

# The semi-Lagrangian semi-implicit time stepping scheme in the ECMWF model IFS

Numerical methods for weather prediction training course  
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# Outline of this lecture

- ◆ Motivation and benefits of semi-Lagrangian, semi-implicit methods
- ◆ A detailed overview of semi-Lagrangian advection (definitions, stability and error analysis, algorithmic details, parallel implementation on spherical geometry domains)
- ◆ The semi-implicit time stepping and how it is combined with a semi-Lagrangian method to solve the full set of the ECMWF model prognostic equations
- ◆ Conservation aspects in the IFS time stepping



# The ECMWF hydrostatic global operational model equation set

$$\frac{D\mathbf{V}_h}{Dt} + f\mathbf{k} \times \mathbf{V}_h + \nabla_h \Phi + R_d T_v \nabla_h \ln p = P_v$$

$$\frac{DT}{Dt} - \frac{\kappa T_v \omega}{(1 + (\delta - 1)q) p} = P_T$$

$$\frac{Dq_x}{Dt} = P_{q_x}$$

$$\frac{\partial}{\partial t} \left( \frac{\partial p}{\partial \eta} \right) + \nabla_h \cdot \left( \mathbf{V}_h \frac{\partial p}{\partial \eta} \right) + \frac{\partial}{\partial \eta} \left( \dot{\eta} \frac{\partial p}{\partial \eta} \right) = 0$$

$$\Phi = \Phi_s - \int_1^\eta R_d T_v \frac{\partial}{\partial \eta} (\ln p) d\eta$$

$\eta$  : hybrid (pressure based) vertical coordinate  $\eta \in [0,1]$

$\mathbf{V}_h$ : horizontal momentum

$T$ : temperature

$T_v$ : virtual temperature,  $\kappa = R_d / c_{pd}$ ,  $\delta = c_{pv} / c_{pd}$

$q_x$ : specific humidity, specific ratios for cloud fields and other tracers  $x$

$\Phi$ : geopotential

$p$  : pressure

$\omega = dp/dt$  : diagnostic vertical velocity

$P$ : physics forcing terms

Continuity equation in terms of full (moist) pressure: the model conserves the total (rather than dry) atmospheric mass. Dry mass conserving version research option, ECMF TM 849 2019, Malardel et al.

- Primitive equation hydrostatic
- A **non-hydrostatic** option is available for **research** purposes but not used operationally
- Spectral Transform with spherical harmonics basis
- Cubic spline Finite Elements in the vertical
- Time stepping: semi-Lagrangian semi-implicit

# The critical role of time stepping

For operational global weather forecasting an accurate and robust weather model which operates at the lowest possible cost is essential

- ◆ Role of time stepping scheme is central into achieving this goal
- ◆ Semi-Lagrangian (SL) semi-implicit (SI) method is ideal
  - ◆ Unconditionally stable SL advection scheme with small phase speed errors and little numerical dispersion
    - Large timesteps can be used (no CFL restriction) without accuracy penalty
    - Multi-tracer efficient
  - ◆ Unconditionally stable SI time stepping for the integration of fast forcing terms
    - No timestep restriction from the integration of “fast forcing” terms (e.g. gravity waves and acoustic terms in non-hydrostatic models)
    - 2<sup>nd</sup> order accurate time scheme

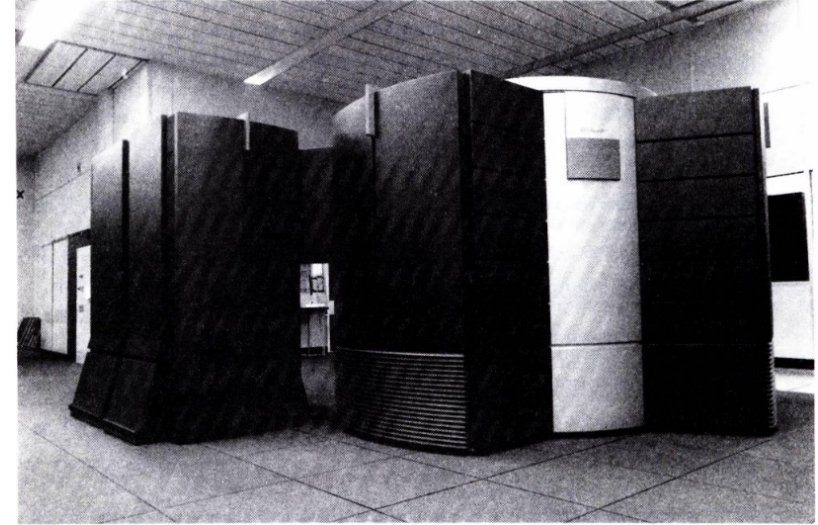


# Benefits of spectral transforms combined with SISL

- ◆ A spectral SISL scheme like the one in IFS offers additional advantages
  - **Derivative terms** in the equations can be computed **extremely accurately** using analytical formulae
  - The common “**pole singularity**” **problem** of regular lat-lon grid-point models **does not apply**
  - The derived **3D Helmholtz** elliptic equation can be **decoupled to a set of small dimension** and **computationally cheap** to solve linear systems thanks to the properties of **Spherical Harmonics** (eigenfunctions of the Laplace operator)
- Advancements in algorithms, hardware and software have enabled us to keep running efficiently the spectral transform method at **ever increasing resolutions**
- ecTrans: a multi-node GPU enabled spectral transform library

# History of SISL method at ECMWF

- 1991: IFS was a **spectral semi-implicit** Eulerian model on a full Gaussian grid at T106 horizontal resolution and 19 levels
  - An increase to T231 L31 resolution was planned
  - This upgrade required at least 12 x available CPU power
  - Funding was available for 4 x CPU increase ...
- Upgrade was made possible by introducing:
  - A **semi-Lagrangian, semi-implicit** scheme on a **reduced Gaussian grid**
  - The new model was 6 x faster!



CRAY Y-MP/8: first HPC to run spectral SISL operationally (1992)

Source: ECMWF newsletter 60, Dec 1992



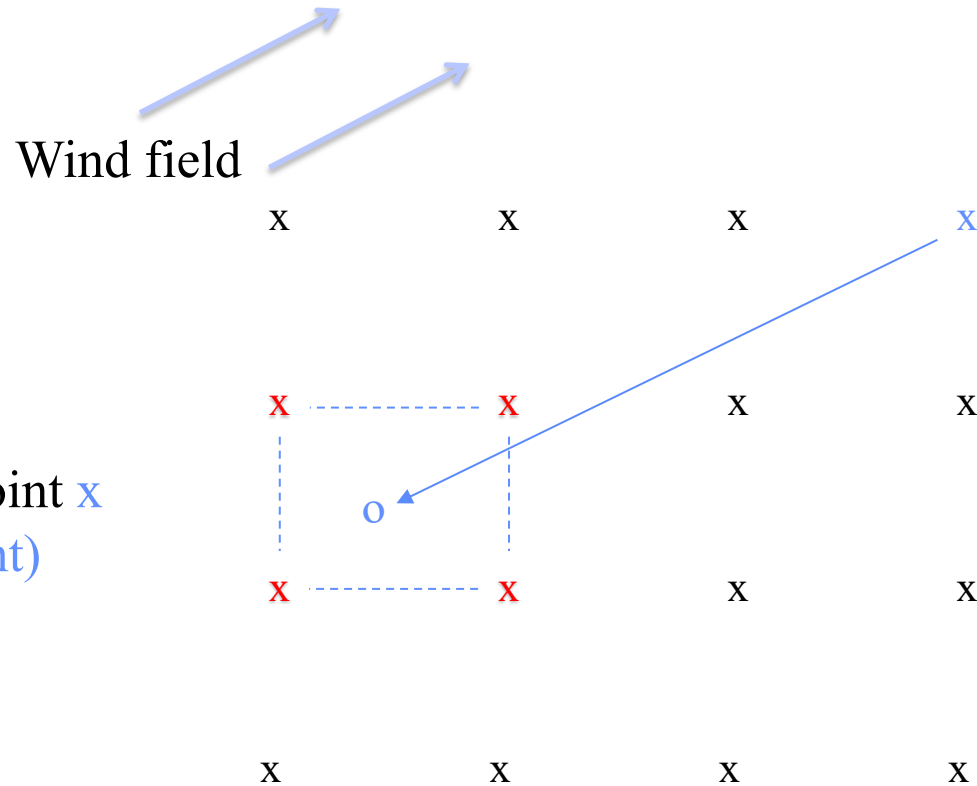
ATOS HPC at Bologna data centre (2022)

# What is a semi-Lagrangian (SL) advection scheme?

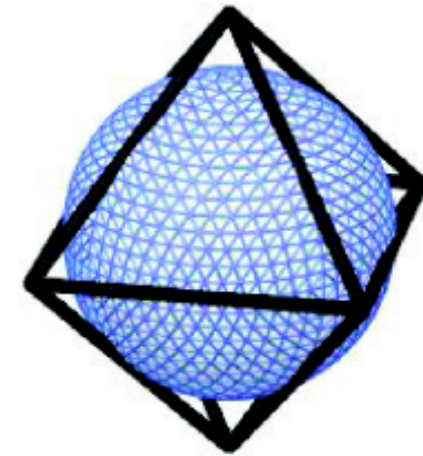
- ◆ Advection (movement of air, moisture, heat, momentum) is a fundamental process in a weather prediction model
- ◆ A SL scheme is a numerical technique for solving advection type PDEs which applies *Lagrangian* “thinking” on grid-point models:
  - ◆ For all discrete fluid elements (parcels) the corresponding upstream points (“backward” trajectories) are computed
    - SL assumes that by the end of each time-step each air parcel arrives at a grid-point location but the location where its trajectory started (departure point) is unknown and must be found.
  - ◆ Gradually evolved from schemes introduced in the '50s, '60s, '70s (Wiin-Nielsen, Krishnamurti, Sawyer, Leith, Purnel)



# Semi-Lagrangian advection in a picture



Octahedral Gaussian grid (see a new grid for IFS ECMWF newsletter 146, Malardel et al) : the grid that IFS uses since cycle 41r2 (2015)

