

# On the parallelisation of CO<sub>2</sub> flux inversion schemes

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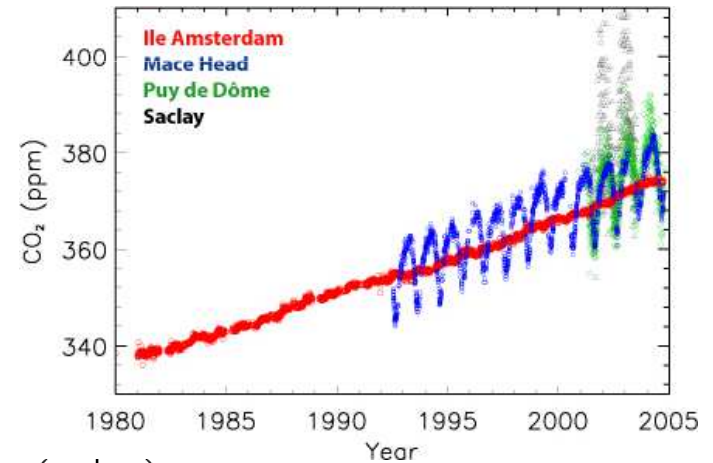
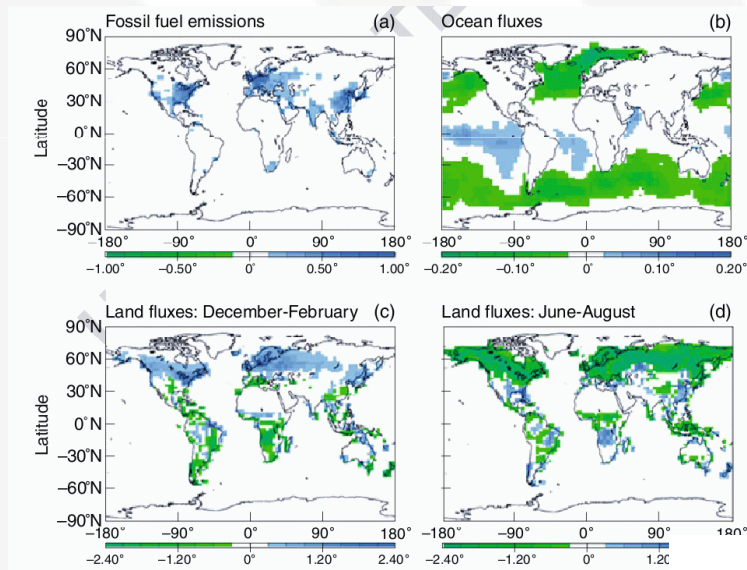


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# Global CO<sub>2</sub> flux estimation

- Non-reversible atmospheric mixing
  - Need a statistical approach to revert the sign of time

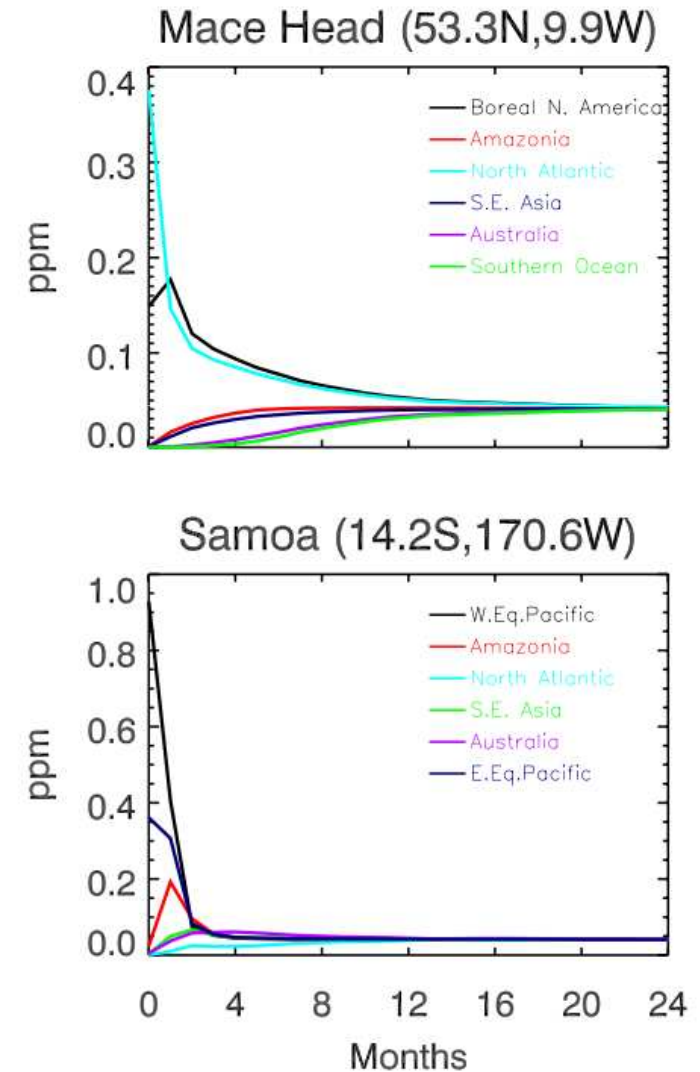


$$p(\mathbf{x}|\mathbf{y}) = \frac{p(\mathbf{x}) \cdot p(\mathbf{y}|\mathbf{x})}{p(\mathbf{y})}$$

