ARCHER webinar

In-situ data analytics for atmospheric modelling

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Met Office NERC Cloud model (MONC)

- Uses Large Eddy Simulation for modelling clouds & atmospheric flows
 - Written in Fortran 2003 due to scientist familiarity, uses MPI for parallelisation
 - Designed to be a community model which will be accessible to be changed by non expert HPC programmers and scale/perform well.
 - For use not just by Met Office scientists, but also those in the wider weather/climate community

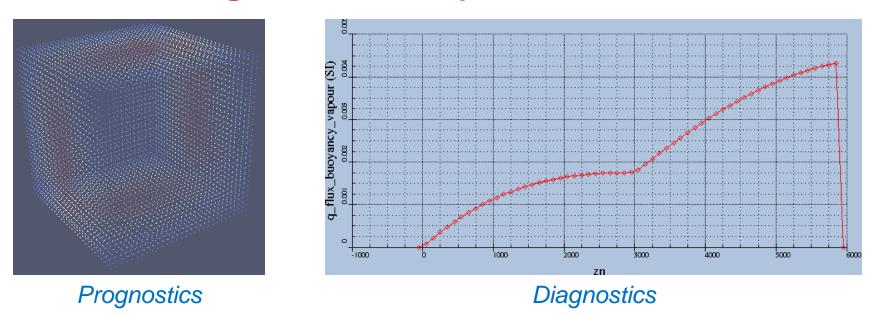


- Replaces an older model, the LEM from the 1980s
 - From 22 million to billions of grid points
 - From 256 cores to many thousands





A challenge for analysis

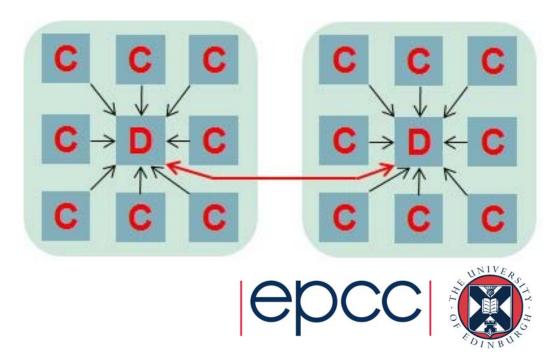


- With much larger domains (billions of grid points) how can we best analyse the data in a scalable fashion?
 - Previous LEM model did this in line with computation, where the model would stop and calculate diagnostics before continuing with computation
 - Could write to disk and analyse offline



In-situ approach

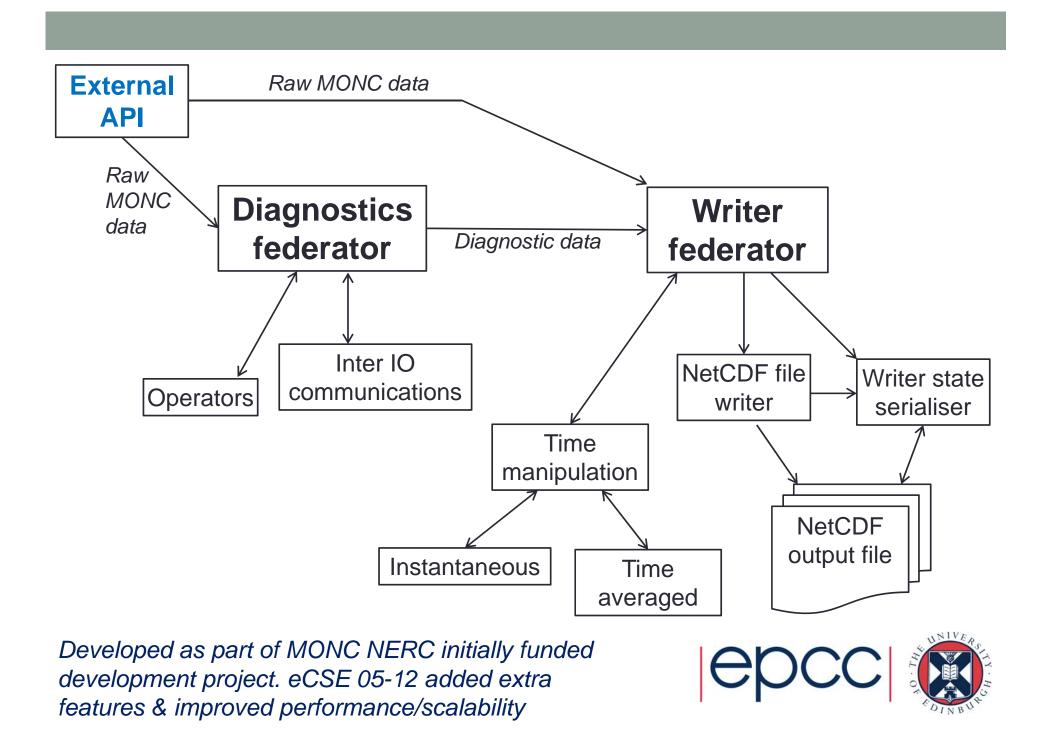
- Have many computational processes and a number of data analytics cores
 - Typically one core per processor is dedicated to IO, serving the other cores running the computational model
 - Computational cores "fire and forget" their data
- In-situ as raw data is never written out
 - Would be too time consuming
- Avoids blocking the computational cores for analytics and IO