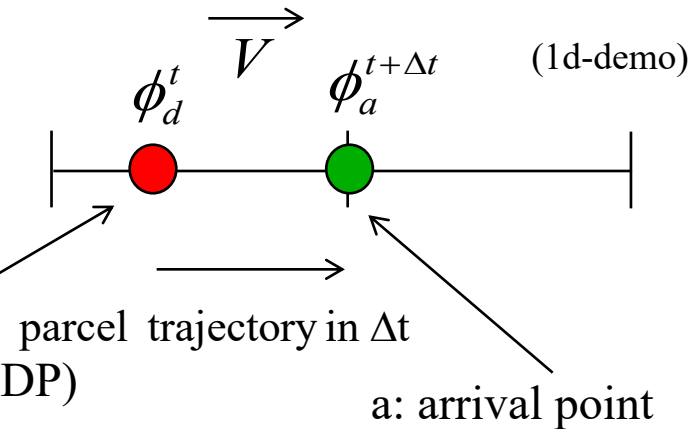


# The SL solution of the advection equation

Passive advection of the tracer mixing ratio  $\phi = \rho_\phi / \rho$  ( $\leftrightarrow$  air density + tracer density advection) in Cartesian coordinates:

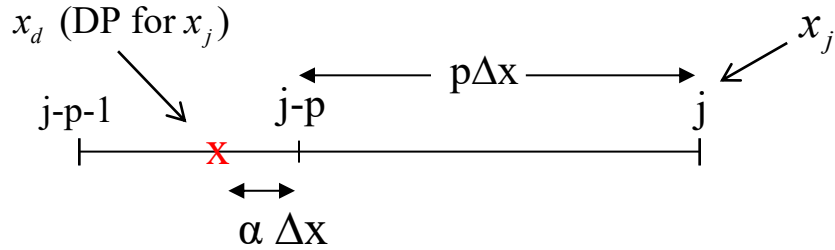
$$\frac{D\phi}{Dt} \equiv \frac{\partial\phi}{\partial t} + V \cdot \nabla\phi = 0, \quad V = (u, v, w)$$

$$\int_{(r_d, t)}^{(r_a, t+\Delta t)} \frac{D\phi}{Dt} Dt = 0 \Rightarrow \phi_a^{t+\Delta t} = \phi_d^t, \quad r = (x, y, z)$$



- ◆ Solution at  $t+\Delta t$  is obtained by finding the DP location and interpolating the available (defined at time  $t$ ) grid-point values of  $\phi$  at the DP
- ◆ Advection term  $V \cdot \nabla\phi$  is not explicitly computed - it is absorbed by the Lagrangian derivative: nonlinear advection problem is reduced to DP search and interpolation!

# Unconditional stability (Bates & McDonald MWR 82)



$$\frac{D\phi}{Dt} \equiv \frac{\partial\phi}{\partial t} + u_0 \frac{\partial\phi}{\partial x} = 0 \quad (\text{constant wind})$$

Departure to arrival pt distance:  $x_j - x_d = u_0 \Delta t = (p + \alpha) \Delta x$   $p$ : integer

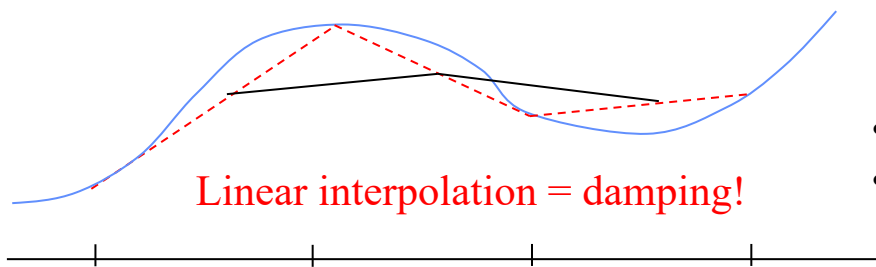
Von Neuman stability analysis:  $\phi_j^n = \phi_0 \lambda^n e^{ikj\Delta x}$

$$\phi_j^{n+1} = \phi_d^n = (1 - \alpha)\phi_{j-p}^n + \alpha\phi_{j-p-1}^n, \quad \alpha = \frac{x_{j-p} - x_d}{\Delta x} \longrightarrow \lambda = [1 - \alpha(1 - e^{-ik\Delta x})] e^{-ipk\Delta x}$$

Amplification factor:  $|\lambda|^2 = 1 - 2\alpha(1 - \alpha)[1 - \cos(k\Delta x)]$

$|\lambda| \leq 1$  if  $0 \leq \alpha \leq 1$   
(interpolation from two nearest points)

NOTE: when  $p=0 \Rightarrow \alpha$  is the CFL number  $\Rightarrow$  SL with linear interpolation is essentially Eulerian upstream differencing!



Linear interpolation = damping!

$$R = \frac{\text{discr freq}}{\text{anal freq}} = \frac{1}{(p + \alpha)k\Delta x} \left[ pk\Delta x + \tan^{-1} \left( \frac{a \sin k\Delta x}{1 - a(1 - \cos k\Delta x)} \right) \right]$$

- No phase error when DP coincides with gridpoint (R=1)
- Small phase error for large CFL

# SL algorithm when winds are constant

Use SL method to solve:

$$\frac{D\phi}{Dt} = 0, \quad V \equiv V_0 = (u_0, v_0, w_0).$$

At the beginning of each step advected variable values  $\phi_j^t$  are available on the computational grid. To compute next time step solution:

- Compute departure point (DP) location:

$$r_{d,j} = r_j - V_0 \Delta t, \quad r = (x, y, z)$$

- Using field values at nearest points surrounding  $r_{d,j}$  interpolate variable  $\phi_j^t$  to obtain solution at future time  $t + \Delta t$  i.e.

$$\phi_j^{t+\Delta t} = \phi_d^t \equiv I(\phi_j^t) \Big|_{r_d}, \quad I: \text{ interpolation operator}$$

Accurate calculation of DP and an accurate interpolation scheme are essential! For accuracy, more sophisticated method required for the DP search in real atmospheric (non-constant wind) flows

# Departure point search in real atmospheric flows: SETTLS

Assume straight line trajectories within a time step. Perform a **2nd order Taylor expansion of an arrival (grid) point to its departure point:**

**Stable Extrapolation Two Time Level Scheme**  
(Hortal, QJRM 2002)

$$r_a(t + \Delta t) = r_d(t) + \Delta t \cdot \left( \frac{Dr}{Dt} \right)_d + \frac{\Delta t^2}{2} \cdot \left( \frac{D^2r}{Dt^2} \right)_{AV} \quad \text{AV: average value along SL trajectory}$$

$$\left( \frac{Dr}{Dt} \right)_d = V_d(t), \quad \left( \frac{D^2r}{Dt^2} \right)_{AV} = \left( \frac{DV}{Dt} \right)_{AV} \approx \frac{V_a(t) - V_d(t - \Delta t)}{\Delta t}$$

Hence,

$$r_a(t + \Delta t) \approx r_d(t) + \frac{\Delta t}{2} \cdot \left( V_a(t) + \{2V(t) - V(t - \Delta t)\}_d \right)$$

Therefore DP can be computed by iterative sequence:

$$r_d^{(0)} = r_a - \Delta t V(r_a, t)$$

$$r_d^{(k)} = r_a - \frac{\Delta t}{2} \cdot \left( V_a(t) + \{2V(t) - V(t - \Delta t)\}_{r_d^{(k-1)}} \right) \quad k = 1, 2, \dots, K$$

Interpolate at  $r_d^{(k-1)}$  In the IFS: K=2 (was 4 until cy48r1)

For convergence of iterative scheme trajectories should not cross. This is equivalent with the Lipschitz condition (see MWR 2016, Diamantakis & Magnusson)

$$\Delta t |\partial V / \partial r| < 1$$

which is less restrictive than CFL for atmospheric flows



# Benefits of SETTLS

- SETTLS has better stability than the classic **2<sup>nd</sup> order mid-point scheme below**:

$$r_d^{(0)} = r_a - \Delta t V(r_a, t)$$

$$r_d^{(k)} = r_a - \Delta t \cdot \left\{ \underbrace{\frac{3}{2} V(t) - \frac{1}{2} V(t - \Delta t)}_{\approx V(t + \Delta t/2)} \right\}_{\frac{r_d^{(k-1)} + r_a}{2}} \quad k = 1, 2, \dots, K$$

