# Advances in Time-Parallel Four Dimensional Data Assimilation in a Modular Software Framework

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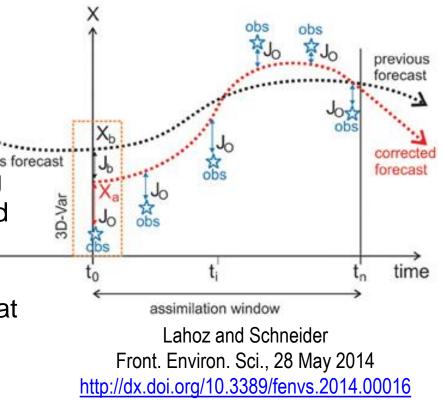


# The HPC group at NOAA/ESRL/GSD

- Strong track record in high performance computing
  - Massively Parallel Fine Grain (MPFG) Computing
  - Graphics Processing Units (GPUs)
  - Many Integrated Core (MIC)
- Wishes to advance the state of the art in data assimilation, in particular, via improved performance and design
  - NOAA/NCEP GSI has a core limit in the hundreds
  - 4D-Var approaches are time consuming
  - 4D-Ensemble approaches are memory & I/O intensive
  - Wish to use a 'great' solver with any model (atmos, ocean...)
- First steps into data assimilation (started this year)

 Four Dimensional Data Assimilation (4D DA) requires having information on the state of the system at many different times

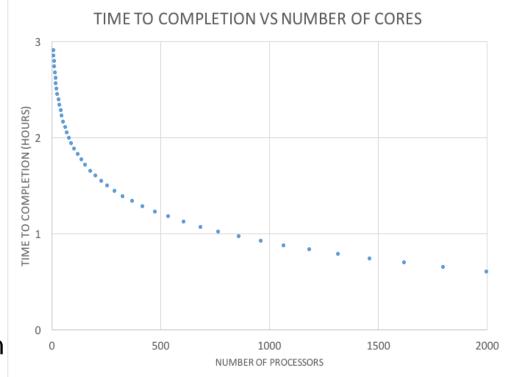
- In some approaches, information at different times is achieved by running a model forward (Tangent-Linear) and backward (Adjoint) in time
- Optimal results with a TL and AD that mimic the true model



• The time spent running the TL and AD is, roughly:

LENGTH OF ASSIM WINDOW

- \* 2 (TL & AD)
- \* 1.5 (TL TAKES LONGER)
- \* 1.5 (AD TAKES LONGER)
- \* NUMBER OF ITERATIONS
- For a 12 hour window, 40 iterations, this value is 54\*40=2160 hours, or 90 days
- This is, perhaps, 6x longer than the forecast itself - this must be improved

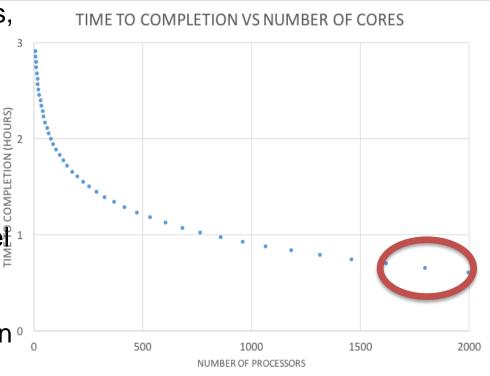


 If the assimilation window could be broken into 48 ¼-hour windows, then run time *could* be closer to 2 model days rather than 90

• Would take ~27-minutes to compute for 1% real-time model

• Achieve scaling when your mode

 If scaling achieved, is the solution of from this time-parallel version just as good?



### **4DVAR Data Assimilation**

4DVAR traditionally involves taking one state (bucket), moving it all the way from the start to finish to start

Time parallel 4DVAR sends a number of states (buckets) from one time to the adjacent time





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WELCOME TIME-PAR OBJ-ORIENT RESULTS SUMMARY

**4DVAR Data Assimilation** 

We did not invent time-parallel 4DVAR – the ECMWF has done this sort of work (Saddle Point), as have others (e.g. Virginia Tech)

Our goal is not to develop a brand new DA system, but to explore promising existing approaches





