

# Automatic Differentiation: A Tool For Data Assimilation and Sensitivity Analysis in Oceanography

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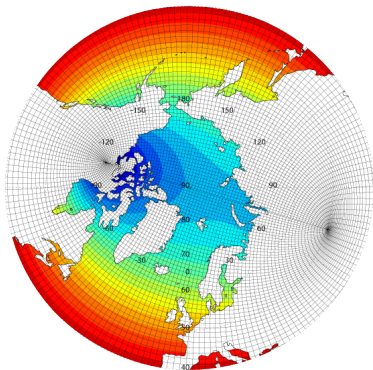
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# Data Assimilation in Oceanography

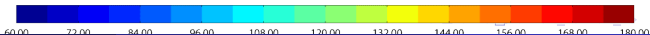
- Model  $\mathcal{M}$  describing the distribution and evolution in space and time of the characteristics of the sea- State Variables  $Y$ -(velocity, temperature, pressure, ..)
- $\mathcal{M}$  depends, among others, on the initial conditions  $Y_0 = X$ .
- Observations : In-Situ, Spatial
- Data assimilation = estimating initial conditions (in our context)

# Direct Model: OPA/NEMO

- Developed at LODYC-LOCEAN-Paris VI
- General ocean circulation model
- Configuration : ORCA 2° :  $i \times j \times k = 180 \times 149 \times 31$



Champ en couleur ( ): Min= 54.98, Max= 171.59, Int= 6.00



# Optimal control theory summary

- Direct model

The system state equation can be given by

$$\begin{cases} y = \mathcal{M}(x) \\ y(0) = x \end{cases}$$

- Cost function:

$$2\mathcal{J}(x) = 2\mathcal{J}^0(x) + 2\mathcal{J}^b(x) = \|\mathcal{H}(y(x)) - y^0\|^2 + \|x - x^b\|^2$$

$\mathcal{H}$  : observation operator  $y^0$  : observations  $x^b$  : background control

- minimization of  $\mathcal{J}(x)$  with respect to the control vector  $x$  using gradient based algorithm

- Tangent linear approximation

$$\mathcal{M}(x + \delta x) = \mathcal{M}(x) + M\delta x, \quad \delta x = x - x^b$$

$$\mathcal{H}(x + \delta x) = \mathcal{H}(x) + M\delta x, \quad \delta x = x - x^b$$

- Cost function:

$$2\mathcal{J}(\delta x) = \|H.M\delta x - d\|^2 + \|\delta x\|^2$$

$$d : y^o - \mathcal{H}(\mathcal{M}(x))$$

- minimization of  $\mathcal{J}(\delta x)$  with respect to the control vector  $\delta x$  using gradient based algorithm
- Quadratic cost function and shorter control vector. Nonlinearities taken by updating  $x^b$ .

# Tangent and Adjoint Model Development for OPA

- Gradient based algorithm
- 4D-Var:

$$\mathcal{J}(x) \longrightarrow \mathcal{M}$$

$$\nabla \mathcal{J}(x) \longrightarrow M^T$$

- Incremental 4D-Var:

$$\mathcal{J}(\delta x) \longrightarrow M$$

$$\nabla \mathcal{J}(\delta x) \longrightarrow M \text{ and } M^T$$

$\mathcal{M}$  : Model OPA

$M$  : Tangent Linear Model of OPA

$M^T$  : Adjoint Model of OPA

# NEMO Tangent and Adjoint Model "NEMOTAM"

## History

- 1992-94: 1st version developed by E. Greiner (LODYC) for OPA4.
  - Applied to 4D-Var with a tropical Atlantic configuration.
- 1995-96: Major rewrite for OPA7 by F. Van den Berghe (CETIIS) and A. Weaver (LODYC).
  - This version was never exploited scientifically.
- 1997-2001: Adapted to OPA8.0 - 8.1 by A. Weaver (LODYC-CERFACS).
  - Developed initially for 4D-Var with a tropical Pacific configuration (TDH).
  - Widely used, primarily for 4D-Var studies.
  - Applied to applications other than data assimilation (singular vectors / optimal perturbations).
- 2002-present: Developed for OPA8.2, free-surface version, by A. Weaver (CERFACS) and C. Deltel (LOCEAN).
  - 1st global ocean version (ORCA2j).
  - Used for 4D-Var in the ENACT project.
  - Currently used by several groups for a variety of applications.

- Hand written tangent and adjoint model
- OPA 9.0 /NEMO: Major new version in Fortran 95.
- Development for OPA 9.0/NEMO using Automatic Differentiation.