“Functional” or “fixed-point” iteration

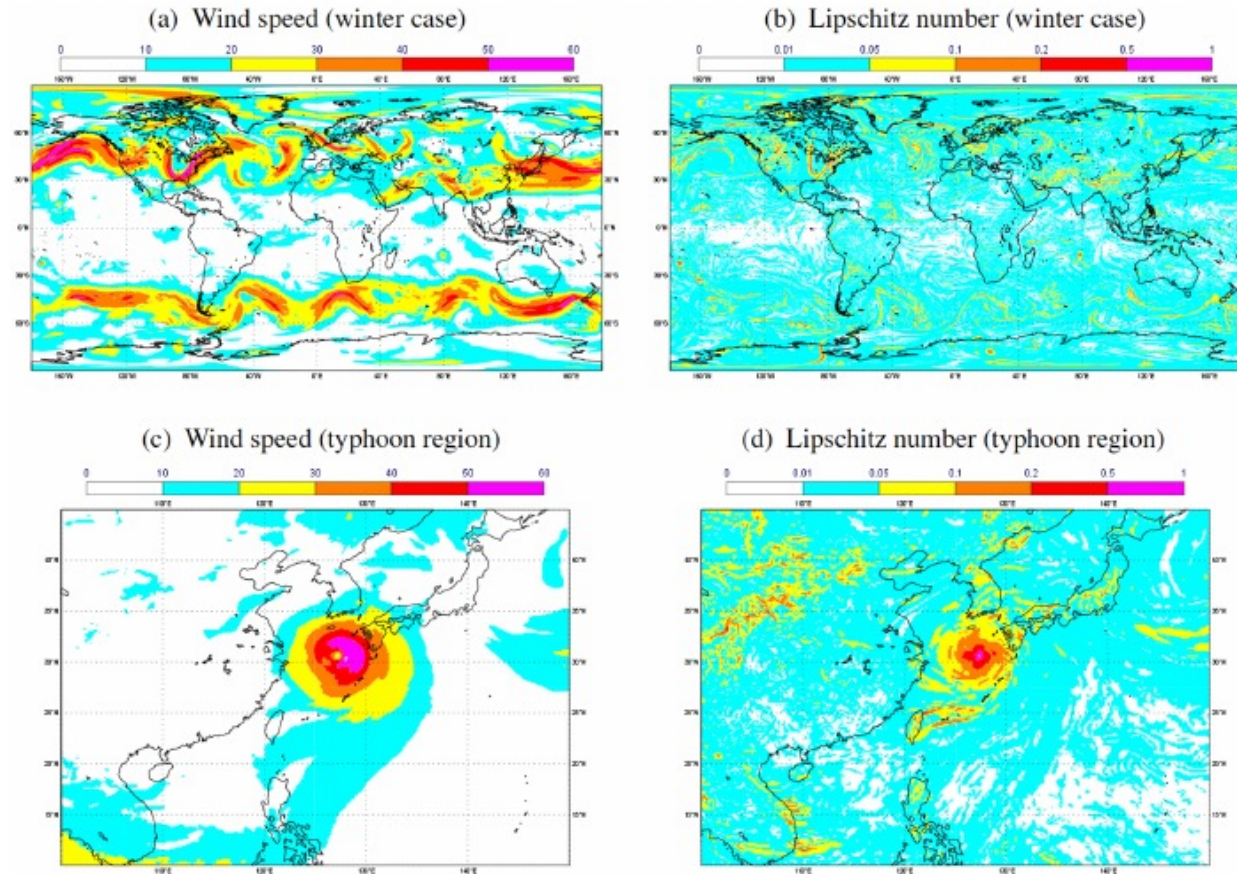
Interpolate to midpoint estimate

- In general, such iterative schemes converge if the Lipschitz number  $< 1$

$$L \equiv \Delta t \left\| \frac{\partial \mathbf{V}}{\partial \mathbf{r}} \right\| \quad \text{Lipschitz (deformational Courant) number}$$

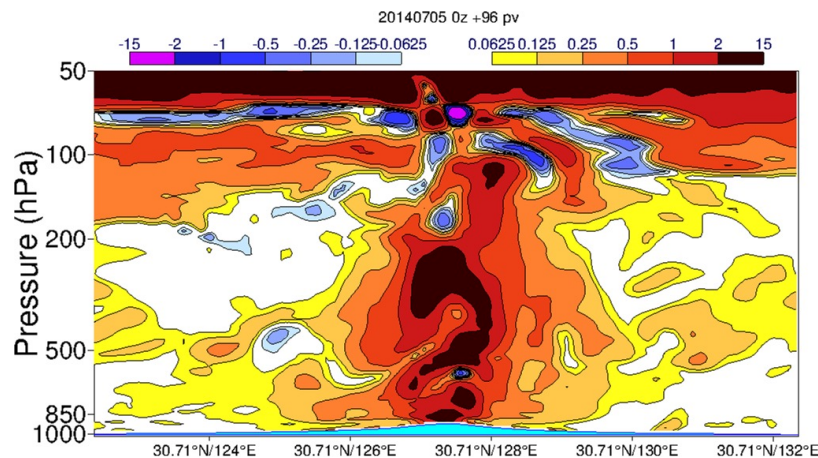
- $L < 1$  is a sufficient condition for convergence
- $L$  is an upper bound of the rate of convergence

# Lipschitz numbers in IFS forecasts

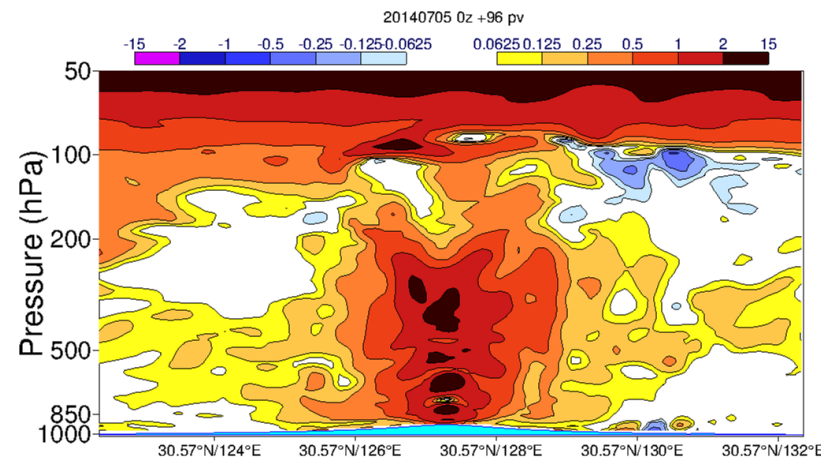


(a), (b): 00UTC 10 January 2014, t+48hrs fc at 500hPa. (c), (d): 00UTC 5 July 2014 t+96 hrs fc at 850hPa

# Side-effects of non-converging DP iterations

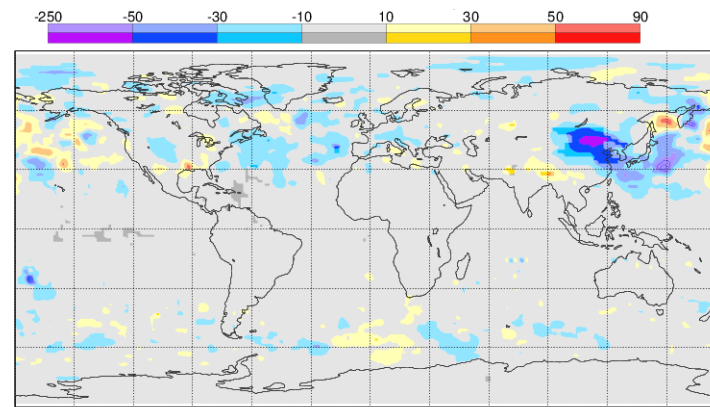


**DP iterations haven't converged: 3 –iterations with old scheme**



**DP iterations have converged: 5-iterations with old scheme or 3-iterations with new scheme**

- **Pre IFS software cycle 48r1: 5 DP iterations** needed for sufficient convergence (Diamantakis & Magnusson, MWR2016 doi:10.1175/MWR-D-15-0432.1)
- **From cycle 48r1: fast convergence in 3 iterations** starting from previous timestep DPs (Diamantakis & Vana, QJRMS 2021, <https://doi.org/10.1002/qj.4224>)



Root Mean Square Error difference for the geopotential height when DP iterations have not sufficiently converged

# SL advection on the sphere in cycle 48r1+

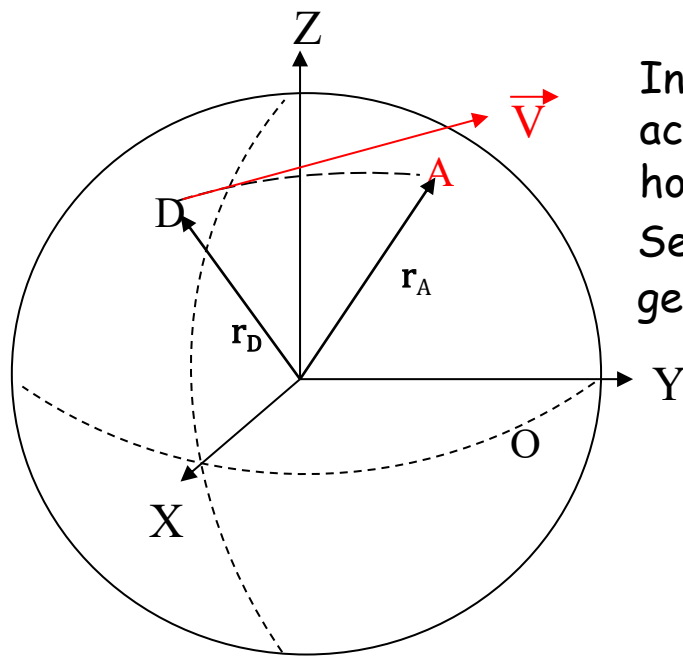
To compute DP on the sphere:

1. Transform horizontal velocities (u,v) in a geocentric Cartesian system (X, Y, Z)
2. Apply SETTLS algorithm to compute  $r_d = (X_d, Y_d, Z_d, \eta_d)$
3. Compute lon/lat of DP from  $(X_d, Y_d, Z_d)$   $\longrightarrow$

$$\lambda_d = ATAN2(Y_d, X_d)$$

$$\theta_d = \arcsin \frac{Z_d}{\sqrt{X_d^2 + Y_d^2 + Z_d^2}}$$

Details of the implementation on the IFS terrain following coordinate in: Diamantakis & Vana QJRMS 2021 10.1002/qj.4224.



