ECMWF's Future challenges in Handling and Manipulating Model and Observational Data

Questions in "Big Data"

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Data Handling

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The Data Chain





Major assimilated datasets





Radiosonde balloons

Receive 300 million observations from 130 sources daily.





Polar, microwave











Meteorological Fields

Operational models produce:

- 13 millions fields daily
- Totalling 8 TB/day





ECMWF products

- 77 million products disseminated ever day, totalling 6 TB.
- Interpolate output fields into user required grids
- Product generation is also subject to a dissemination schedule (time critical)
- Products also served via web visualisation services

7.2	9.9	3.6	0.4	8.3	0.2	0.5	0.1	9.1	6.7
0.3	8.8	1.8	0.5	0.3	0.1	2.7	0.1	7.9	6.9
7.1	9.2	3.6	0.4	8.3	0.2	6.5	3.3	5.5	5.3
2.2	1.1	1.7	0.7	3.5	2.4	0.8	1.9	9.0	6.7
5.1	0.9	1.9	8.9	5.9	0.4	1.5	2.0	7.7	0.7
6.2	0.4	1.4	9.8	9.9	7.7	0.9	3.2	7.2	4.8
8.1	1.4	4.4	0.4	0.3	7.2	3.5	3.4	1.1	9.7
7.0	3.6	4.9	0.7	6.8	1.2	0.1	2.2	6.6	6.0
0.2	7.7	3.6	3.1	8.6	0.5	9.5	0.8	5.6	5.0
3.2	7.2	3.1	0.4	0.9	0.3	0.7	0.4	0.2	0.0





Questions in "Big Data"

"There are no right answers to wrong questions" - Ursula Le Guin



"Big Data is the term for a collection of data sets so large and complex that it becomes difficult to process using on-hand database management tools or traditional data processing applications. The challenges include capture, curation, storage, search, sharing, transfer, analysis and visualization."

"Big Data", Wikipedia, retrieved 2014

"Big Data is high volume, high velocity, and/or high variety information assets that require new forms of processing to enable enhanced decision making, insight discovery and process optimization."

"3D Data Management: Controlling Data Volume, Velocity and Variety", D. Laney, Gartner, 2001

The 3 V's of Big Data



V is for Volume







V is for Volume





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V is for Volume

ECMWF Archive





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V is for Velocity

• ECMWF's archive grows exponentially:



- -r is around 0.5, which is a 50% increase per year
- The rate of added data also grows exponentially at the same rate! $\frac{\partial V_0(1+r)^t}{\partial t} = V_0 log(1+r)(1+r)^t = A_0(1+r)^t$
- In 1995, the size of the archive was increasing at a rate of 14 TB/year.
- In 2014, the size of the archive increases at a rate higher than 65 TB/day with peaks of 100 TB/day



V is for Variety



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Future Challenges

... more of the same?



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Resolution	Grid size	Grid Points	Field Size (in memory)
T319	62.5 km	204 k	1.6 MB
T511	39 km	524 k	4 MB
T799	25 km	1.2 M	9.6 MB
T1279	16 km	2.1 M	16.8 MB
T2047	10 km	8.4 M	67.2 MB
T3999	5 km	20 M	160 MB
T7999	2.5 km	80 M	640 MB

As memory per core diminishes (think GPU's) ...

... this may have serious implications on the interpolation software!



Archive size vs. Supercomputer power



As has been impacting on the archive size...



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Nothing of this is new We have always been dealing with this issues...

What changed?



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"The Free Lunch is Over". H. Sutter, Dr. Dobb's Journal, 30(3), March 2005



But what about "real" performance?



CPU Performance Growth (single-threaded)

"A Look Back at Single-Threaded CPU Performance", J. Pershing Feb 2012



More registers, vector units, branch prediction ...

... but also harder to achieve!



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Storage Density Growth – Multiple Technologies

"Tape based magnetic recording: technology landscape comparisons with hard disk drive and flash roadmaps", R. Fontana et al, IBM Research Division, 2011





HDD Storage Growth

"GPFS Scans 10 Billion Files in 43 minutes". R. Freitas, et al. IBM Research Division, 2011



Volume is linearly proportional to area density Recently follows 25-40% CAGR...