# Time scales

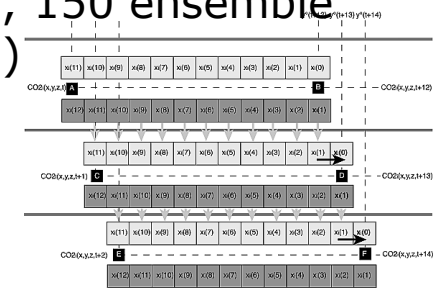
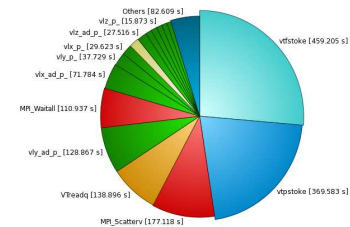
- Long inversion windows needed for long-lived gases
- Interhemispheric exchange time of  $\sim 1.4$  years (Patra et al. 2011)

Monthly average basis functions calculated as one-month flux pulses transported forward from each source region and sampled at a particular observation site. Bruhwiler et al. (2005)



# Parallelisation strategies

- Use current supercomputing infrastructures
- Solution 1: variational formulation, parallelise the atmospheric transport model with MPI
  - No simplification
  - Tedious effort
  - Technical hard point: I/O
  - Ex: 32-yr inversion in  $\sim 2$  months on 8 CPUs (Chevallier et al. 2009)
- Solution 2: ensemble formulation, split the state vector  $\mathbf{x}$  into independent pieces
  - Compromise between mixing time and length of  $\mathbf{x}$
  - Additional localisation simplifications needed
  - Physical consistency is achieved by giving up statistical consistency
  - Ex: NOAA's CarbonTracker (Ensemble Kalman smoother, 150 ensemble members; 5-week inversion window; Peters et al. 1997)

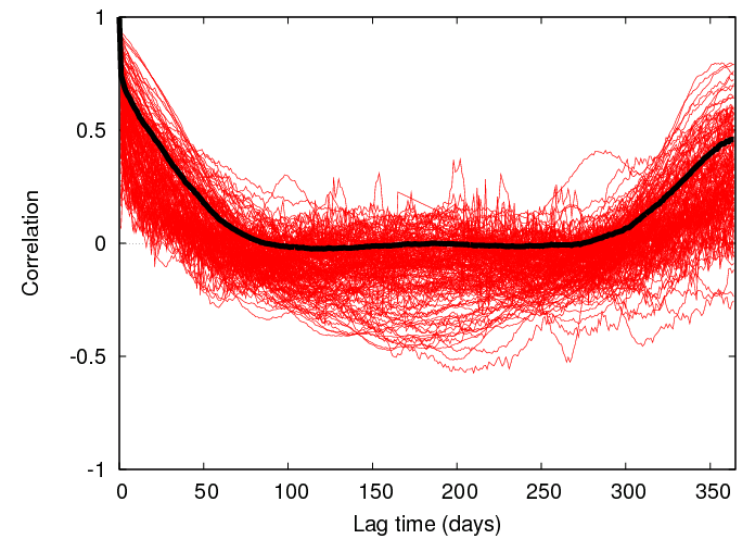
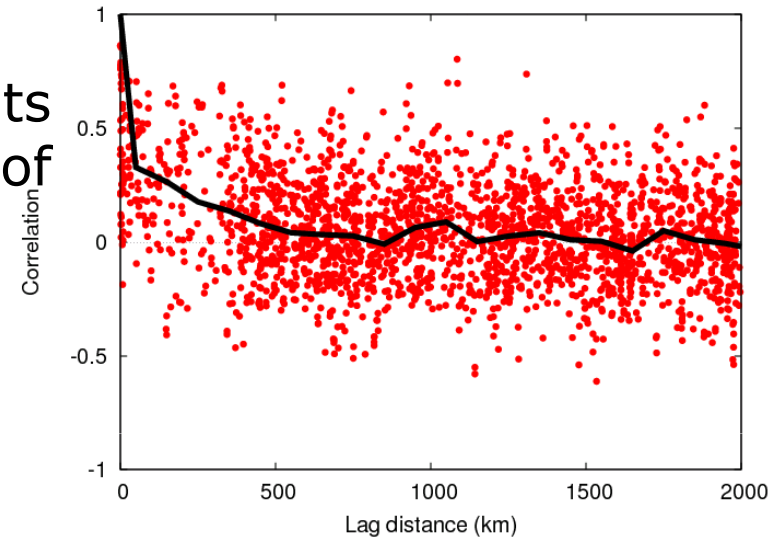
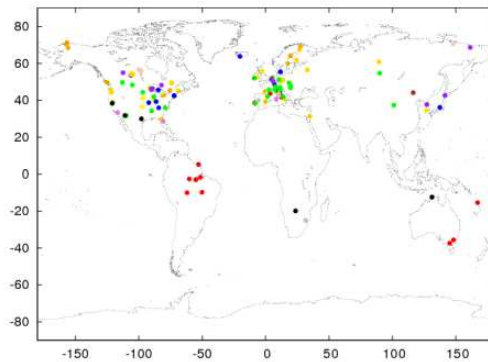


# Structure of the prior CO<sub>2</sub> flux errors

- Use daily-mean eddy-covariance flux measurements to assign the error statistics of the prior fluxes