# Automatic Differentiation (AD) of Computer Programs

- Every programming language provides a limited number of elementary mathematical functions
- computer program, no matter how complicated, may be viewed as the composition of these so-called intrinsic functions

$$P = \{f_1; f_2; \dots; f_{p-1}; f_p\} \text{ implement } F = f_p \circ f_{p-1} \circ \dots \circ f_1$$

- Derivatives for the intrinsic functions are combined using the chaine rule

$$F'(x_0 = x) = f'_p(x_{p-1}) \cdot f'_{p-1}(x_{p-2}) \cdots f'_1(x_0); x_i = f_i(x_{i-1})$$

# Automatic Differentiation (AD) of Computer Programs

- But calculating and multiplying jacobians is too expensive

$$F'(x_0 = x) = f'_p(x_{p-1}) \cdot f'_{p-1}(x_{p-2}) \cdots f'_1(x_0)$$

- Tangent mode

$$y = F'(x) \cdot \dot{x} = f'_p(x_{p-1}) \cdot f'_{p-1}(x_{p-2}) \cdots f'_1(x_0) \cdot \dot{x}$$

- Reverse mode

$$\bar{x} = F'^T(x) \cdot \bar{y} = f_1'^T(x_0) \cdots f_{p-1}'^T(x_{p-2}) \cdot f_p'^T(x_{p-1}) \cdot \bar{y}$$

- Which mode ?

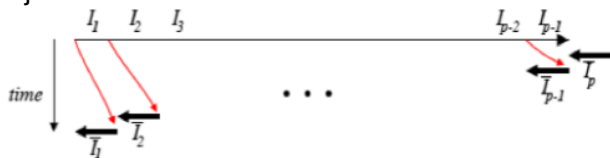
$$F : \mathbb{R}^m \longrightarrow \mathbb{R}^n$$

Tangent mode:  $m \leq n$

Reverse mode:  $m \gg n$  (e.g. data assimilation)

# Automatic Differentiation (AD) of Computer Programs : Recompute vs Restore

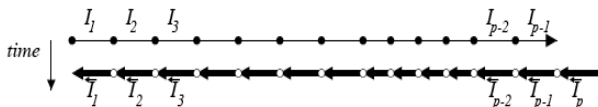
- Form of adjoint code:



- Recompute all strategy:

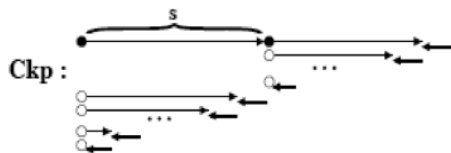


- Restore all strategy:

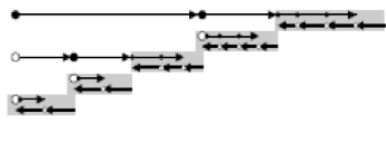
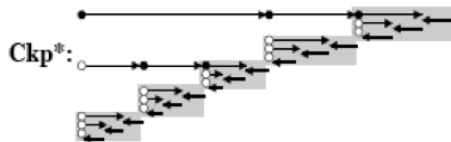
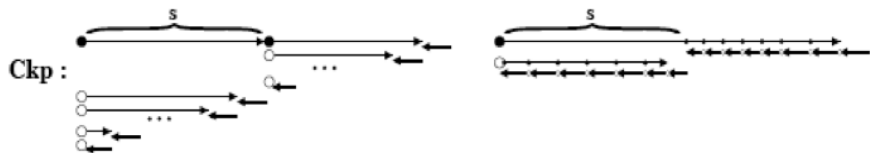


# Automatic Differentiation (AD) of Computer Programs : Checkpointing

Recompute-All :



Store-All :



# Validation and Performances: Correctness test

- Compute a directional derivative  $\nabla F(x)\dot{x}$  for a random direction  $\dot{x}$  using finite difference and the tangent linear mode AD.
- Compute via the reverse mode a single adjoint  $\bar{x} = \nabla F(x)^T \bar{y}$  for  $\bar{y}$  equals  $\dot{y}$  the output of the tangent linear mode.
- Check the following equality within the limits of the machine precision

$$\lim_{\varepsilon \rightarrow 0} \left| \frac{F(x + \varepsilon \dot{x}) - F(x - \varepsilon \dot{x})}{2\varepsilon} \right|^2 = \dot{y}\dot{y} = \bar{x}\dot{x}$$

- Dot product test for 1000 iterations

Divided differences ( $\varepsilon = 10^{-7}$ )	4.405352760987440e+08
AD (tangent linear)	4.405346876439977e+08
AD (adjoint)	4.405346876439867e+08

# Validation and Performances: Binomial Checkpointing

- Multilevel checkpointing e.g. MIT-gcm not optimal
- Optimal checkpointing 'treeverse/revolve'