... but transaction rate hasn't kept up!

This means that we may have the capacity, but maybe not the bandwidth ...



• "No Free Lunch" → Improve our software

- Explore new Algorithms that expose ...
 - Concurrent computations (eg. map-reduce)
 - Data locality (eg. FEM discretisations)
 - Computational intensity (CPU usage/MB transferred)
- Software must cope with changes Flexibility
 - Best use of new hardware (eg. use high-level DSL)
 - Unknown future for parallel platforms
 - Be able to adapt to changes in system architecture



Can we do it?

We have already started...

- + OOPS project for Assimilation
- + IFS Co-Array Fortran
- + PantaRhei project



ECMWF's Meteorological Archival and Retrieval System

- A managed archive, **not a file system**
 - Users not aware of the location of the data
 - Retrievals expressed in meteorological terms
- Data is kept forever:
 - Dataset becomes more useful once enough data has been accumulated
 - Deleting old data in an exponentially growing archive is meaningless
- Consists of 3 layers:
 - FDB cache at the HPC level (~80% hit ratio)
 - DHS HDD cache (~80% hit ratio)
 - HPSS Tape system



ECMWF's Meteorological Archival and Retrieval System

- **Fully distributed** (migrated 2012)
 - 15 servers for metadata and data movers
 - 40 PB primary archive
 - 1 PB of disk cache (2.5%)
 - 110 billion fields in 8.5 million files
 - 200 million objects/65 TB added daily
 - 7000 registered users
 - 650 daily active users
 - 100 TB retrieved per day, in 1.5 million requests

Users and # Requests **not** directly under our control...

Scale with # Users / Requests !





A meteorological language

parameter=totype=festep=1levtype=plevels=1grid=2	0110101/to/20110131, emperature/geopotential, orecast, 2/to/240/by/12, ressure levels, 000/850/500/200, /2, 10/20/10/0
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• This request represents 31*2*20*4 = 4960 fields

Indirection is key to Scalability



IFS I/O Layer

As IFS improves its scalability ...

- GRIB encoding is likely to become a bottleneck
 - GRIB encoding requires full field (involves data gather)
 - Currently done within IFS

→ Introduce an I/O layer (indirection)

- Achieve adaptability to changing paradigms:
 - Do data gather on our side?
 - Implement IO Server?
 - Encode GRIB in parallel? Defer encoding?
 - Encode in a parallel format (NetCDF4? Other?)

→ Very important to optimize the whole data chain ←



Other Data Chain Components

Currently under development ...

- Observations
 - COPE project: real time processing
- IFS I/O (cached storage)
 - FDB5: transactional & integrated with MARS
- Interpolation and Product Generation
 - New interpolation package (MIR)
 - ATLAS Framework co-developed with PantaRhei
 - Looking into FEM data-structures and functional spaces

Needing future attention ...

- Visualisation
- Encoding fields (GRIB, NetCDF)



Summary

- **ECMWF** Data Chain faces **the Big Data 3V's** scalability challenges...
- Need to develop **concurrent** approaches to **all** data chain components:
 - Observation Processing
 - Data Encoding
 - Data Storage
 - Interpolation and Product Generation
 - Visualisation
- I/O transaction rates are not keeping up with growth
 - Avoid I/O by pipelining between data-chain components?
 - Move processing closer to the data?
 - Meteorology "Cloud Services"?



We are hiring !

Visit <u>www.ecmwf.int</u> > Employment

- Scalability Program
 Work in the Data Handling Team
- Looking for experts in:
 - High Performance Computing
 - GPU's, Accelerators
 - Algorithms



Come and help us solve these challenges ...



Questions?

* No dwarfs were used in the production of this presentation

**** OK, except maybe one called MapReduce...**

"The Landscape of Parallel Computing Research: A View from Berkeley", Asanovic et al, December 2006 (aka 13 Berlekey Dwarfs)