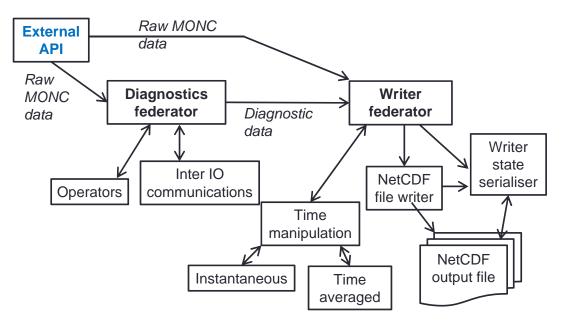
Existing approaches.....

- Some existing approaches:
 - XIOS
 - Damaris
 - ADIOS
 - Unified Model IO server
- We need:
 - To support dynamic time stepping
 - Checkpoint-restart of the IO server itself
 - Bit reproducibility
 - Scalability and performance
 - Easy configuration & extendibility

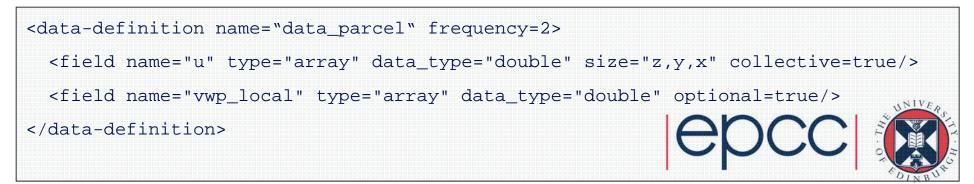




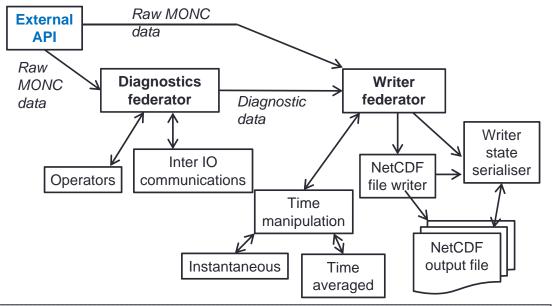
- Define the data from MONC
 - Arrays, scalars or maps
 - Mandatory (default) or optional
 - A unique subset of a field (collective) or not. If collective need to provide sizes per dimension; z, zn, y ,x and qfields
 - Integer, double, float, boolean, string