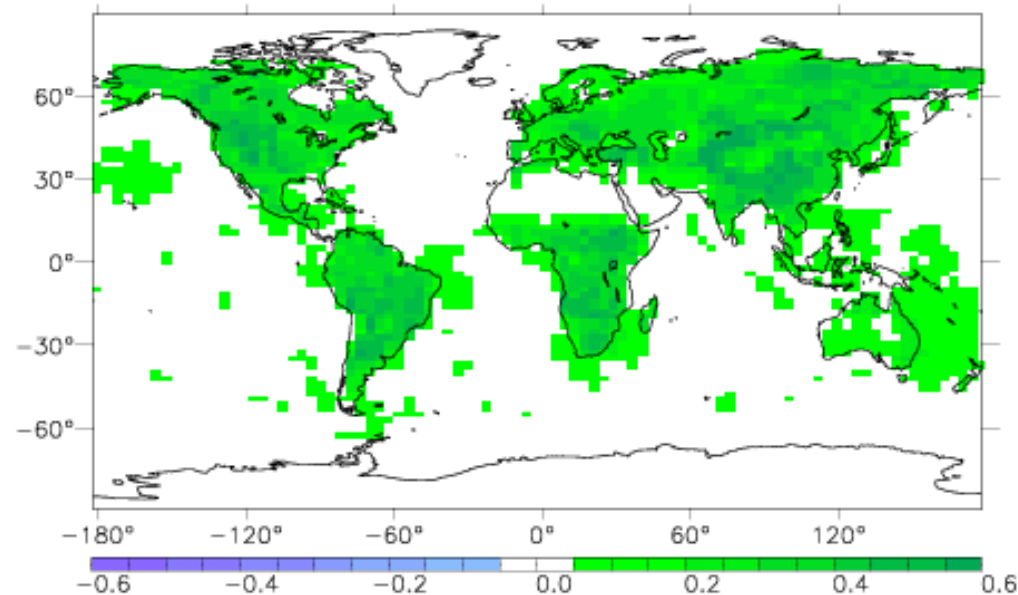
- Results:
  - Small spatial correlations
  - Large temporal correlations



# Full-rank results

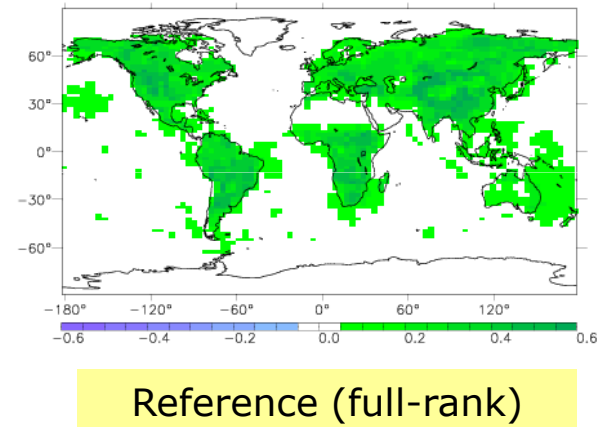
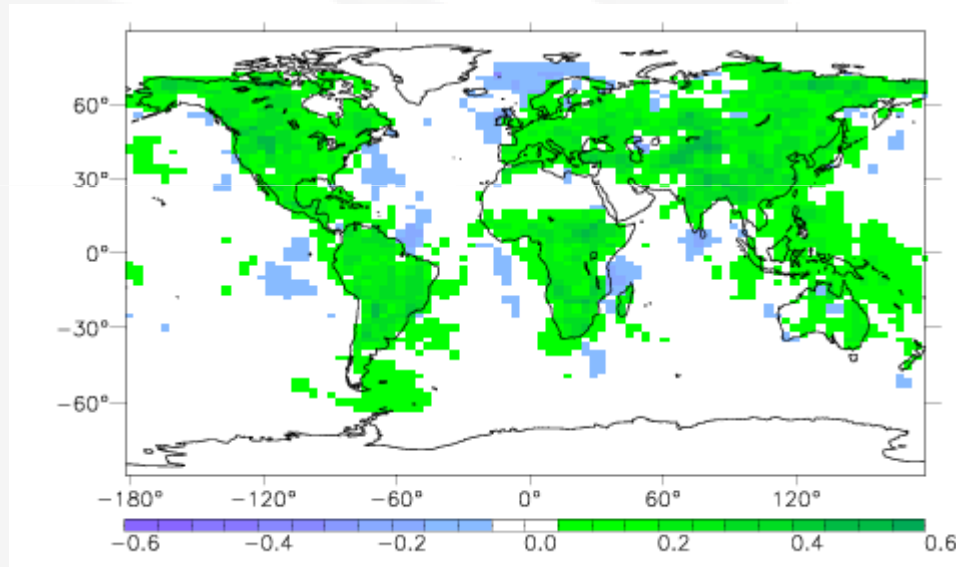
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- Theoretical uncertainty reduction ( $1 - \sigma_a / \sigma_b$ ) provided by the GOSAT satellite observations for the estimation of weekly-mean CO<sub>2</sub> surface fluxes.
  - $\sigma_a$  is the posterior error standard deviation and  $\sigma_b$  is the prior error standard deviation



# Low temporal resolution

- Same as previous, but the inversion is performed with a simplified prior error temporal error covariance matrix.

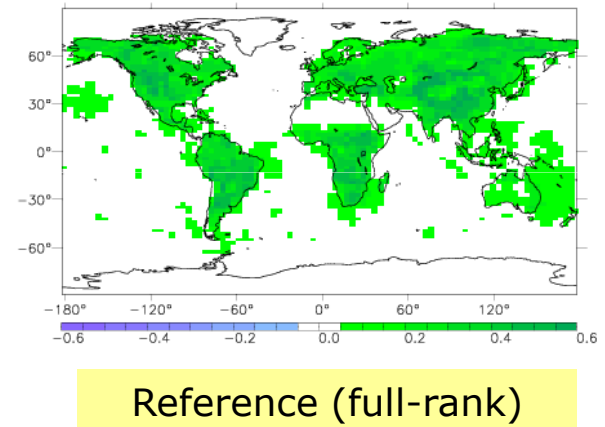
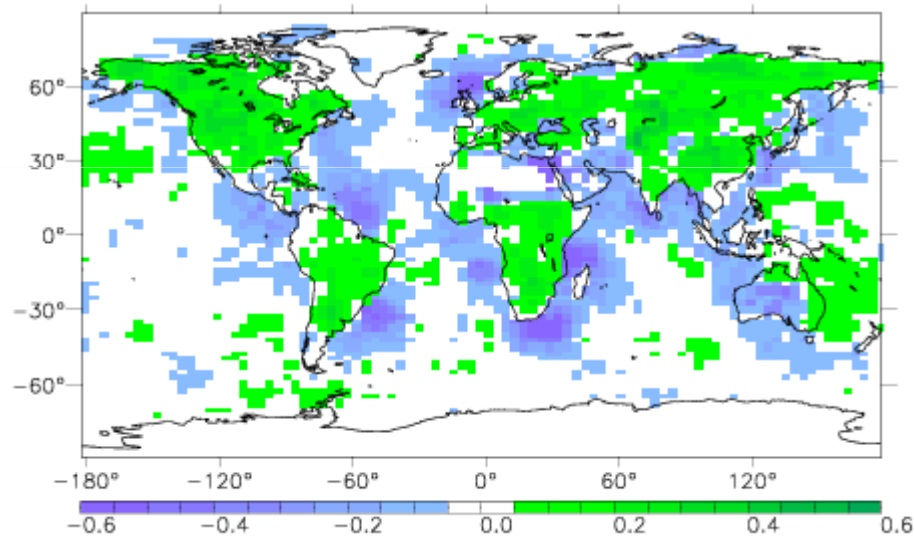