WELCOME TIME-PAR OBJ-ORIENT RESULTS SUMMARY

# Software Design - The Current State

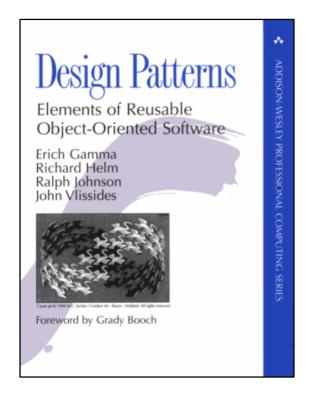
- Legacy Fortran 77/90 codes 100K+ lines
- Developers lack training & expertise in modern software engineering practices
- Poor software designs
  - Difficult to extend with new capabilities
  - Difficult to debug
  - Lack interoperability
  - Hinder R2O
  - Slow scientific progress

# Addressing Design Deficiencies

- Maximize cohesion of software components
  - Components cannot be subdivided without exposing internal state.
- Minimize coupling of software components
  - Changes to A should not require changes to B
- Object-Oriented Design
  - Facilitates high cohesion and low coupling
  - Adopted by Software Industry 20+ years ago!

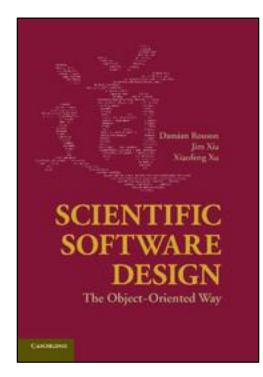
# Addressing Design Deficiencies

- Design Patterns (1994)
- Arsenal of well-tested recipes for solving recurring design problems
- Arguably revolutionized objectoriented software development



Addressing Design Deficiencies

- Scientific Software Design (2011)
- Patterns for scientific applications (Fortran/C++)
- Design is as important as performance



# Is Object Oriented Design The Solution?

- Why aren't we using modern software design principles?
  - Historical lack of language support (pre-Fortran 2003)
  - "Make it work, then make it fast, then forget it" culture
  - Design to accommodate future requirements not a priority?
  - Bad Performance?
- Will object-oriented design make it too slow?
- Performance is key to meet Operations deliver deadlines but...
  - Inflexible designs hurt R2O A less tangible cost
  - Total cost of ownership includes cost of code maintenance

We Are Investigating

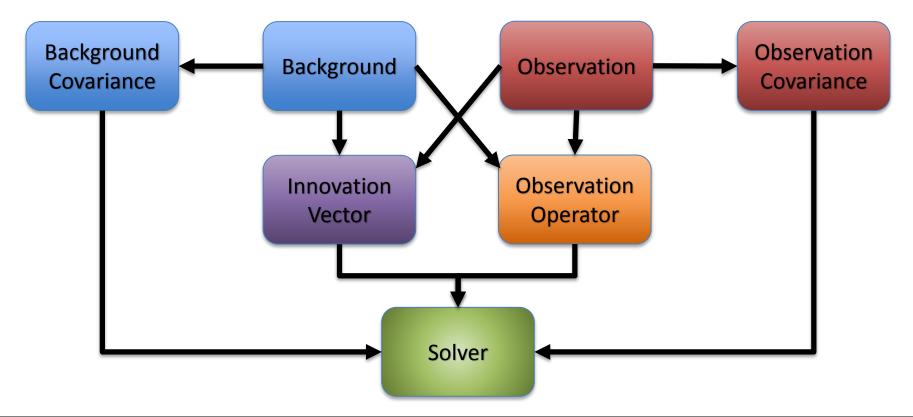
- What is the performance impact of object oriented design?
  - Let's measure it!
- Construct a data assimilation prototype
  - Design from scratch using "old school" methods
  - Provides baseline verification of correctness
  - Provides baseline performance measurements
- Construct a copy of the prototype
  - Progressively refactor and apply object oriented techniques
  - Measure correctness and performance against baseline

# We Are Investigating

- Incremental approach
  - Facilitates identification of designs that cause performance penalties
  - Facilitates accumulation of patterns for techniques to avoid performance penalties
- Explore Fortran 2003/2008 features
  - Limitations of Fortran 2003/2008 object oriented features
  - Limitations of compiler support for 2003/2008 features
  - Interactions with OpenMP and OpenACC
- May consider other languages in the future