In SL transport, vector quantities transported from D to A must be rotated to account for curvature effects: multiply with a "rotation matrix" R the interpolated horizontal wind vector at D

See Temperton et al QJRMS 2001, Staniforth et al QJRMS 2010 (provides general formula independent of  $\varphi = \text{angle } \widehat{DOA}$  between position vectors  $r_A$  and  $r_D$ )

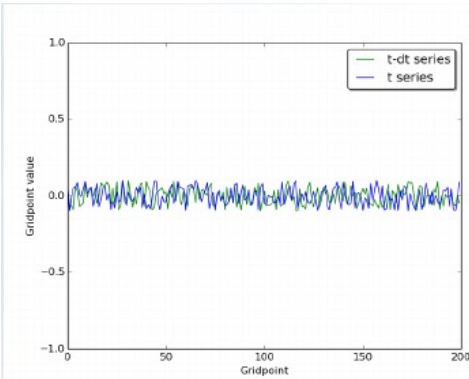
$$\begin{pmatrix} u_A \\ v_A \end{pmatrix} = \underbrace{\begin{pmatrix} p & q \\ -q & p \end{pmatrix}}_{R(V_D): \text{rotation matrix}} \begin{pmatrix} u_D \\ v_D \end{pmatrix}, \quad q = \frac{(\sin \theta_A + \sin \theta_D) \sin(\lambda_A - \lambda_D)}{1 + \cos \varphi}$$

$$p = \frac{\cos \theta_A \cos \theta_D + (1 + \sin \theta_A \sin \theta_D) \cos(\lambda_A - \lambda_D)}{1 + \cos \varphi}$$

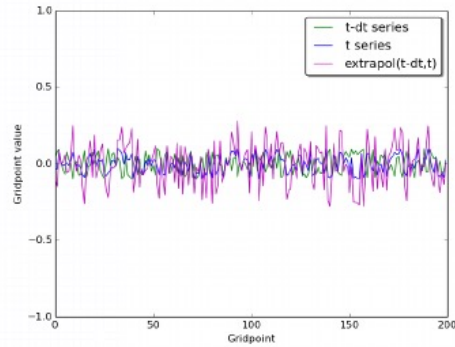
# SETTLS adjustment for stratospheric warming predictions

- In “Sudden Stratospheric Warmings” noise is seen in upper stratosphere and model underpredicts the temperature
- Noise due to **time extrapolation** of vertical velocity in SETTLS
- Solution: use non-extrapolating 1<sup>st</sup> order scheme for gridpoints with sudden changes in vertical velocity in 2 consecutive steps

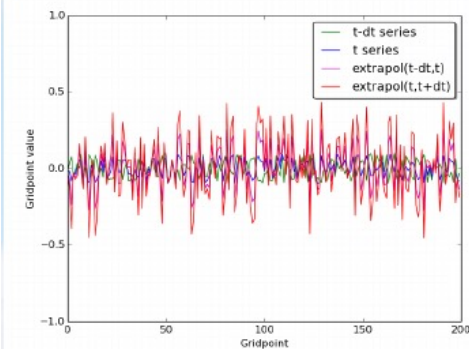
Impact of SETTLS time-extrapolation on noisy and smooth data



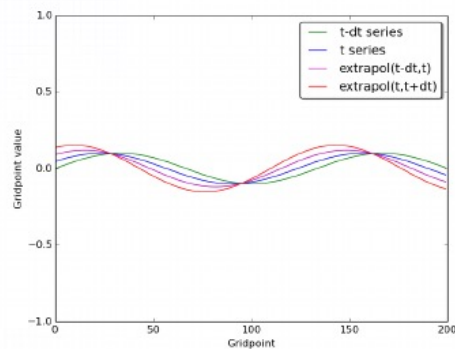
(a) Input t-series



(b)  $w^{ext1} = 2w^t - w^{t-\Delta t}$



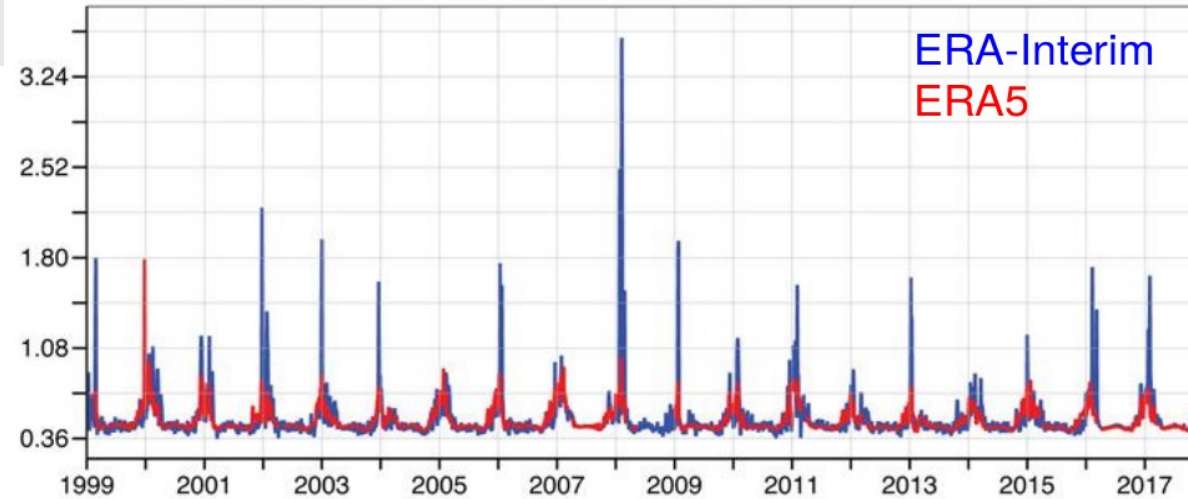
(c)  $w^{ext2} = 2w^{ext1} - w^t$



(d) Smooth data

Much better representation of SSW due to SETTLS improvements

NH winter SSWs

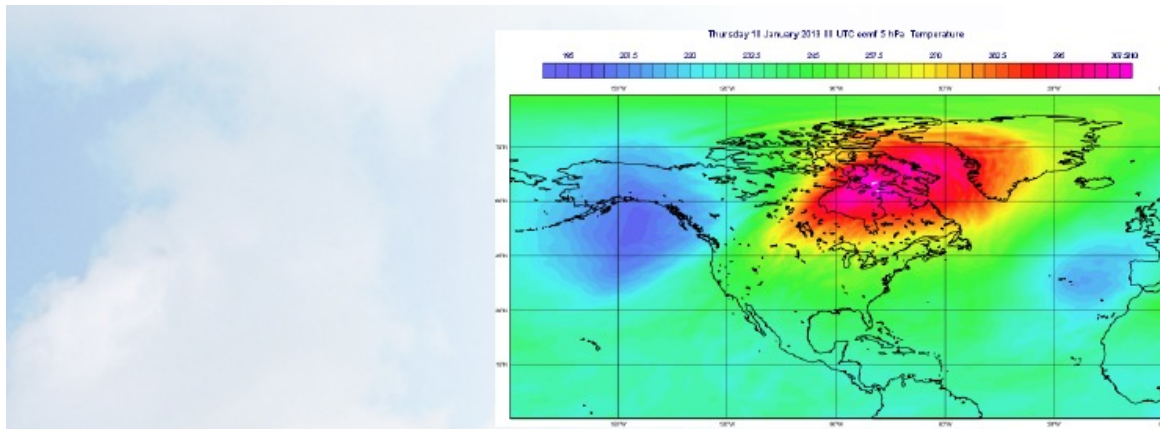


Standard deviation of MW radiances observed vs simulated temperature fields of ERA-Interim (blue) and ERA5 (red) using satellite channel (noaa15) peaking around 5hpa.

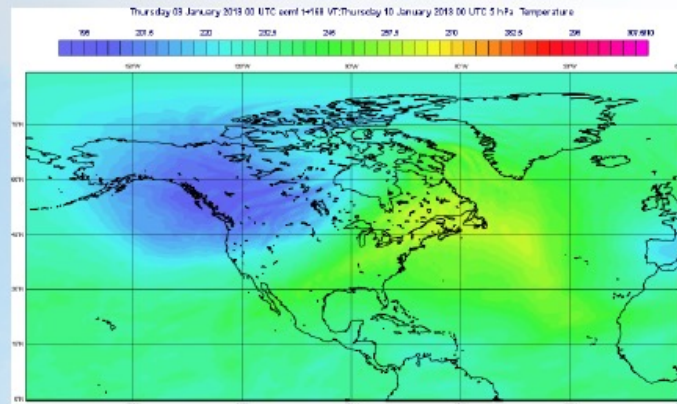
*T. McNally, A. Simmons*

Reference: “Improving ECMWF forecasts of sudden stratospheric warmings”, ECMWF newsletter No.141 Autumn 2014

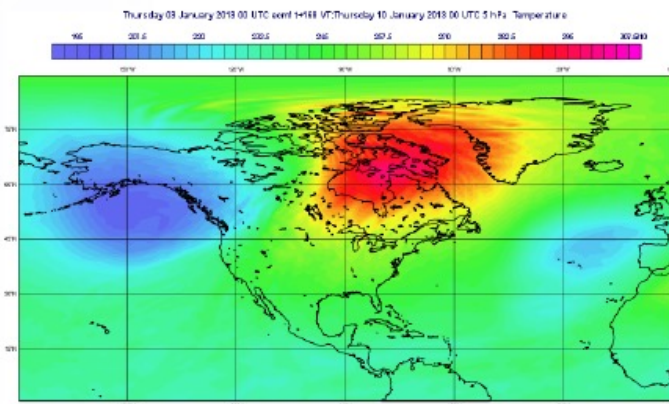
# Major SSW January 2013



(a) Analysis



Original SETTLS t+7day



Improved SETTLS switching off 2<sup>nd</sup> order time-extrapolation in regions of oscillations: t+7days

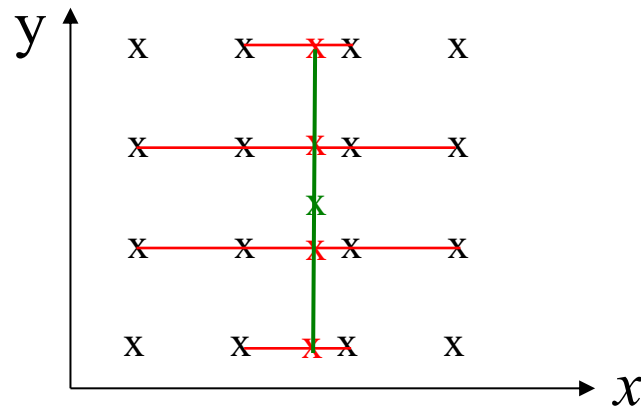
# Interpolation in the IFS semi-Lagrangian scheme

After computing the departure points we need to:

- **Interpolate** the advected field at the **DP** using neighbouring gridpoints
- **Interpolation weights are computed once per time step** and are the same for all advected variables: **multi-tracer efficient scheme**