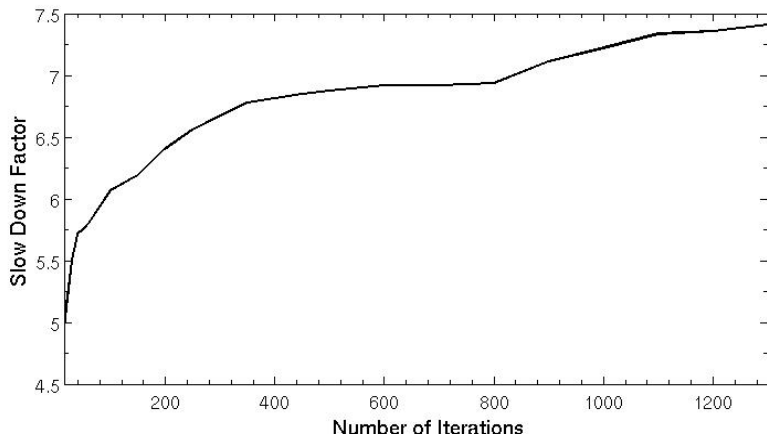
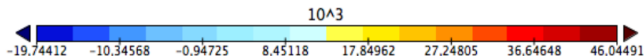
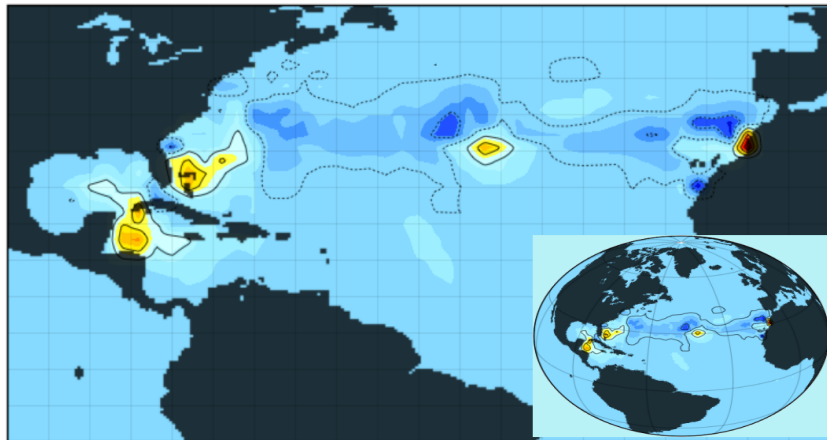


Figure: Slow dow factor vesus Max number of iterations

# Validation and Performances: Sensitivity Analysis

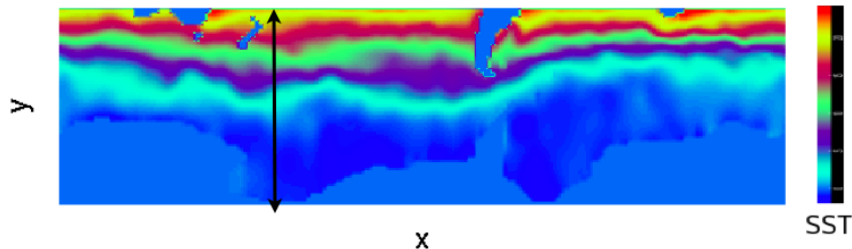
- The sensitivity of the north atlantic heat transport at  $29^{\circ}\text{N}$ , to changes in temperature at the ocean surface.



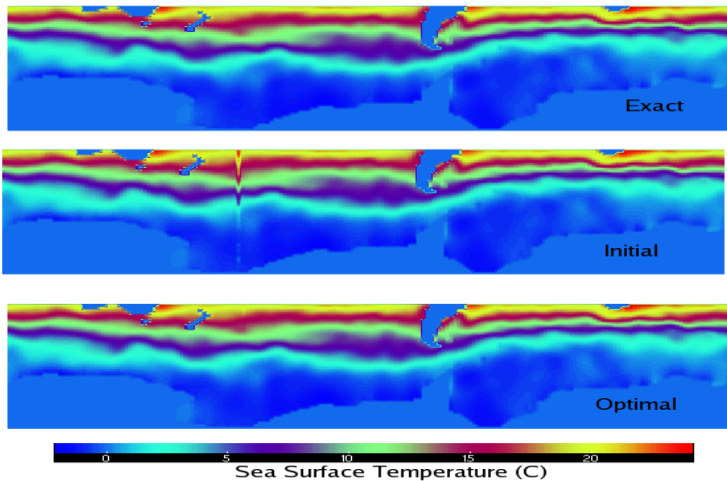


# Twin Experiment

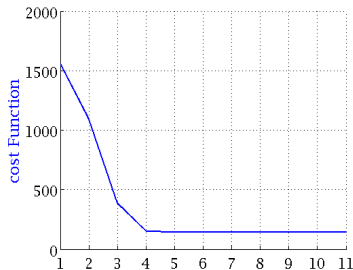
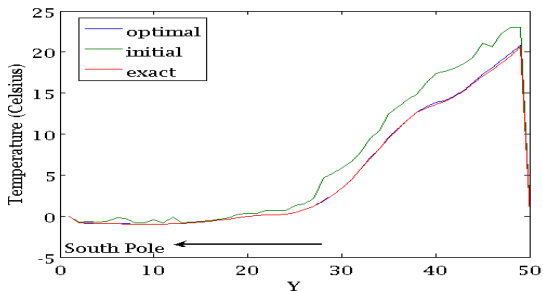
- Zoom on Antarctic
- Fully nonlinear approach
- Distributed observations
- Estimating sea surface temperature at  $x=60$  (longitude)



# Twin Experiment: Results



# Twin Experiments: Results



- PCG Solver
- Wind stress assimilation for a stationary solution