- The positive impact of the GOAT data is reduced compared to the optimal case. The maximum values become less than 50% and some negative values occur for ocean grid points.



# Low spatial resolution

- Same as reference, but the inversion is performed with a truncated prior error spatial error covariance matrix.
  - 200 leading evs (from 7000) kept for each week



- The positive impact of the GOAT data is reduced to values less than 30% over land and becomes significantly negative (i.e. the inverted fluxes are worse than the prior fluxes) over some parts of the ocean basins.



# Parallelisation strategies: $y$

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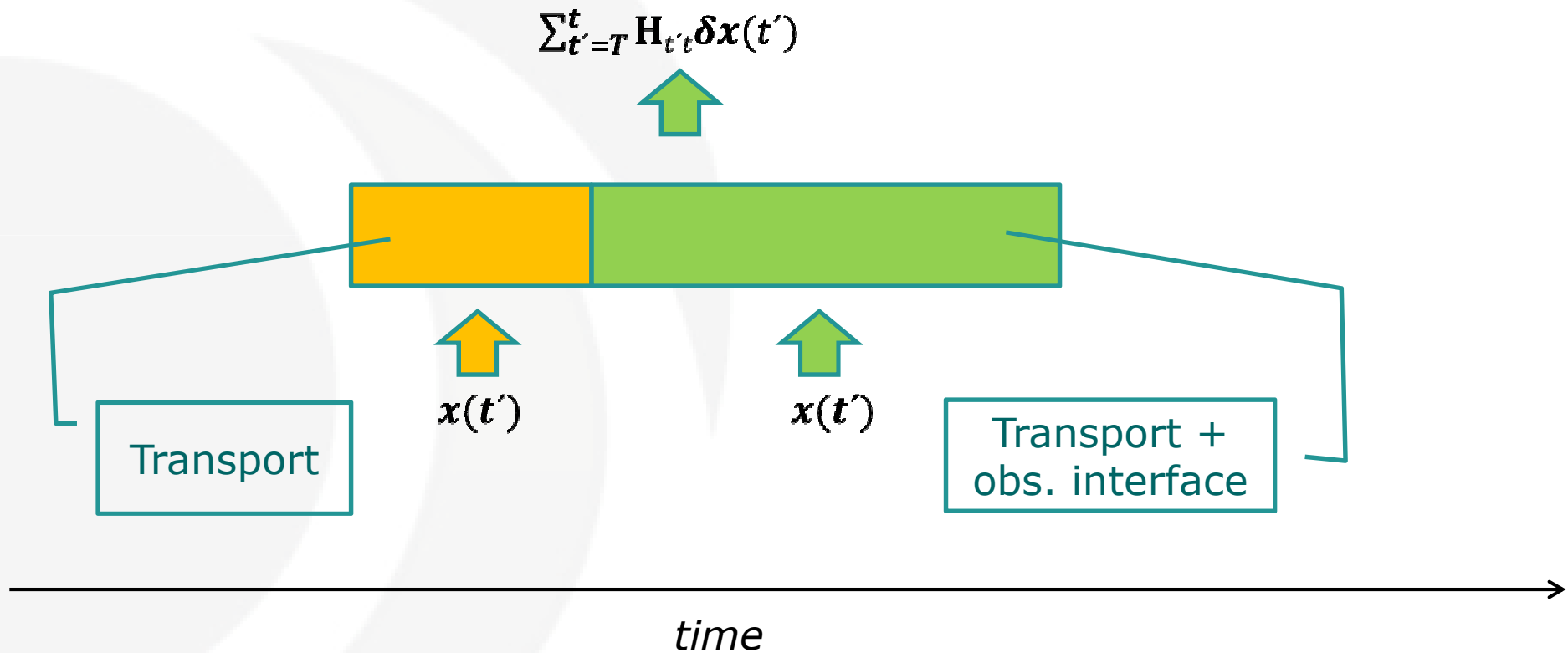
$$J(\mathbf{x}) = (\mathbf{x} - \mathbf{x}_b)^T \mathbf{B}^{-1} (\mathbf{x} - \mathbf{x}_b) + (\mathbf{H}\mathbf{x} - \mathbf{y})^T \mathbf{R}^{-1} (\mathbf{H}\mathbf{x} - \mathbf{y})$$