ECMWF model uses quasi-monotone quasi-cubic Lagrange interpolation (quintic in the vertical for temperature, specific humidity ref Polichtchouk et al ECMWF newsletter 163)

Cubic Lagrange interpolation:  $\phi(x) = \sum_{i=1}^4 w_i(x)\phi_i$ ,  $w_i(x) = \frac{\prod_{k \neq i}^4 (x - x_k)}{\prod_{k \neq i} (x_i - x_k)}$  Faster alternative currently investigated: sweep-quadratic interpolation GMD 17, 2024



## Fully cubic interpolation:

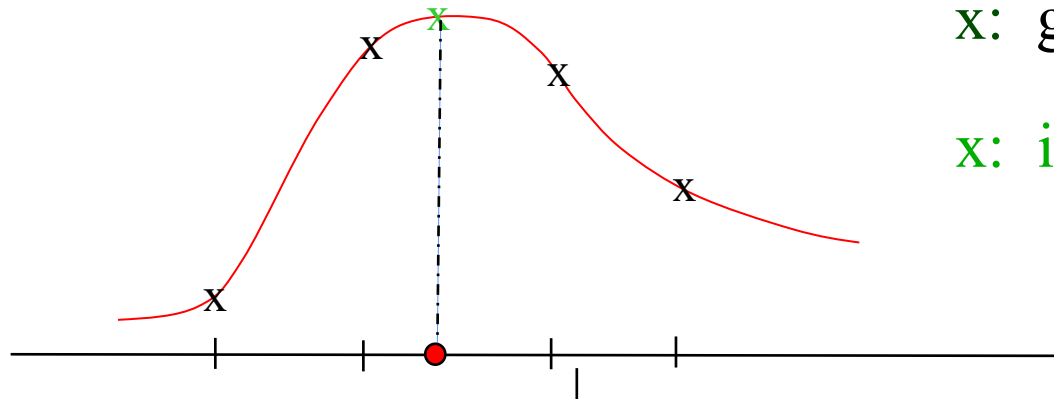
- 5 cubic interpolations in 2D
- 21 in 3D (64pt stencil)

**Quasi-cubic interpolation** reduces computational cost maintaining accuracy:

- **3\*cubic+2\*linear** interpolations in 2D
- **7\*cubic+10\*linear** in 3D (32 pt stencil)
  - Cubic interpolation at nearest to the departure point rows
  - Linear interpolation for remaining rows

# Limiter for shape-preserving (locally monotonic) interpolation

- Creation of “artificial” maxima /minima

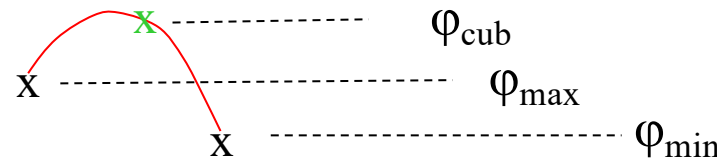


x: grid point values

x: interpolated value

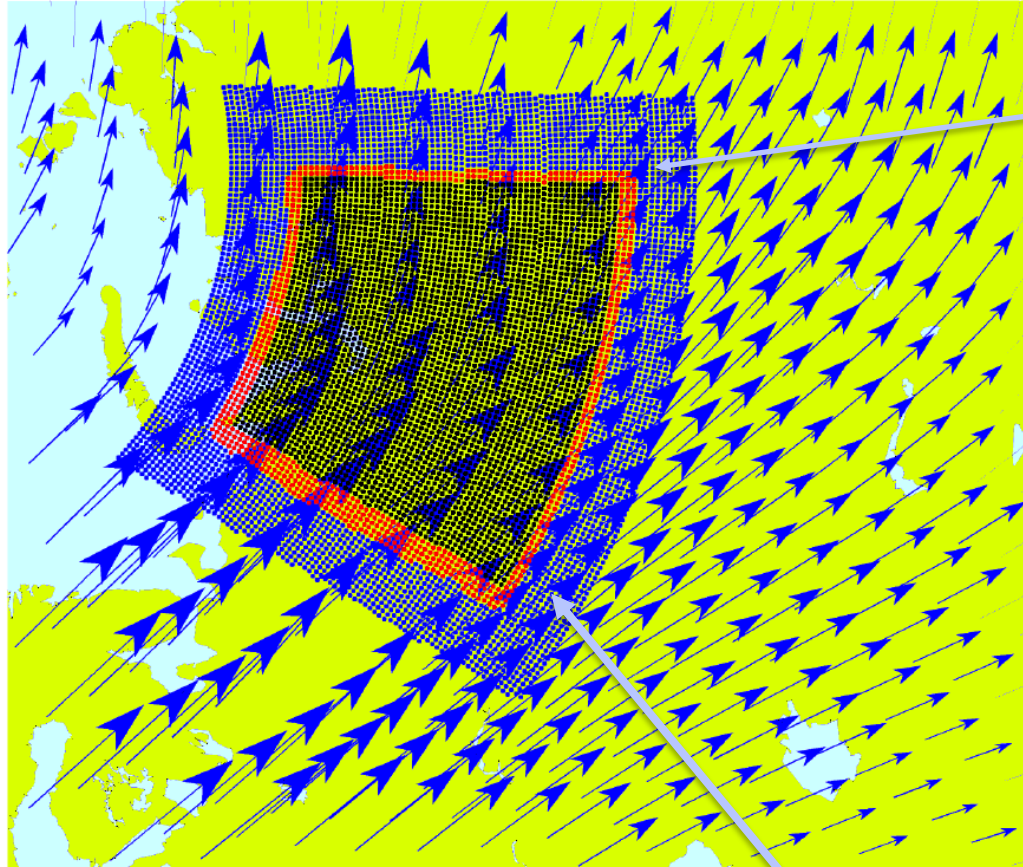
- Shape-preserving (quasi-monotone) cubic interpolation QMSL scheme (Bermejo & Staniforth, MWR 1992)

$$\varphi_{qm} = \max(\varphi_{\min}, \min(\varphi_{\max}, \varphi_{cub}))$$



# Parallel implementation

Interpolation at the DP near the edges of MPI domains requires data from a neighbouring domain (note that DP may lie at a different domain)



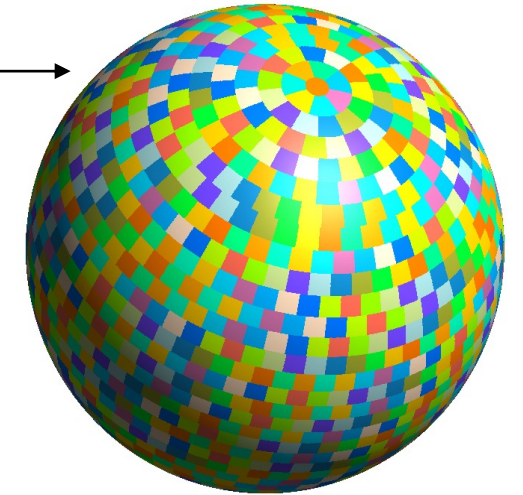
Blue: Halo region

Equal region domain decomposition + MPI and openMP parallel

Halo width for MPI assumes a maximum wind speed larger than the ones observed in the atmosphere e.g. 250m/s

Two levels of communication:

- Entire wind halo filled for the DP iterations
- When the DP is known then only a smaller sub-region around the DP needs to be filled
- No need to fetch data from remote processors at the expense of extra memory use



# Combining semi-Lagrangian with semi-implicit (SI) time stepping

- ◆ A nonlinear system of m-prognostic equations must be solved:

$$\frac{DX}{Dt} = M(X), \quad X = (X_1, X_2, \dots, X_m) \quad \text{e.g. } X=(u,v,T,p,q,\dots)$$

- ◆ Integrate along SL trajectory using a “trapezoidal” 2<sup>nd</sup> order approximation to obtain semi-implicit (Crank-Nicolson) scheme:

$$X^{t+\Delta t} - X_d^t = \int_t^{t+\Delta t} M(X) dt \Rightarrow X^{t+\Delta t} - X_d^t = \frac{\Delta t}{2} (M_d^t + M^{t+\Delta t})$$

Subscript d denotes interpolation at the departure point

Subscript is omitted when variables are at arrival (grid) points

- ◆ Use isothermal reference profiles to linearise the “fast” terms of the right-hand side M and split them to a linear and a residual part:

$$\mathfrak{R} = M - L$$

R: nonlinear residual terms; these are changing slowly and can be integrated explicitly

L: “Fast linearized” (e.g. GW) terms. These must be integrated implicitly for stability

# IFS-SISL for NWP prognostic equations

With splitting in fast linear and slow nonlinear residual terms the two-time-level, 2<sup>nd</sup> order IFS discretization (Temperton et al, QJRMS 2001) becomes:

$$\frac{X^{t+\Delta t} - X_d^t}{\Delta t} = \frac{1}{2} (L_d^t + L^{t+\Delta t}) + \frac{1}{2} \overbrace{(\mathcal{R}_d^{t+\Delta t/2} + \mathcal{R}^{t+\Delta t/2})}^{\text{time-extrapolated nonlinear res}}$$

terms interpolated at the DP

The 2nd right hand side term in brackets is an extrapolation & approximation (space/time) at the trajectory mid-point i.e.  $\approx \mathcal{R}_M^{t+\Delta t/2}$  and can be substituted by the SETTLS expansion:

$$\mathcal{R}_M^{t+\Delta t/2} = \mathcal{R}_d^t + \frac{\Delta t}{2} \left( \frac{d\mathcal{R}}{dt} \right)_{AV} \approx \mathcal{R}_d^t + \frac{\Delta t}{2} \frac{\mathcal{R}^t - \mathcal{R}_d^{t-\Delta t}}{\Delta t}$$