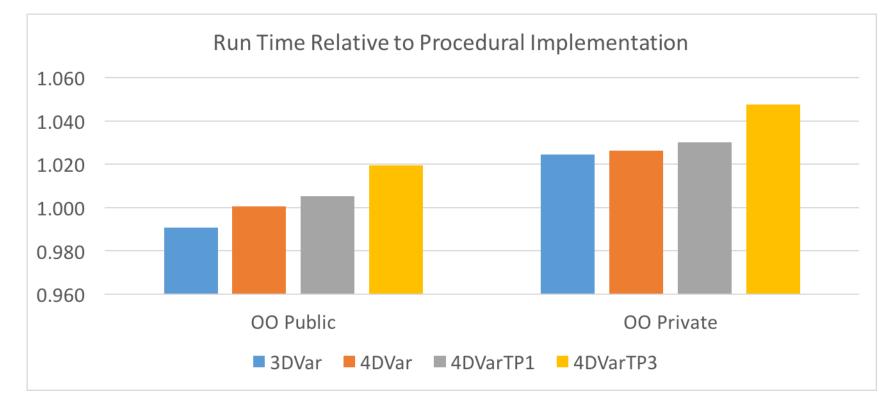
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### What We Have So Far



### Results - Object Oriented Design

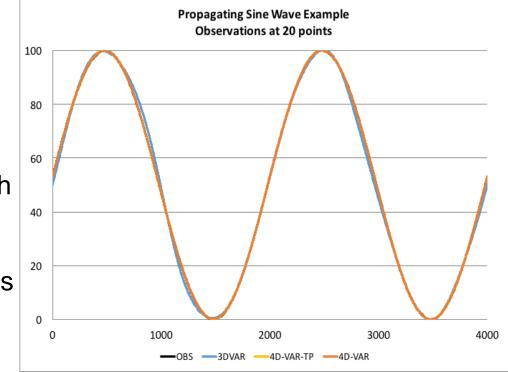
### Performance of Object Oriented Implementation



#### WELCOME TIME-PAR OBJ-ORIENT RESULTS SUMMARY

# Test 1: The eastward propagation of a 1D sine wave

- Tested on a single case
- Assimilation over a 3 "hour" assimilation window
- •Time parallel 3 'windows', each its own OMP thread
- 4000 gridpoints, 20 observations
- TL and AD simple: derivative of sine is cosine

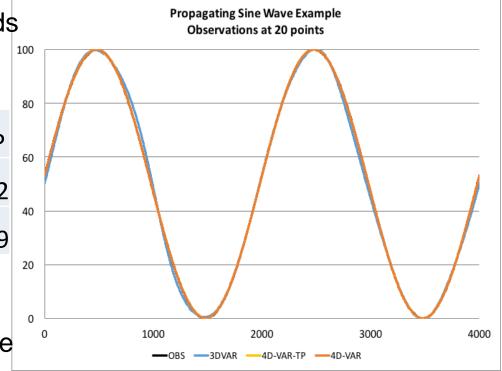


# Test 1: The eastward propagation of a 1D sine wave

• Results show that all the methods produce an improved analysis of the sine wave

|      | 3D-VAR | 4D-VAR  | 4D-VAR-TP |
|------|--------|---------|-----------|
| BIAS | 0.2805 | -0.0326 | -0.0302   |
| RMSE | 2.4913 | 0.0073  | 0.0049    |

 The question then is: does the time-parallel 4DVAR take less time to complete?

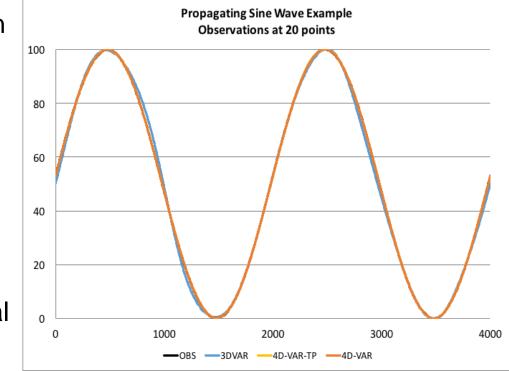


# Test 1: The eastward propagation of a 1D sine wave

• Timing results from this case (in seconds):

| 3DVAR      | 35.0  |
|------------|-------|
| 4DVAR      | 108.8 |
| 4DVAR-TP-1 | 108.6 |
| 4DVAR-TP-3 | 44.2  |

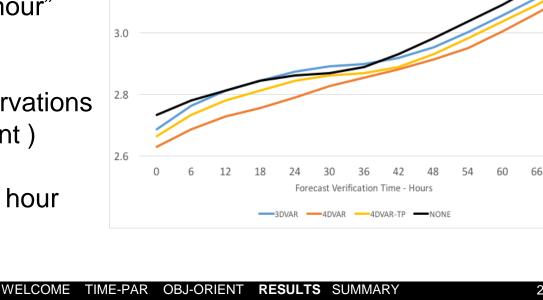
 Results show that using the 3 OMP-THREAD Time Parallel 4DVAR results in a substantial reduction in run-time length