- Solution 3: variational formulation; process observation sets in parallel
  - All observations are linked through atmospheric transport
  - Split the inversion window into overlapping segments
  - Replace  $\delta\mathbf{y}(t) = \sum_{t'=-\infty}^t \mathbf{H}_{t't} \delta\mathbf{x}(t') + \varepsilon$   
with  $\delta\mathbf{y}(t) = \sum_{t'=T}^t \mathbf{H}_{t't} \delta\mathbf{x}(t') + \delta b(T) + \varepsilon_2$
  - $\delta b(T)$  = global scalar, mass increment of C in the atmosphere
  - The sum term can be computed in parallel
  - $\delta b(T)$  can be computed *afterwards*
  - *Ensemble method*, but without a statistical ensemble

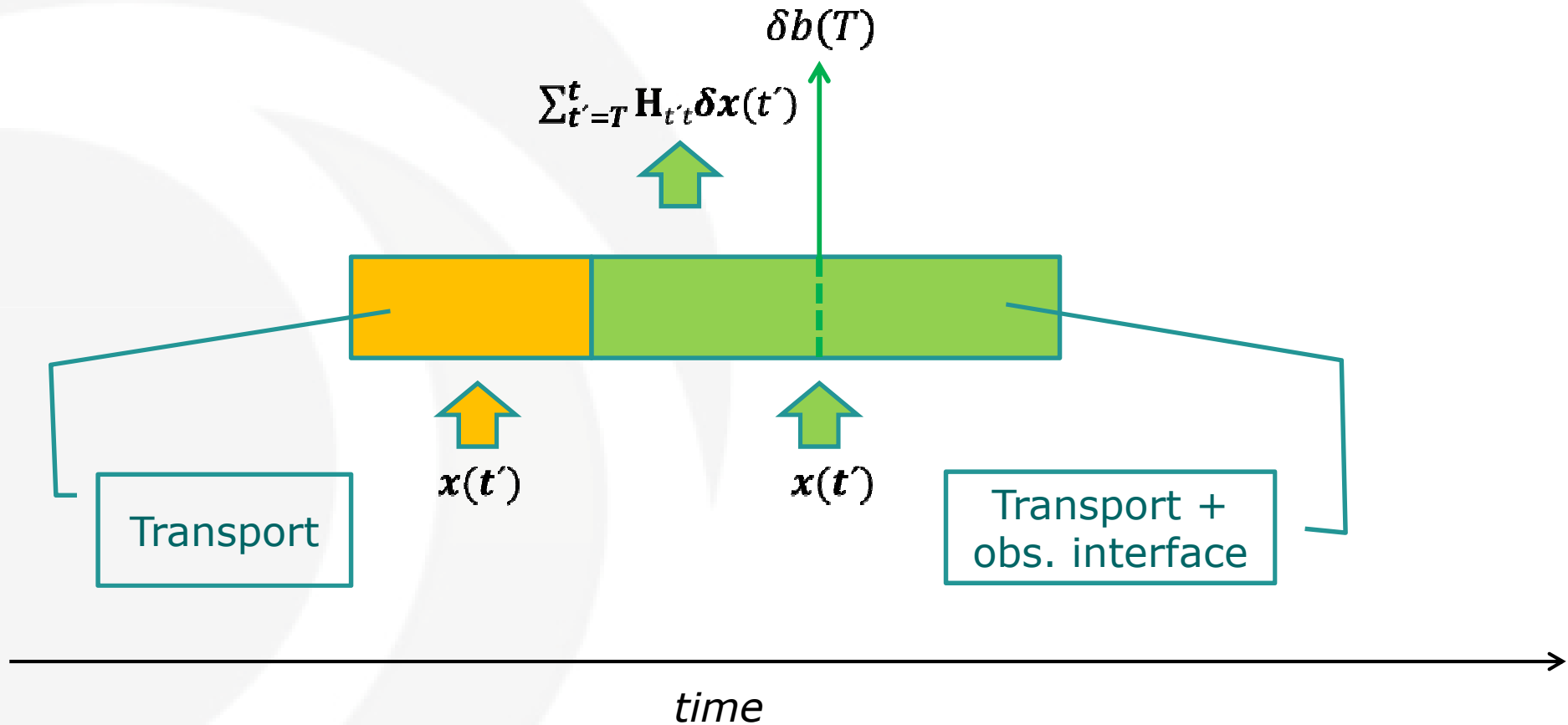
- At the highest level, the inversion requires a single cost