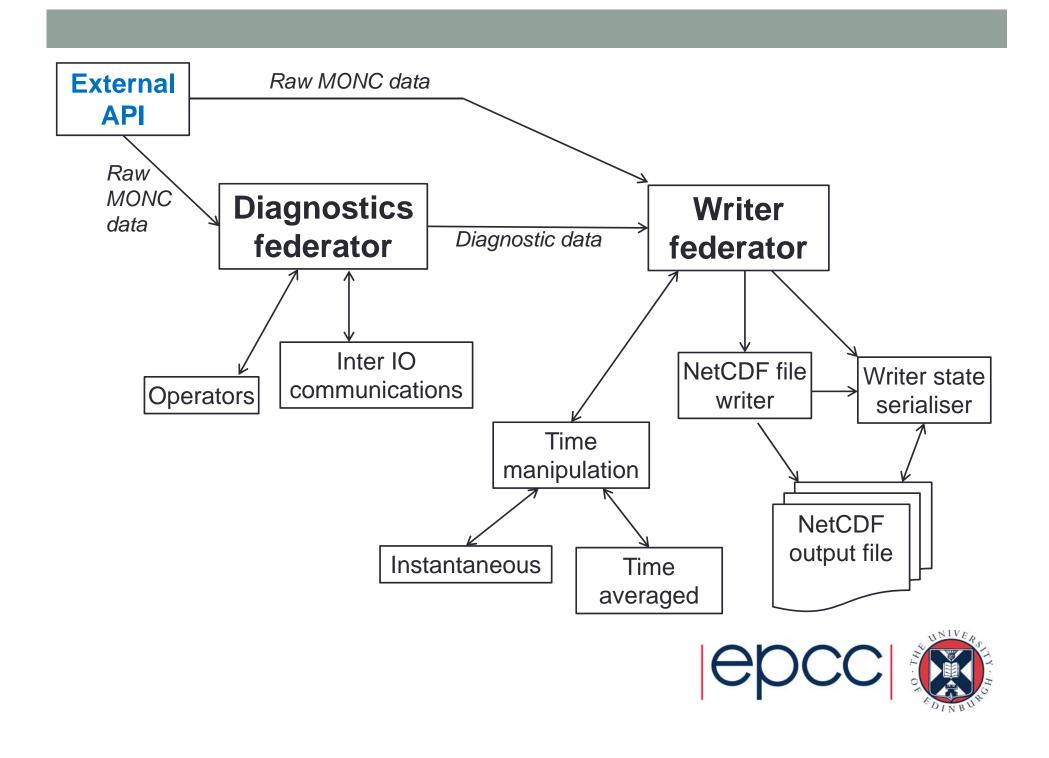
- The IO server expects this data every "frequency" timesteps
- On registration the MONC process is told what data it should send & when. MONC process tells the IO server the sizes.



- Define the diagnostics & its attributes
 - Along with how to generate this diagnostic
- Organised as communication and operators

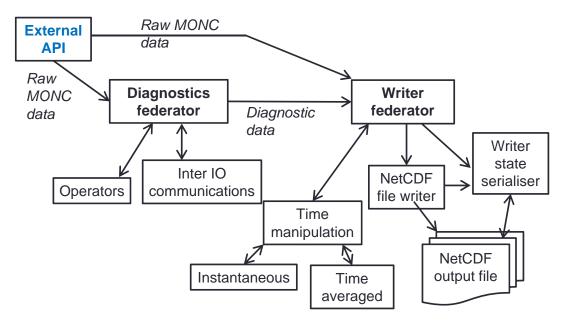


```
<data-handling>
<data-handling>
<data-handling>
<diagnostic field="VWP_mean" type="scalar" data_type="double" units="kg/m^2">
<operator name="localreduce" operator="sum" result="VWP_mean_loc_reduced"
field="vwp_local"/>
<communication name="reduction" operator="sum" result="VWP_mean_g"
field="VWP_mean_loc_reduced" root="auto"/>
<operator name="arithmetic" result="VWP_mean"
equation="VWP_mean_g/({x_size}*{y_size})"/>
</diagnostic>
<data-handling>
```



User configuration - writing

<proup name="3d_fields"></proup>	
<member name="w"></member>	
<member name="u"></member>	