3.4

3.2

- Test 2: The Lorenz96 Model
- Tested over a 100 "day" simulation
- Data assimilation every 6 "hours"
- Assimilation over a 3 "hour" assimilation window
- 40 gridpoints / 20 observations (every other gridpoint)
- F=8.02, Initial Offset=1 hour



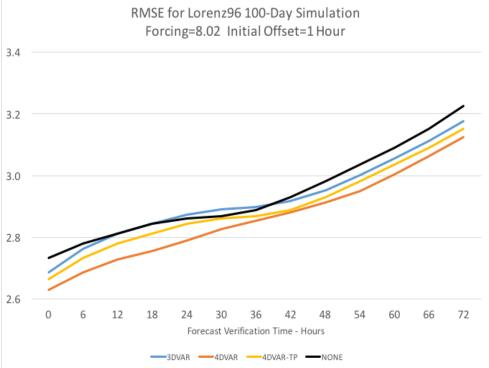
RMSE for Lorenz96 100-Day Simulation

Forcing=8.02 Initial Offset=1 Hour

72

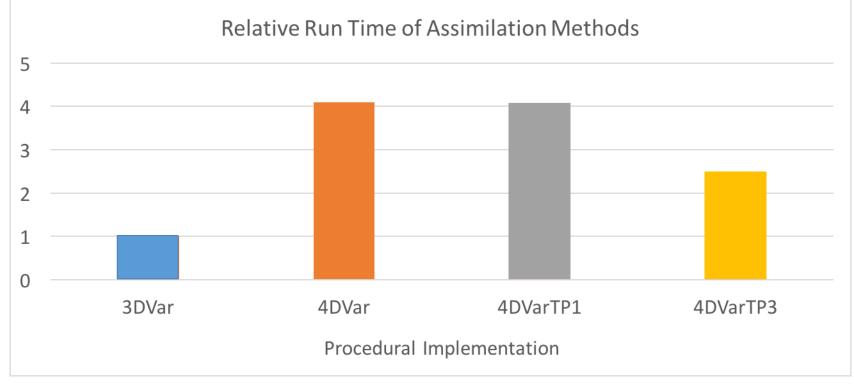
### Test 2: The Lorenz96 Model

- The time parallel 4DVAR (yellow line) performed better than 3DVAR, but not quite as well as 4DVAR
- No great performance statistics here – the 40-point problem was not taxing
- Nonetheless the Time Parallel 4DVAR results encourage us to continue on



# Results – Time to Completion

# Performance of Procedural Implementation Lorenz Model with 4000 points



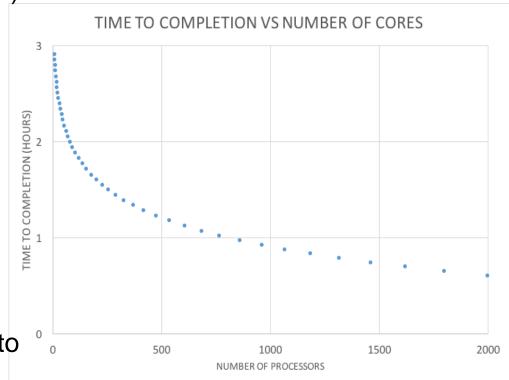
### Conclusions

- Our simple experiments have shown
  - Time-Parallel 4D DA can result in reduced time-to-completion
  - Object oriented design resulted in minimal increases in time-tocompletion
  - Accuracy of analyses and forecasts, with this method, are not unreasonable

• We expect that the benefits of Time-Parallel 4D DA will be greatest when the forecast model (and the Tangent-Linear and Adjoint thereof) have past the point of scaling well