# Scheme for the TL computation

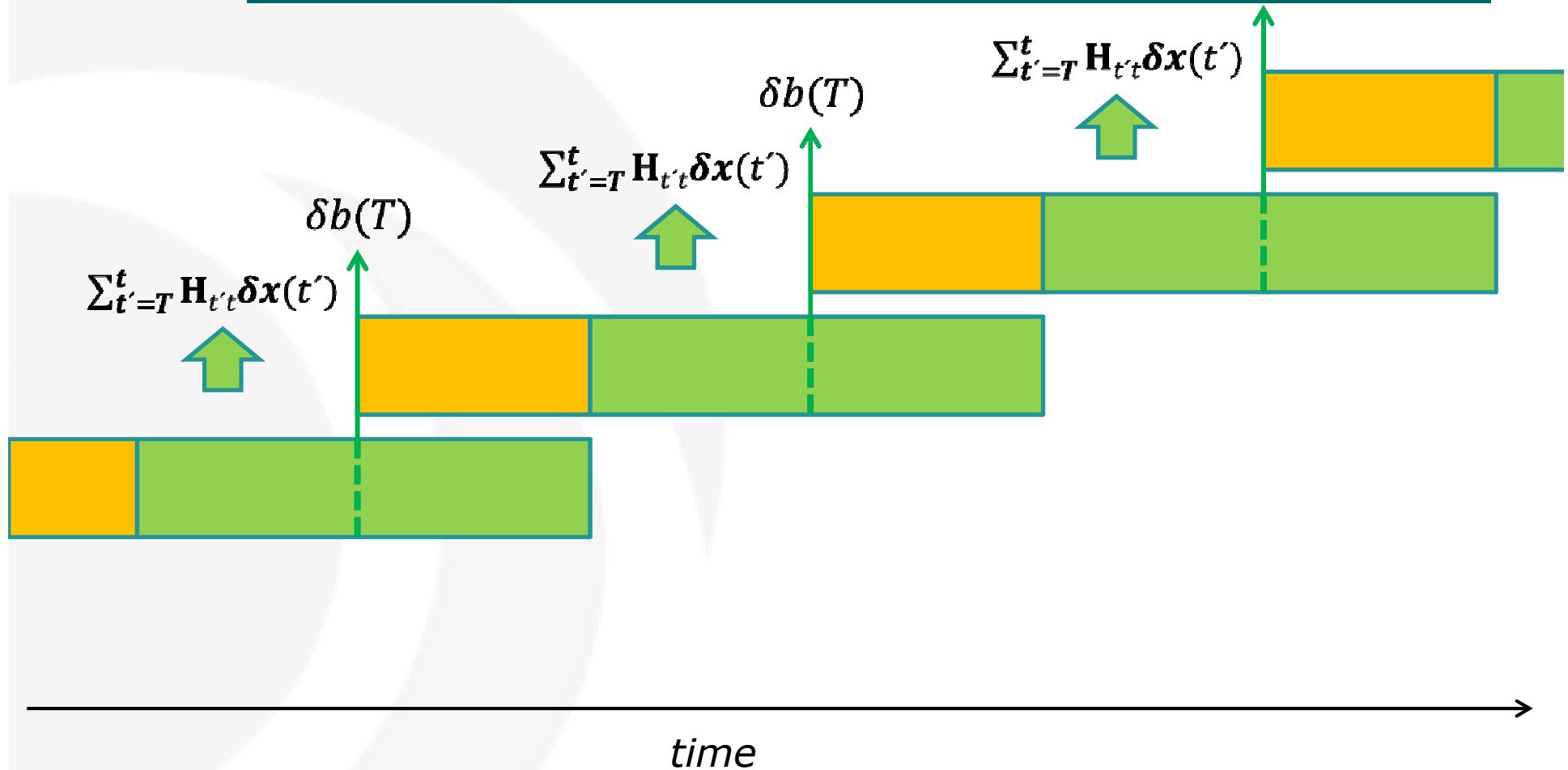
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# Scheme for the TL computation

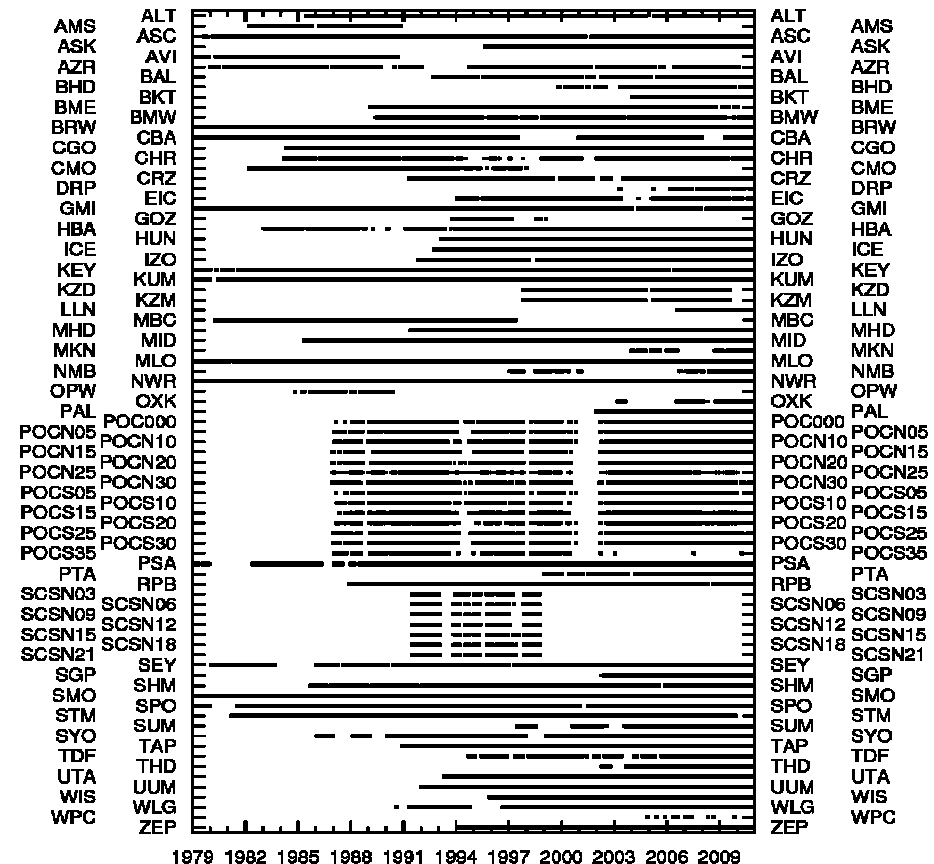
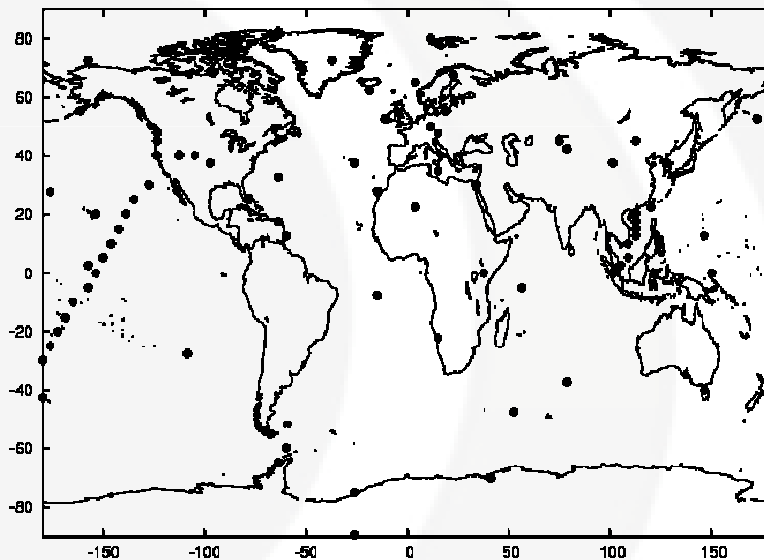


