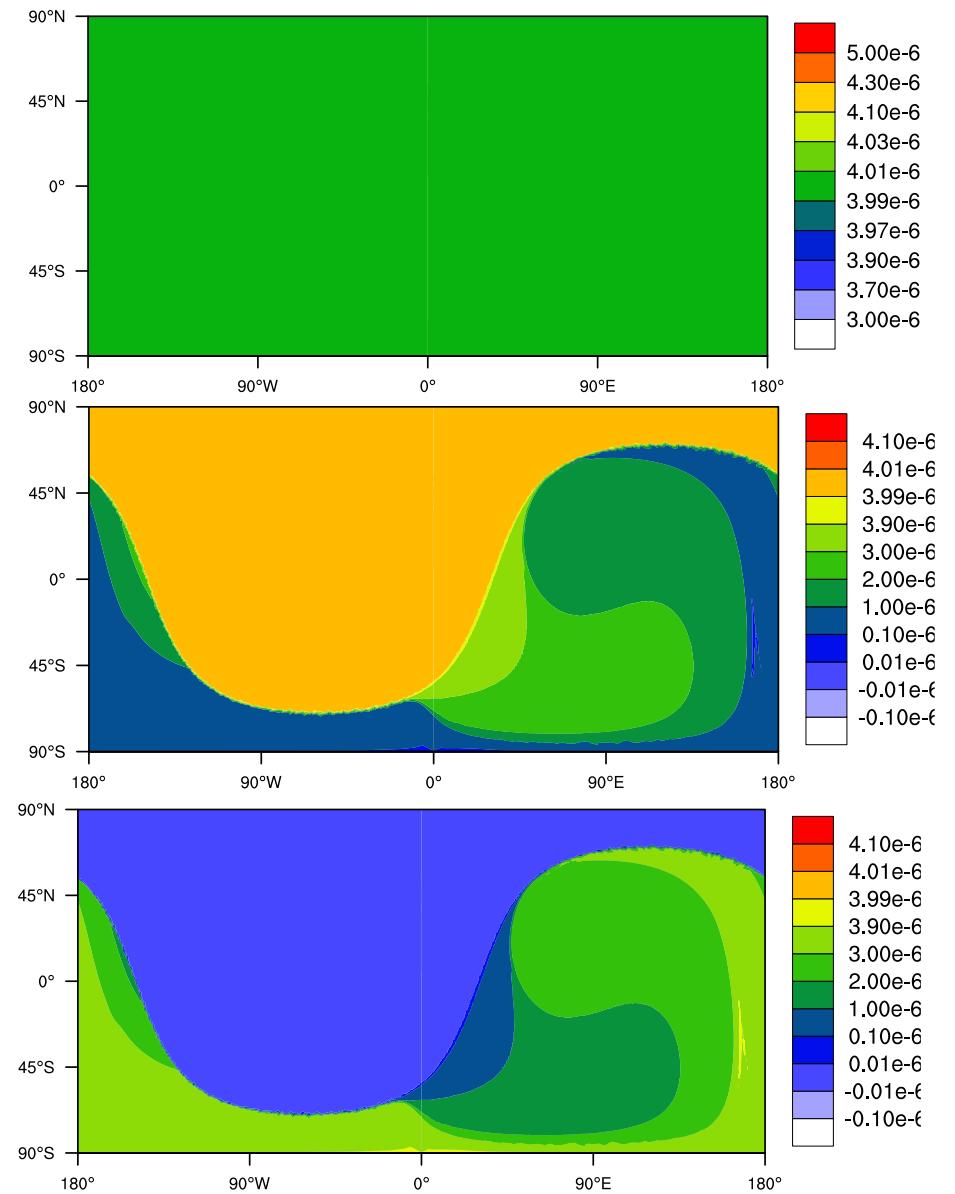
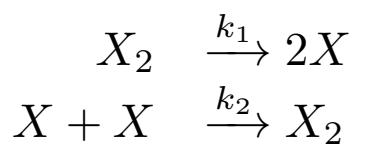
Mass conservation



For the same tracer distribution and flow field, the error in total tracer mass reveals the lack of mass conservation in the underlying SESL scheme and the recovery of conservation by the OBT approach

Linear correlation

OBT SESL performs well on challenging idealized chemistry test where total sum of species should remain constant as long as advection scheme preserves linear relationships.



Conclusions



Optimization-based tracer transport offers a robust and flexible alternative to traditional limiter-based transport techniques.

- Ensures global mass conservation and bounds preservation,
- Provably preserves linear tracer correlations,
- Robust and efficient (cost similar to conventional slope limiters),
- Formulation applicable to finite volume, finite element, and spectral element discretizations.
- Spectral element optimization-based transport has been implemented in the High-Order Method Modeling Environment (HOMME), the code on which DOE's E3SM dynamical core is based.