Reuse of configuration

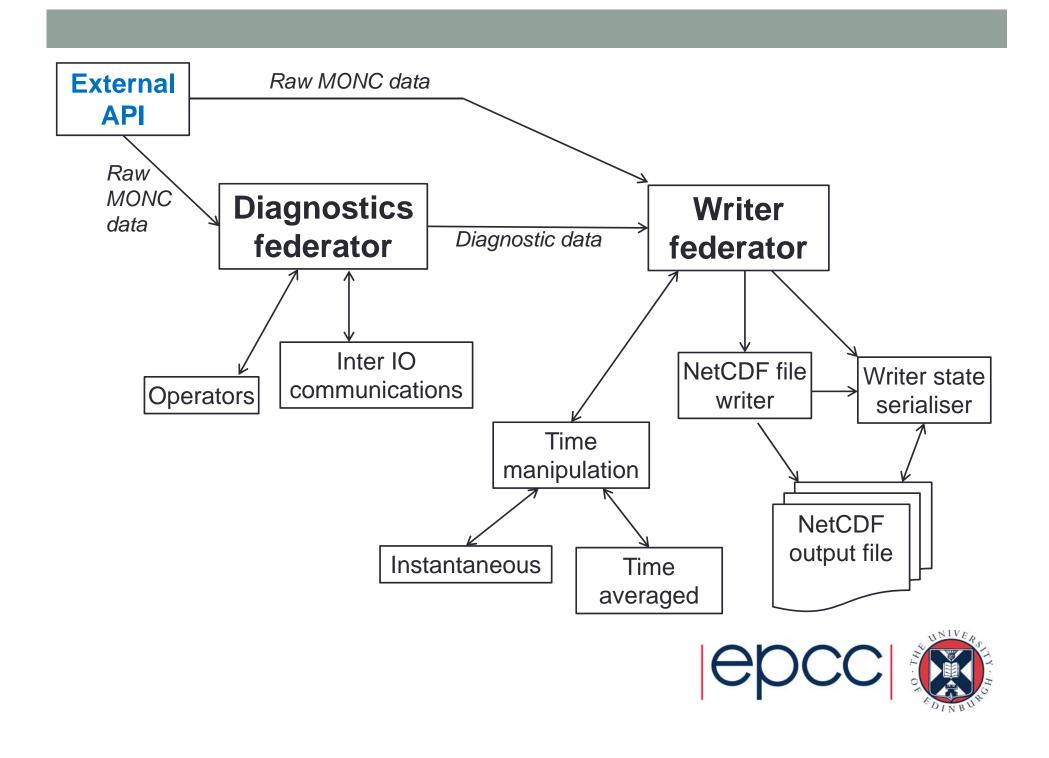
- Numerous existing XML configurations provided which can be included by the user
- Raises the issue of name conflicts
 - Handled by the concept of namespaces

#include "checkpoint.xml"

#include "profile_fields.xml"

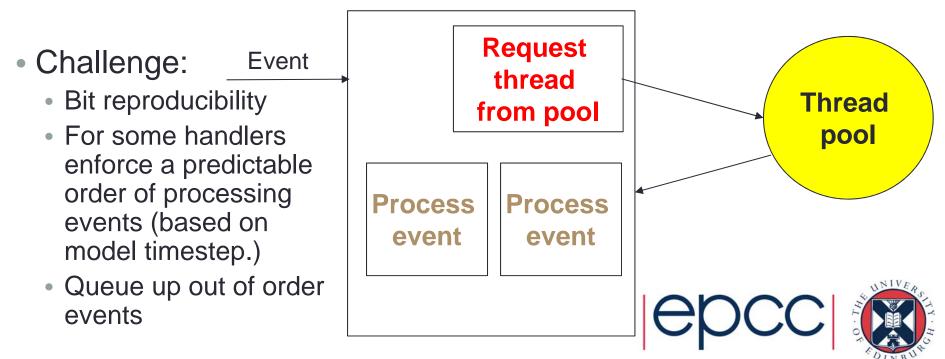
#include "scalar_fields.xml"





Event handling

- The federators and their sub activities are event handlers
 - Process events concurrently by assigning these to these from a pool
 - Aids asynchronous data handling
 - As soon as data arrives process it
 - Internal state of event handlers needs protection (mutexes/rw locks)



Inter-IO communications challenge

External

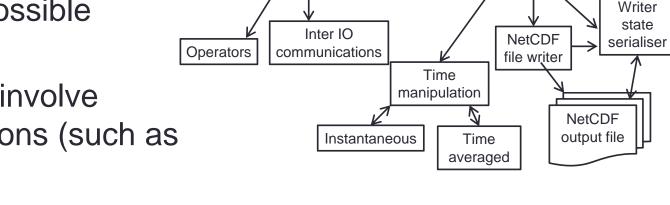
API

Raw

data

MONC

- We promote asynchronicity and processing of events out of order where possible
- Many inter IO communications involve collective operations (such as a reduction)



Diagnostic

data

Raw MONC

Diagnostics

federator

data

- We would like to use MPI, but issue order of collectives matters (i.e. if IO server 1 issues a reduce on field A and then B, then all other IO servers must issue reductions in that same order)
- But ensuring this would require additional overhead and/or coordination
- Solution: Abstract through active messaging