# Scheme for the TL computation



# Parallelisation strategies: y

- $\delta y(t) = \sum_{t'=T}^t \mathbf{H}_{t't} \delta x(t') + \delta b(T) + \epsilon_2$ 
  - Test on a 32-yr period, with a (3-mnth mixing) and (1-yr obs) configuration. Lanczos-based minimizer, 30 iterations

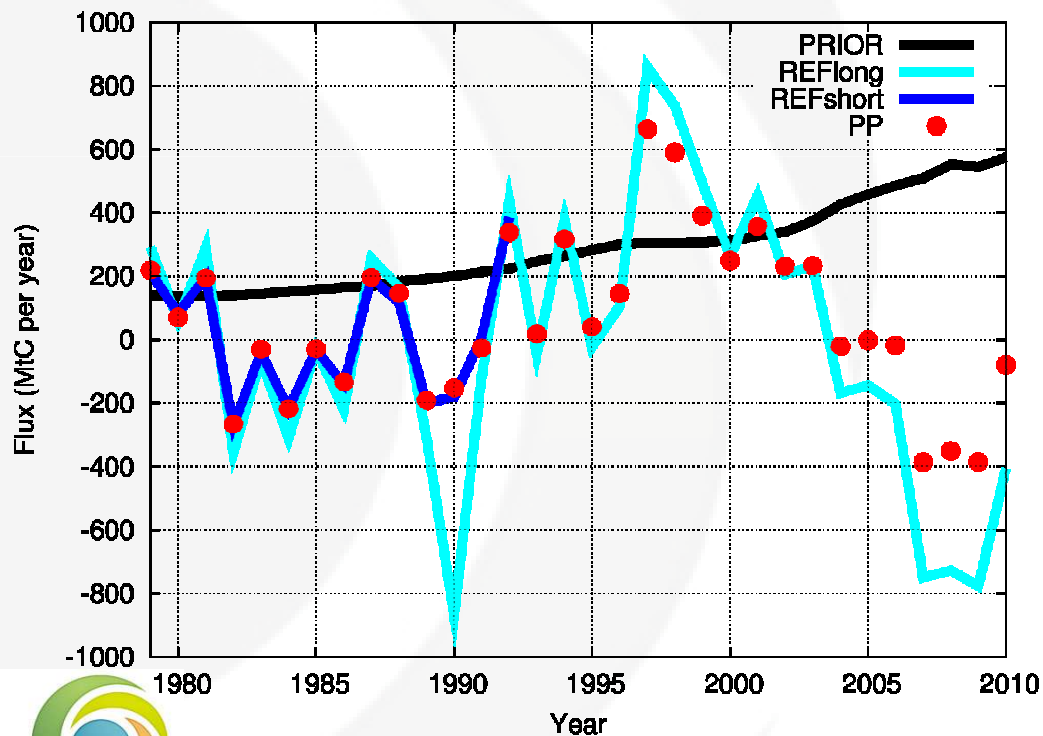


Flask data from



# Parallelisation strategies: $y$

- $\delta y(t) = \sum_{t'=T}^t \mathbf{H}_{t't} \delta x(t') + \delta b(T) + \varepsilon_2$ 
  - Test on a 32-yr period, with a (3-mnth mixing) and (1-yr obs) configuration. Lanczos-based minimizer, 30 iterations



Time series of the annual total CO<sub>2</sub> fluxes (including fossil fuel) in region Tropical Asia.

**PRIOR** = prior fluxes

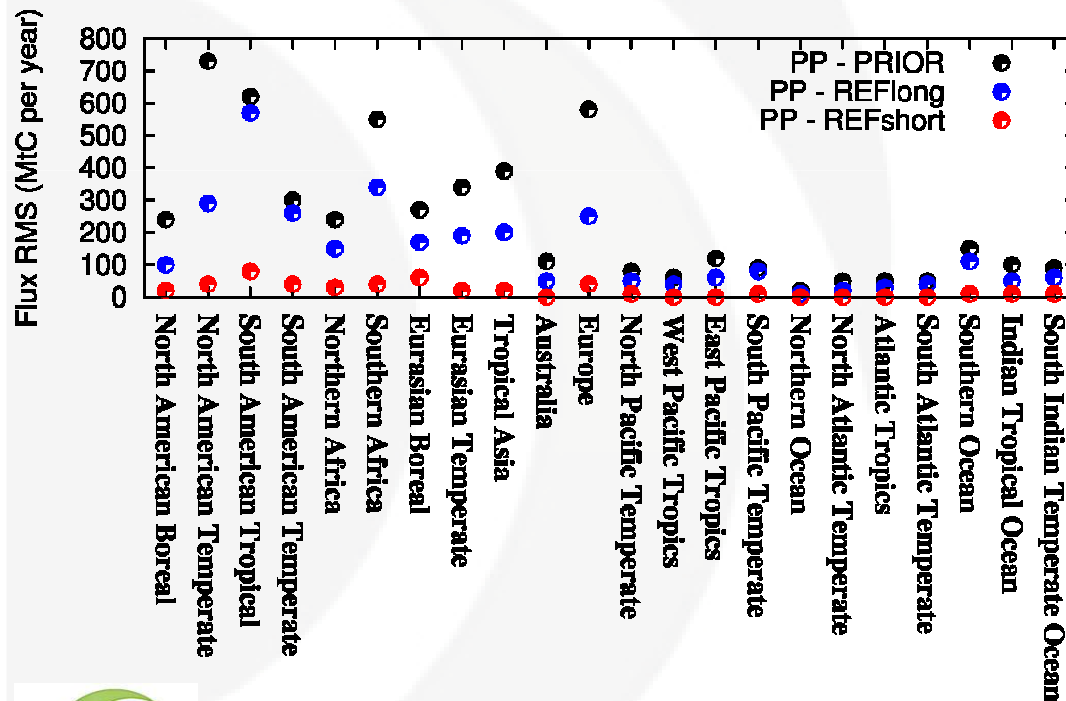
**REFlong** = inverted fluxes for the reference 1979-2010 inversion