Re-arranging terms, yields the familiar **SETTLS** formula resulting in a 2<sup>nd</sup> order discretization scheme

$$\frac{X^{t+\Delta t} - X_d^t}{\Delta t} = \frac{1}{2} (L_d^t + L^{t+\Delta t}) + \mathcal{R}_M^{t+\Delta t/2}, \quad \mathcal{R}_M^{t+\Delta t/2} = \frac{1}{2} \left( \mathcal{R}^t + \{2\mathcal{R}^t - \mathcal{R}^{t-\Delta t}\}_d \right)$$

all right-hand side terms are given

# Helmholtz equation

- ◆ Elimination of variables in the semi-implicit discretized equations, leads to a single **Helmholtz elliptic equation** in terms of horizontal wind **divergence**
- ◆ Helmholtz equation is solved in **spectral space** at the end of each timestep
- ◆ Constant reference profiles  $\Rightarrow$  constant coefficient **Helmholtz equation**
- ◆ Using spherical Harmonics properties Helmholtz equation can be solved very cheaply with a direct diagonal solver (or 5-diagonal when Coriolis terms are implicit)
  - Having a cheap Helmholtz solver + being able to use large  $\Delta t$  (due to unconditional stability and good dispersion properties of SISL) contributes to high computational efficiency
- ◆ Remaining prognostic variables are computed with back substitution



# Application to the IFS Hydrostatic model

$$\frac{D\mathbf{V}_h}{Dt} + f\mathbf{k} \times \mathbf{V}_h + \nabla_h \Phi + R_d T_v \nabla_h \ln p = P_v$$

$$\frac{DT}{Dt} - \frac{\kappa T_v \omega}{(1 + (\delta - 1)q)p} = P_T, \quad \frac{Dq_x}{Dt} = P_{q_x}$$

$$\frac{D}{Dt} (\ln p_s) = \mathbf{V}_h \cdot \nabla_h (\ln p_s) - \frac{1}{p_s} \int_0^1 \nabla_h \cdot (\mathbf{V}_h \frac{\partial p}{\partial \eta}) d\eta$$

$$\Phi = \Phi_s - \int_1^\eta R_d T_v \frac{\partial}{\partial \eta} (\ln p) d\eta$$

$\eta$  : terrain following (pressure based) vertical coordinate

$\mathbf{V}_h$  : horizontal momentum  $\mathbf{V}_h = (u, v)$

$\nabla_h$  : horizontal gradient

$T_v$  : virtual temperature

$q_x$  : humidity and moist tracers,  $\delta = c_{pv}/c_{pd}$

$\Phi$  : geopotential

$p, p_s$  : pressure, surface pressure

$\omega = dp/dt$  : diagnostic vertical velocity

$P$  : physics forcing terms

Identify fast nonlinear terms of this equation set and linearise equations:

$$\frac{DX}{Dt} = M(X), \quad \mathfrak{R} \equiv M(X) - LX, \quad X = \mathbf{V}_h, T, \ln p_s$$

nonlinear but slow changing

linear but fast changing

# Deriving Helmholtz equation

- For simplicity assume dry dynamics ( $T=T_v$ )
- Also assume that Coriolis terms are incorporated in  $V_h$  (advective form):  $X = V_h + 2\Omega \times r$

$$\frac{X^{t+\Delta t} - X_d^t}{\Delta t} = 0.5(L_d^t + L^{t+\Delta t}) + \mathfrak{R}_M^{t+\Delta t/2}$$

Note: following convention terms without subscript are assumed to be on a grid-point



$$X^{t+\Delta t} - 0.5\Delta t L^{t+\Delta t} = \underbrace{X_d^t + 0.5\Delta t L_d^t + \Delta t \mathfrak{R}_M^{t+\Delta t/2}}_{\equiv X^* \text{ (known part)}}$$

$$L \equiv -\nabla_h(\gamma T + R_d T_{ref} \ln p_s), -\underline{\underline{\tau}} D, -\underline{\underline{\nu}} D \quad \text{for } X = D, T, \ln p_s$$

Linearised terms for different equations

$$\begin{aligned} D^{t+\Delta t} + 0.5\Delta t \nabla_h^2(\gamma T^{t+\Delta t} + R_d T_{ref} \ln p_s^{t+\Delta t}) &= D^* \\ T^{t+\Delta t} + 0.5\Delta t \underline{\underline{\tau}} D^{t+\Delta t} &= T^* \\ \ln p_s^{t+\Delta t} + 0.5\Delta t \underline{\underline{\nu}} D^{t+\Delta t} &= P^* \end{aligned}$$

Momentum equation in terms of divergence  $D$  has been derived by applying the  $\nabla \cdot$  operator in horizontal momentum component discrete equation

$T_{ref}$ : constant temperature reference profile

$\underline{\underline{\gamma}}, \underline{\underline{\tau}}, \underline{\underline{\nu}}$ : operators defined in Ritchie et al MWR vol123, 1995

$D$ : horizontal divergence

$T, \ln p_s$  can now be eliminated deriving a single elliptic equation in terms of  $D$



# Solving Helmholtz equation in spectral space

Prognostic variables are eliminated to derive a Helmholtz equation wrt to D in spectral space:

$$\left( \underline{\underline{I}} - \alpha^2 \Delta t^2 (\underline{\underline{\gamma}} \underline{\underline{\tau}} + R_d T_{ref} \underline{\underline{v}}) \nabla_h^2 \right) D^{t+\Delta t} = RHS \quad \text{RHS contains all known terms (at time t)}$$

[ $\alpha=1/2$  (Crank-Nicolson), using  $\alpha$ -value slightly  $>0.5$  is often used by other models to control unwanted oscillations]

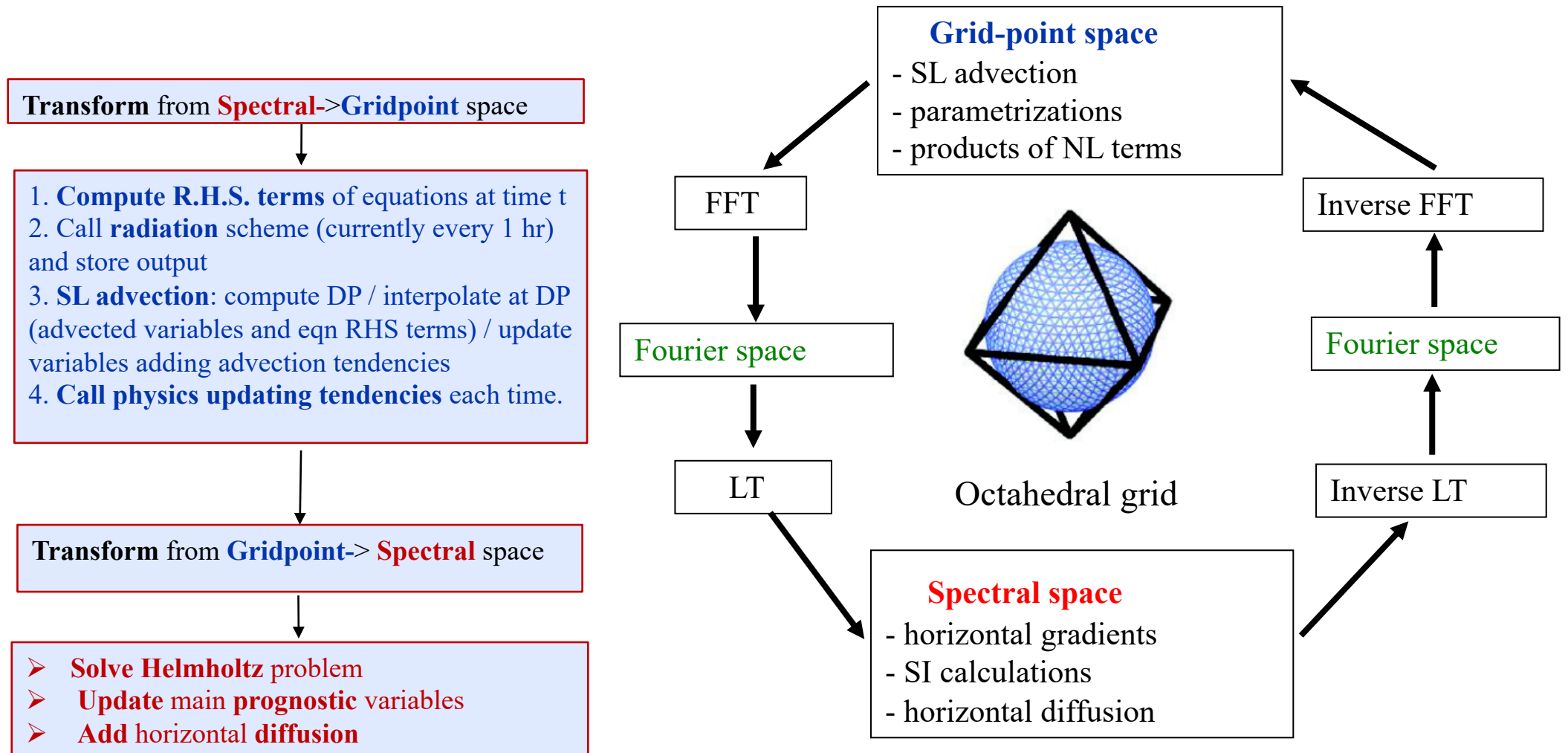
Define: 
$$\underline{\underline{\Gamma}} \equiv \alpha^2 \Delta t^2 (\underline{\underline{\gamma}} \underline{\underline{\tau}} + R_d T_r \underline{\underline{v}}) \implies \left( \underline{\underline{I}} - \underline{\underline{\Gamma}} \nabla_h^2 \right) D^{t+\Delta t} = RHS$$

- $\Gamma$  is constant in time, depends on the vertical discretization and couples all vertical levels
- In a spectral model, the Laplacian operator can be substituted analytically using properties of spherical harmonics:

$$\nabla^2 D_n^m = -\frac{n(n+1)}{r_0} D_n^m \implies \left( \underline{\underline{I}} + \frac{n(n+1)}{r_0^2} \underline{\underline{\Gamma}} \right) D_n^m = B_n^m \quad (\text{B is the RHS – known right hand side})$$

- The above system has a diagonal structure and can be easily solved for each (m,n) => **Computationally Efficient!**
- Once divergence D at new time level is found the remaining fields can be computed through back-substitution