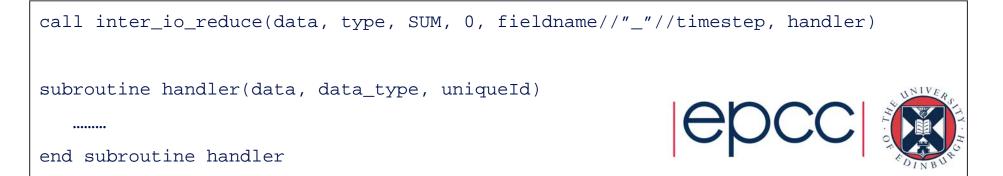


Writer

federator

Active messaging for inter IO comms

- These communication calls additionally provide
 - UniqueID: matching collectives even if they are issued out of order
 - Callback : Handling procedure called on the root when the data arrives
- inter_io_reduce(data, data_type, operation, root, uniqueId, callback)
 - When this reduction is completed on the root, a thread is activated from the pool and calls the handling function (typically in the event handler)
 - The Unique ID here is the concatenation of field name and timestep
 - Built upon MPI P2P communication calls



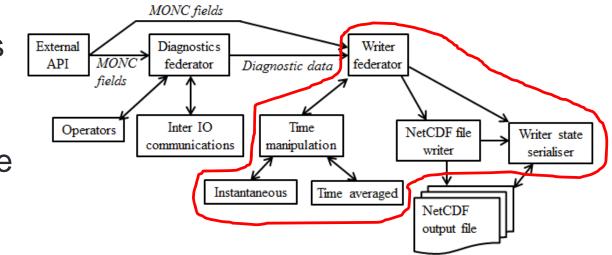
Active messaging for synchronisation

- File writing is done by NetCDF
 - But this is not thread safe, so crucial that only one thread per IO server process calls NetCDF functions concurrently
 - Many NetCDF calls are collective (i.e. will block until called on all processes in the communicator.)
 - NetCDF *close* is an example of this, where each process will block here until same call issued on all other processes
 - Which we don't want, as it will block access to NetCDF (and MPI)
- Active messaging barrier calls the handling function on every process once a barrier has been issued by all processes

```
call inter_io_barrier(filename_uniqueID, closeHandler)
subroutine closeHandler(uniqueId)
call close_netcdf_file(......)
end subroutine closeHandler
```

Checkpointing

- We need to support checkpoint-restart of the IO server and analytics
 - This is challenging due to the amount of asynchronicity, especially in the analytics
- Wait for all analytics to that point to complete and just store the state of the writer federator

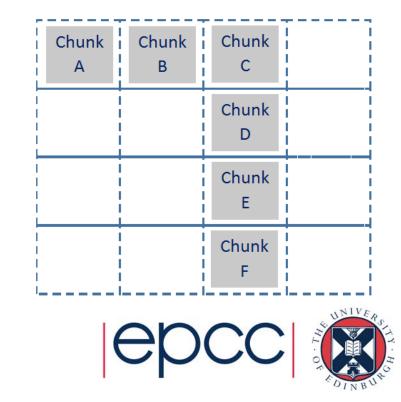


- Two step process
 - Walk the state to determine the amount of memory needed & lock it
 - Serialise state into buffer, write this and unlock

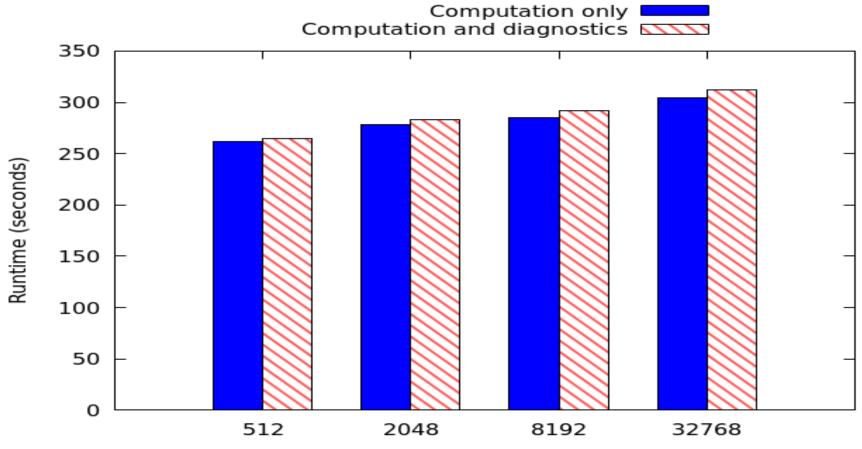


Prognostic writing optimisation

- IO servers servicing many computational cores means that the data is naturally split up
 - For prognostic field writes this can be a problem as it is very inefficient to do lots of independent writes to the file
- Want to perform minimal collective writes instead
 - Search through the domain of local computational cores and merge data chunks together where possible to produce smaller number of large contiguous chunks
 - Must do the same number of writes on every core, some perform empty writes if not enough chunks