**REFshort** = inverted fluxes for the reference 1979-1992 inversion

**PP** = 1979-2010 Parallelised inversion

# Parallelisation strategies: y

- $\delta y(t) = \sum_{t'=T}^t \mathbf{H}_{t't} \delta x(t') + \delta b(T) + \varepsilon_2$ 
  - Test on a 32-yr period, with a (3-mnth mixing) and (1-yr obs) configuration. Lanczos-based minimizer, 30 iterations



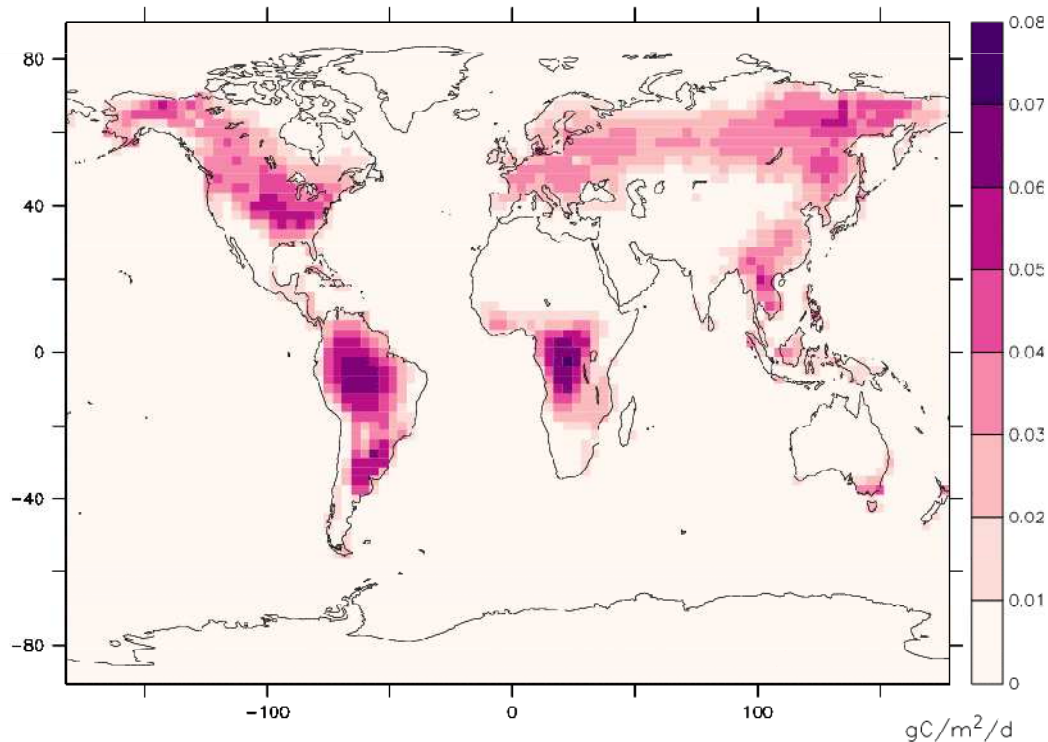
RMS differences for annual regional CO<sub>2</sub> fluxes between

- the PP inversion and its prior for the period 1979-2010 (**PP-PRIOR**),
- the 1979-2010 PP inversion and the 1979-2010 reference inversion for the same period (**PP-REFlong**)
- the 1979-2010 PP inversion and the 1979-1992 reference inversion for 1979-1991 (**PP-REFshort**)  
 ~ 10% of increments



# Parallelisation strategies: $y$

- $\delta y(t) = \sum_{t'=T}^t \mathbf{H}_{t't} \delta x(t') + \delta b(T) + \epsilon_2$ 
  - Test on a 32-yr period, with a (3-mnth mixing) and (1-yr obs) configuration. Lanczos-based minimizer, 30 iterations



RMS differences between the 1979-2010 PP inversion and the reference 1979-1992 inversion when computing the inverted weekly fluxes.

The statistics are computed for the 1979-1991 period.

< 11% of inversion increments

# Parallelisation strategies: $y$

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- $\delta y(t) = \sum_{t'=T}^t \mathbf{H}_{t't} \delta x(t') + \delta b(T) + \epsilon_2$ 
  - Test on a 32-yr period, with a (3-mnth mixing) and (1-yr obs) configuration. Lanczos-based minimizer, 30 iterations
  - Cost for 1 iteration (1 TL + 1 AD):

	Reference	// version
Nb of CPUs	8	32
CPU time	352h	192h
Wall clock time	44h	6h
Disk space	155GB	180GB
Swap space per CPU	2.5GB	20.5GB

# Conclusion

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- Parallelisation is less natural for variational DA systems than for ensemble ones, but smart strategies are possible
- For flux inversion, our strategy allows
  - Benefitting from the high-resolution of the variational formulation
  - Keeping the physical and statistical consistency of the inversion over long periods (decades)
  - Improved numerical stability
  - 7x faster execution speed