# A simplified overview of IFS time-stepping



FFT: Fast Fourier Transform, LT: Legendre Transform

# A note on SL and its truncation error

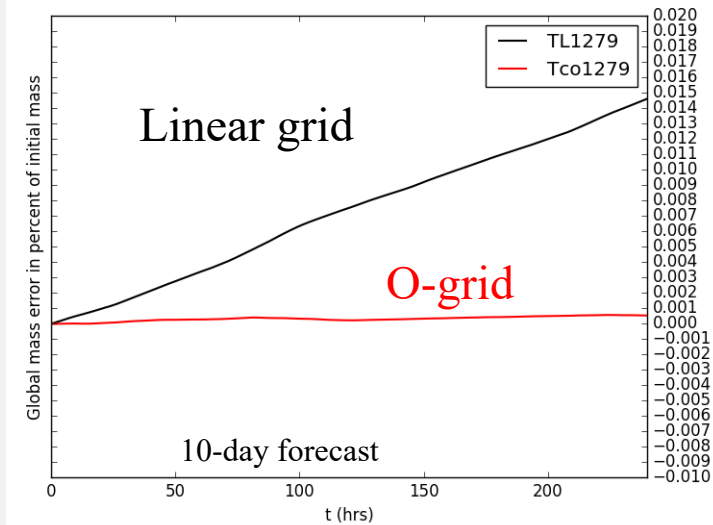
- ◆ Falcone, Feretti SIAM J. 1998: *the leading order truncation error term for a SL method applied to a 1D constant wind advection equation with an interpolation formula of order  $p$  on a grid with constant spacing  $\Delta x$  and an order  $k$  time-integration method with timestep  $\Delta t$  for the DP is  $O(\Delta t^k + (\Delta x)^{p+1}/\Delta t)$*
- ◆ Resolution refinement and timestep reduction should be applied simultaneously rather than separately to improve accuracy
  - ◆ small  $\Delta t$  improves the accuracy of the DP calculation
  - ◆ However, with unnecessarily short  $\Delta t$  too many interpolations and therefore more diffusion from them
  - ◆ Smaller  $\Delta x$  reduces spatial truncation errors
  - ◆  $\Delta t, \Delta x$  ratio must be adjusted together to optimize accuracy



# Mass conservation in semi-Lagrangian advection

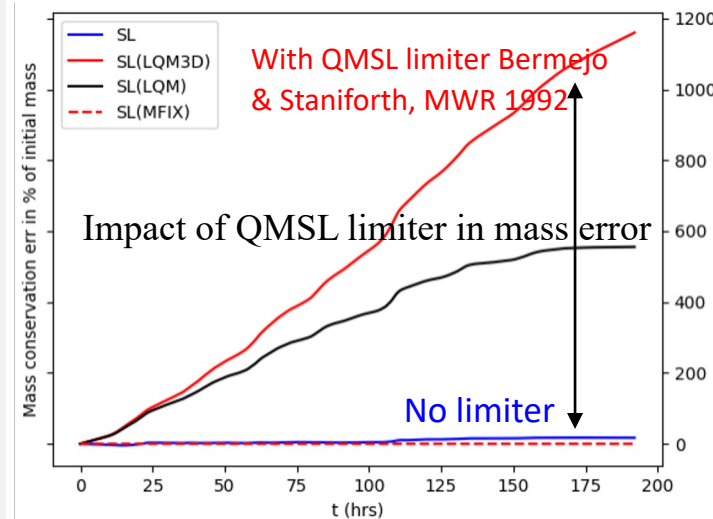
Mass conservation: important for atmospheric composition forecasts, for long range forecasts, climate and overall for high-resolutions

- Semi-Lagrangian time-stepping does not conserve mass, energy, momentum
- Why? (i) continuity expressed in non-conservation form (ii) interpolation errors
- Mass conservation errors depend on the characteristics of the transported variable:
  - Small for smooth tracers and total mass of air
  - Large for localised tracers with large gradients: **monotone interpolation limiters amplify** cons errors!
  - Tracers flowing near the surface exhibit larger conservation errors (stronger vertical advection and boundary condition influence)



With O-grid **total air** mass conservation error is very small in double precision

Mass errors as percent of initial mass

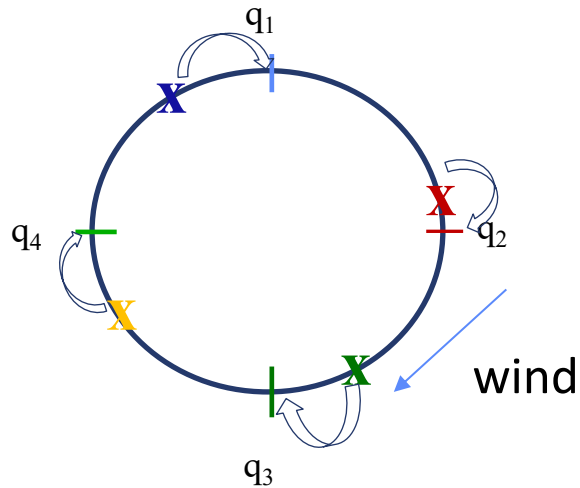


Case study: artificial **discontinuous** tracer 4x5 degrees rectangle placed on the near surface level near Shanghai:

- Large mass conservation error growth in time
- Monotone limiter greatly amplifies those

# Why SL does not conserve? A simple 1D demo

$$q = \frac{\rho_t}{\rho}, \quad \Delta x = \text{const}, \quad m = q\rho\Delta x$$



x: a departure point associated with grid-point

For a uniform grid and non-divergent constant wind flow the SL method without limiter can conserve a tracer mass if the background air density does not change but not in the general case

**Case A: constant wind on uniform grid** (departure point distances from grid-points are equal which results into constant interpolation SL weights on each grid cell)

$$q_j^{n+1} = (1 - \alpha) q_j^n + \alpha q_{j-1}^n, \quad q_0 = q_N, \quad N=4$$

$$M^{n+1} = \sum_{j=1}^N q_j^{n+1} \rho \Delta x = \sum_{j=1}^N [(1 - \alpha) q_j^n + \alpha q_{j-1}^n] \rho \Delta x = \sum_{j=1}^N q_j^n \rho \Delta x = M^n$$

**Case B: non-constant wind or non-uniform grid** (variable departure point distances)

$$q_j^{n+1} = (1 - \alpha_j) q_j^n + \alpha_j q_{j-1}^n, \quad q_0 = q_N$$

$$M^{n+1} = \sum_{j=1}^N q_j^{n+1} \rho \Delta x = \sum_{j=1}^N [(1 - \alpha_j) q_j^n + \alpha_j q_{j-1}^n] \rho \Delta x$$

**Conservation error**

$$= \sum_{j=1}^N q_j^n \rho \Delta x + \sum_{j=1}^N [\alpha_j (q_j^n - q_{j-1}^n)] \rho \Delta x = M^n + \sum_{j=1}^N [\alpha_j (q_j^n - q_{j-1}^n)] \rho \Delta x$$



# Enforcing conservation with mass fixers

- A simple mass fixer (rescaling) is applied on surface pressure field to keep air mass constant in time
- A tracer mass fixer is also applied on water tracers, GHG gases, aerosols
  - The tracer mass fixer used is a locally weighted scheme (ECMWF TM 819, 2017 Diamantakis & Agusti-Panareda, scheme based on Bermejo & Conde MWR 2002) which gives more skilful tracer concentration predictions apart of correcting their global mass error

$$\phi_{jk} = \phi_{jk}^{adv} - \lambda w_{jk}, \quad \lambda = \frac{\delta M}{\underbrace{\sum_j A_j \sum_k w_{jk} \frac{\Delta p_{jk}^{adv}}{g}}_{\text{mass integral}}}, \quad \delta M = M(\phi_{\chi}^{adv}) - M(\phi_{\chi}^n)$$

Corrected tracer mixing ratio →  $\phi_{jk}$

Tracer mixing ratio after advection →  $\phi_{jk}^{adv}$

Lagrange multiplier →  $\lambda$

$\delta M$ : mass conservation error in a timestep after SL advection

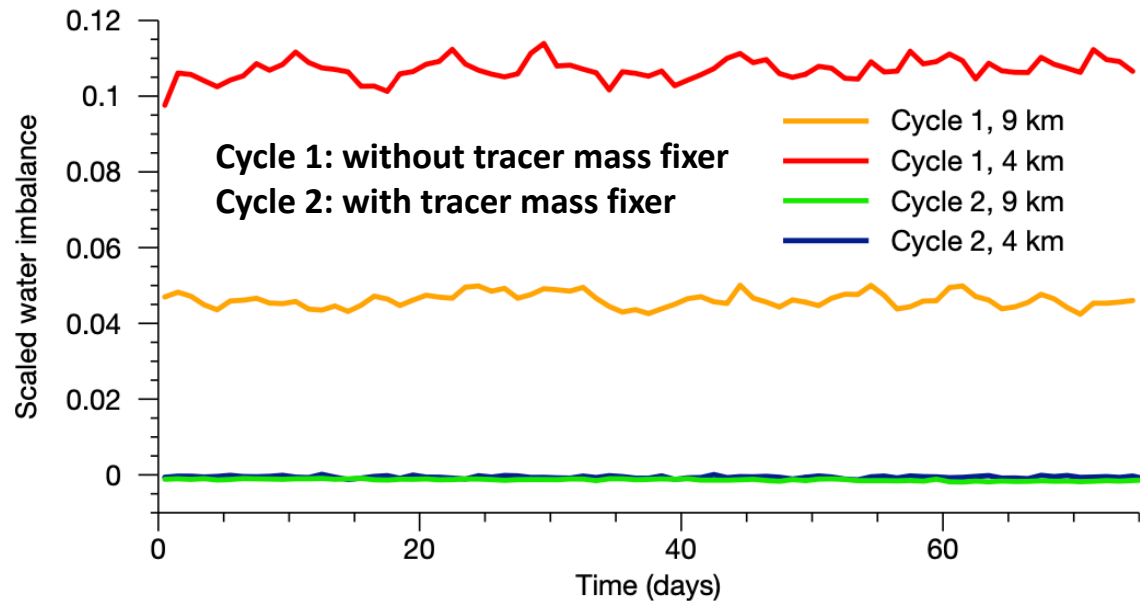
$M$  total mass for tracer  $\phi$

$w_{jk}$  is a weight that depends on the sign of  $\delta M$ , it is proportional to the interpolation truncation error and the mass content of grid-box that corresponds to  $jk$