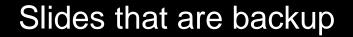
• Time-Parallel provides a new avenue to achieve scaling

- From Govett et al (BAMS paper) G11 NIM (3.75KM resolution) using 64\*20=1280 K80 GPUs runs in 1.6% of forecast time
- 12-hour forecast takes 12minutes (could do only ONE iteration of 4DVAR)
- •Time-parallel could do 40 iterations in ~30 minutes if the iterations could be subdivided into ~48 sections (60,000 K80s, 30,000 Pascals)



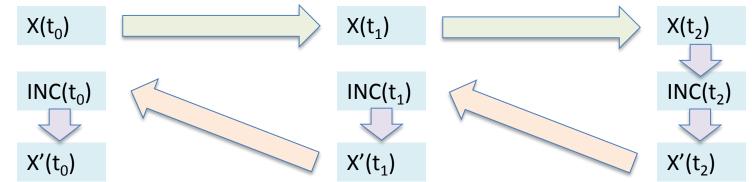
### Thoughts for the future

- In the near term, we shall:
  - Make use of symmetry of matrices
  - Work to reduce amount of memory use by each thread
  - Test on an atmospheric model, likely QG T103L3
- In the longer term, we must:
  - Think of how this would work for a global model running at nonhydrostatic scales
  - Would like to keep all the states in memory, but that might be too much memory use. But if we write to disk, will that really slow things down?
  - Work to incorporate ensembles, making for a hybrid framework (48 windows, 51 ensemble members, shared approaches?) (BLAS/LAPACK – opportunities for better performance?)
  - Apply object oriented design patterns and measure impact

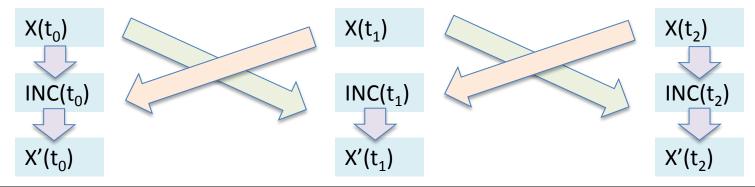


4D Data Assimilation – Time Parallel

### 4DVAR traditionally involves:



### Time-Parallel Approach does:



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WELCOME TIME-PAR OBJ-ORIENT RESULTS SUMMARY

4DVAR Data Assimilation - Traditional

4DVAR traditionally involves:

- Calculating a cost function  $2J(\mathbf{x}) = (\mathbf{x} - \mathbf{x}_b)^T \mathbf{B}^{-1} (\mathbf{x} - \mathbf{x}_b) + (\mathbf{x} - \mathbf{x}_b)^T \mathbf{H}^T \mathbf{R}^{-1} \mathbf{H} (\mathbf{x} - \mathbf{x}_b)$   $- [\mathbf{y}_0 - H(\mathbf{x}_b)]^T \mathbf{R}^{-1} \mathbf{H} (\mathbf{x} - \mathbf{x}_b)$   $- (\mathbf{x} - \mathbf{x}_b)^T \mathbf{H}^T \mathbf{R}^{-1} [\mathbf{y}_0 - H(\mathbf{x}_b)]$   $+ [\mathbf{y}_0 - H(\mathbf{x}_b)]^T \mathbf{R}^{-1} [\mathbf{y}_0 - H(\mathbf{x}_b)]$
- Calculating an increment to the present guess, 'x  $\nabla J(\mathbf{x}) = \mathbf{B}^{-1}(\mathbf{x} - \mathbf{x}_b) + \mathbf{H}^T \mathbf{R}^{-1} \mathbf{H}(\mathbf{x} - \mathbf{x}_b) - \mathbf{H}^T \mathbf{R}^{-1} \mathbf{H}[\mathbf{y}_0 - H(\mathbf{x}_b)]$
- The question then is: what is "x"?  $\mathbf{x} = \frac{1}{2} \Big[ \operatorname{TL}(\mathbf{x}_{prior}) + \operatorname{AD}(\mathbf{x}_{prior}) \Big] \quad t = 0, \quad \mathbf{x} = \frac{1}{2} \Big[ \mathbf{x} + \operatorname{AD}(\mathbf{x}) \Big] \quad t = T, \quad \mathbf{x} = \frac{1}{2} \Big[ \mathbf{x} + \operatorname{TL}(\mathbf{x}) \Big]$