Performance and scalability



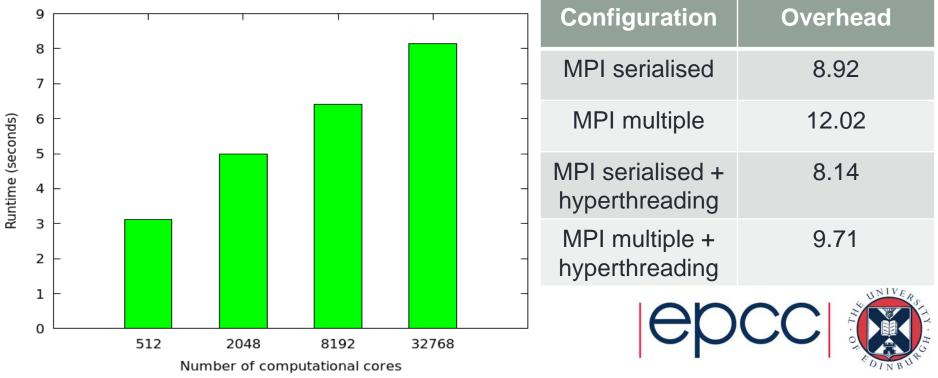
Number of computational cores

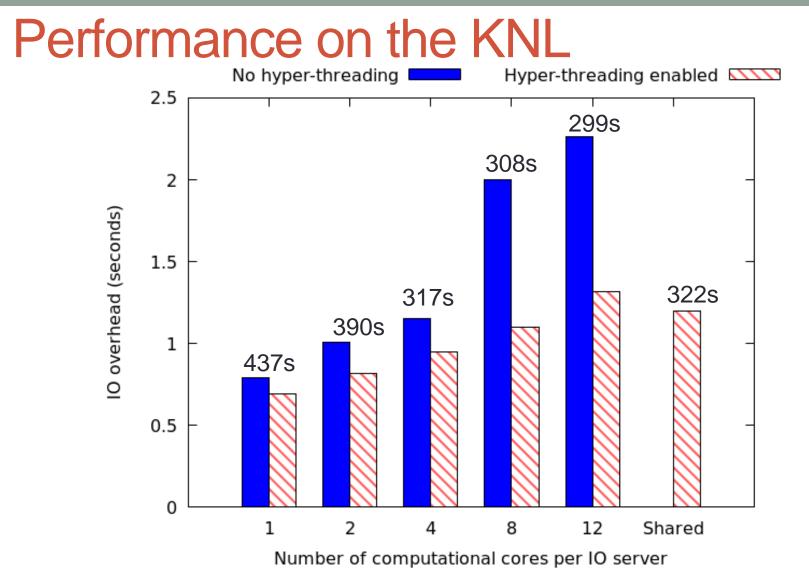
- Standard MONC stratus cloud test case
 - Weak scaling on Cray XC30, 65536 local grid points
 - 232 diagnostic values every timestep, time averaged over 10 model seconds. File written every 100 model seconds. Run terminates after 2000 model seconds.



IO overhead as a metric

- To measure the performance of the IO server and different configurations we adopt an overhead metric
 - This is the time difference from the MONC communication that should trigger a write, to that write having being physically performed





- Cray XC40, 64 core KNL 7210
 - Same stratus test case. 3.3 million grid points
 - As MONC is not multi-threaded can we run one IO server per MONC on the hyper-thread?

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Conclusions and further work

- We have discussed our approach for in-situ data analytics and IO
 - Which performs and scales well up to 32768 computational cores
 - As well as the architecture, challenges and lessons learnt from implementing this
- Extend the active messaging layer to build upon something other than MPI
- Plug in other writing mechanisms such as visualisation tools
- Extract this from MONC to enable others to integrate with their models