*Correction computed by the mass fixer is the solution of a constrained optimization problem that ensures that its global norm is minimized subject to the constraint that global mass remains constant*

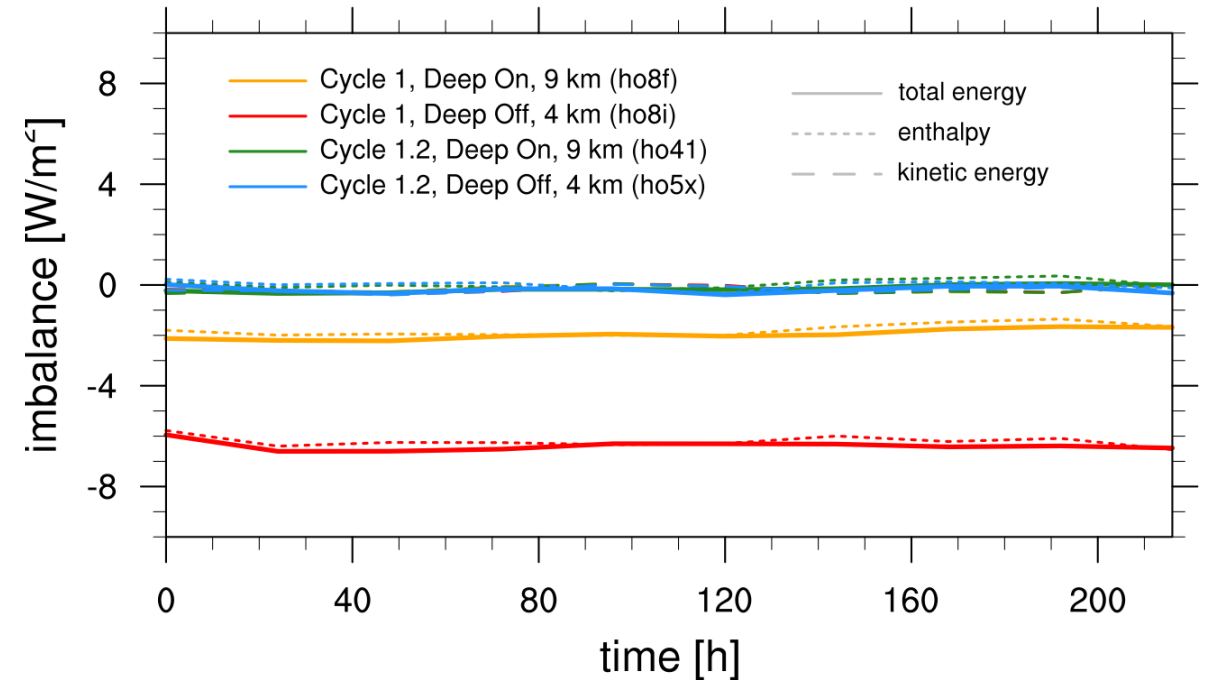
# Fixing water leakage in IFS

Mass fixer on moist tracers (humidity, clouds): improvement in precipitation scores and overall skill of ENS forecasts



**Total water conservation error as a fraction of total precipitation in long integrations**

- 10% surplus is reduced to nearly 0% with tracer mass fixer



**Total Energy leakage reduction with fixer:**

- 2 W/m<sup>2</sup> -> -0.15 (deep conv on)
- 6 W/m<sup>2</sup> -> -0.32 (deep conv off)

Reference: ECMWF newsletter 172, p14

Plots and diagnostics by Tobias Becker from nextGEMS project runs

- HE project supporting CO2MVS (inversions of greenhouse gases)
- **Focus on numerical schemes + evaluation**
  - Development of case studies & diagnostics to understand origin & evolution of mass conservation errors in IFS SL advection
  - Design / implementation of improved algorithms and analysis of their impact on conservation & transport accuracy

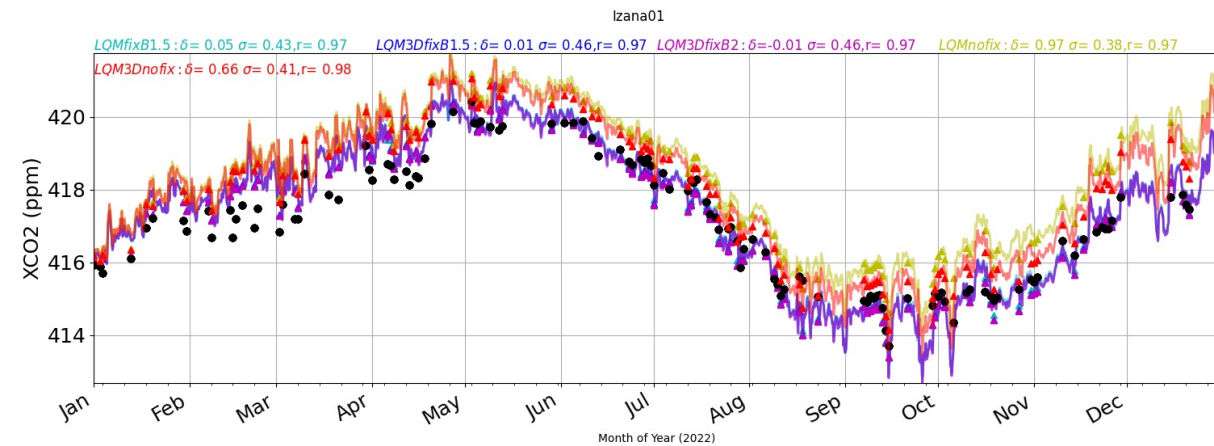
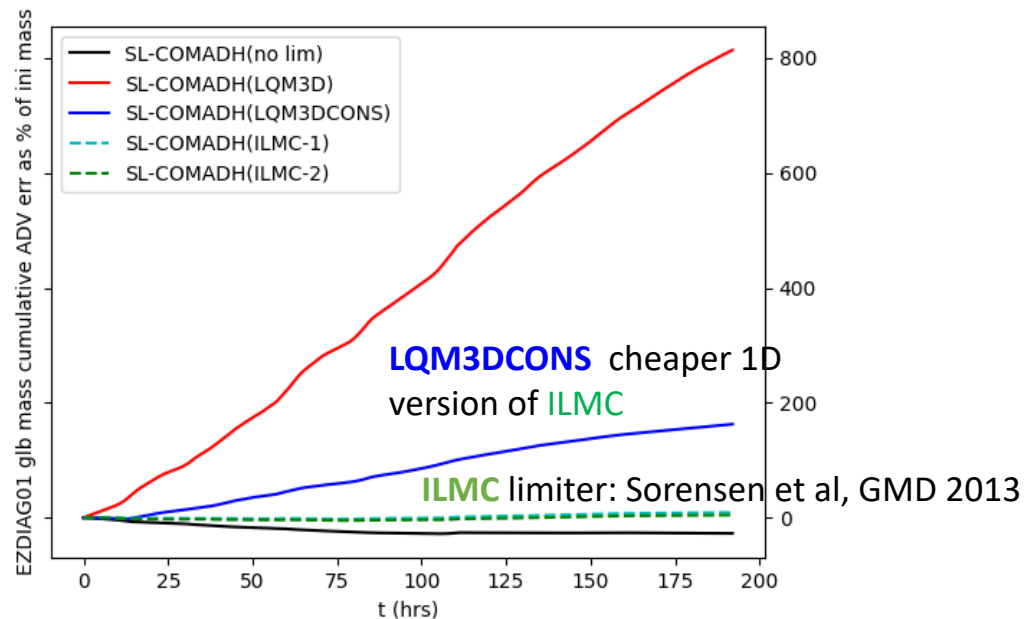
**Project reports with detailed information on SL advection transport scheme evaluation and improvements:**

<https://www.catrine-project.eu/sites/default/files/2025-04/CATRINE-D1-1-V1.1.pdf>

<https://www.catrine-project.eu/sites/default/files/2025-06/CATRINE-D1-2-V1.0.pdf>

## Case study: artificial tracer initialized at the near surface level

Reduction of conservation error by  $\sim O(100)$  due to **conservative limiter**



Simulated time-series of CO<sub>2</sub> concentration versus observations at Izana site using CATRINE optimized fluxes (plot courtesy of A. Agusti-Panareda). **Conservation improves CO<sub>2</sub> simulation accuracy.** Mass fixer (purple, blue) compared to run without mass fixer (yellow, red)

# Some references cited here and further relevant ones

- ◆ Diamantakis & Vana (QJRMS 2021): "A fast converging and concise algorithm for computing the departure points in semi-Lagrangian weather and climate models"
- ◆ ECMWF Tech Memo 819 2017: "A positive definite tracer mass fixer for high-resolution weather and atmospheric composition forecasts"
- ◆ Diamantakis & Magnusson (MWR 2016): "Sensitivity of the ECMWF Model to Semi-Lagrangian Departure Point Iterations "
- ◆ S. Fletcher book: *Semi-Lagrangian Advection Methods and Their Applications in Geoscience* (2020)
- ◆ Hortal (QJRMS 2002): "The development and testing of a new two-time level semi-Lagrangian scheme (SETTLS) in the ECMWF model"
- ◆ Ritchie et al (MWR 1995): "Implementation of the Semi-Lagrangian Method in a High-Resolution version of the ECMWF forecast model"
- ◆ Staniforth & Cote (MWR 1990): "Semi Lagrangian schemes for Atmospheric models"
- ◆ Temperton, Hortal, Simmons (QJRMS 2001): "A two-time-level SL global spectral model"

