Optimizing IFS observation processing: - a case study in scalability enhancement

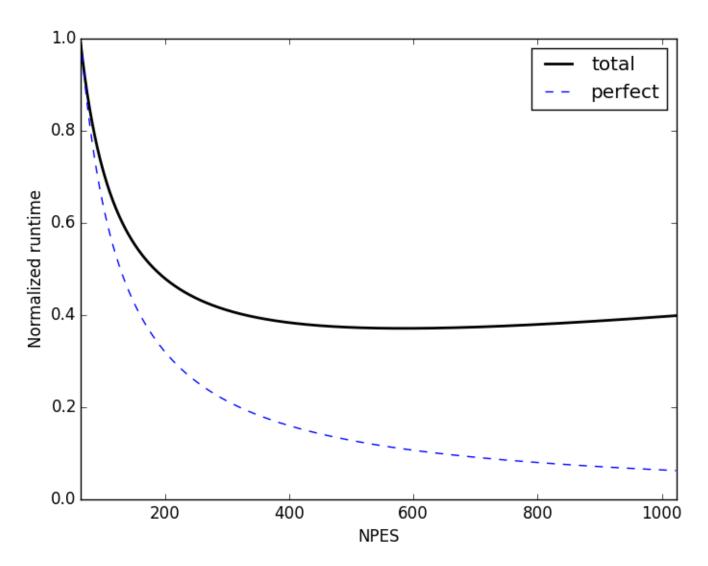
Peter Lean

Thanks - Alan Geer, Deborah Salmond, John Hague and Niels Bormann

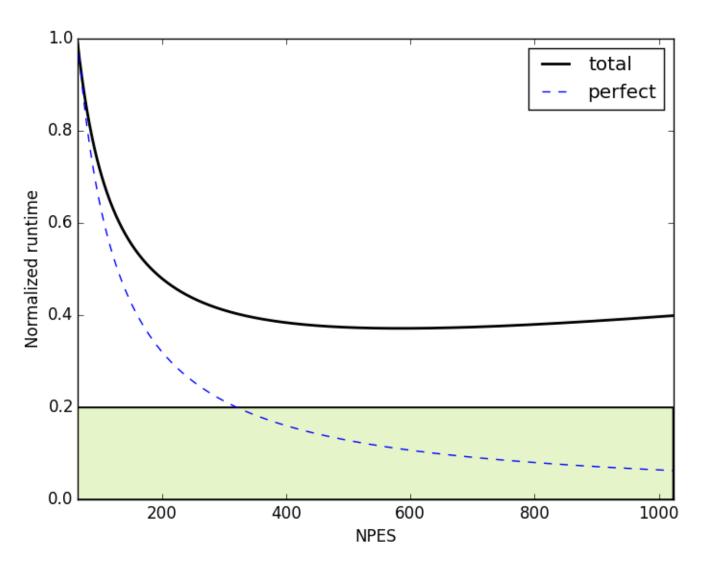
17th Workshop on High Performance Computing in Meteorology

October 2016

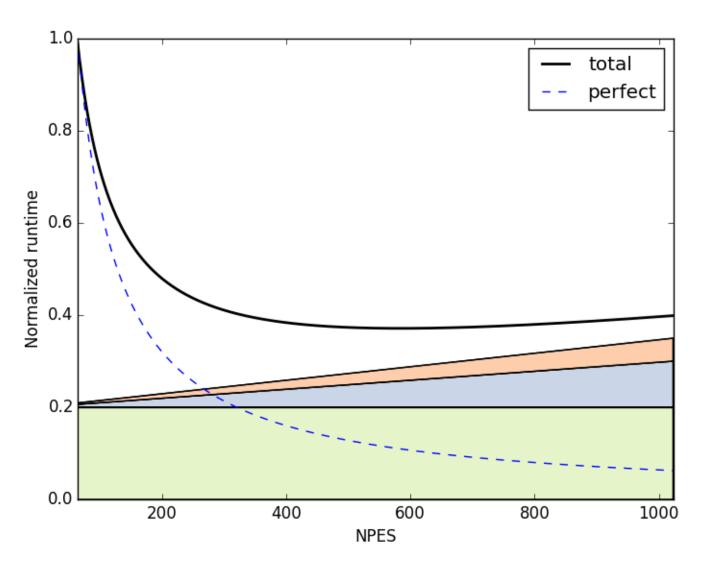




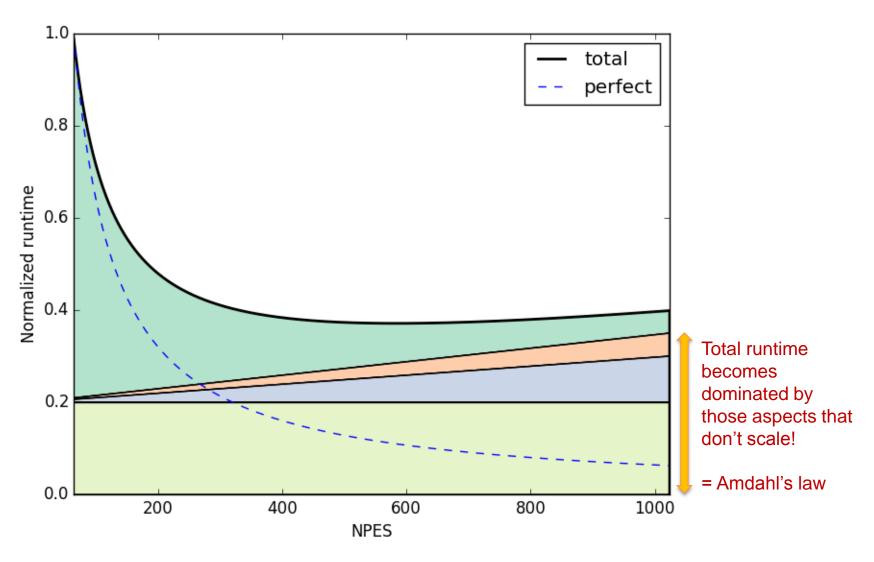




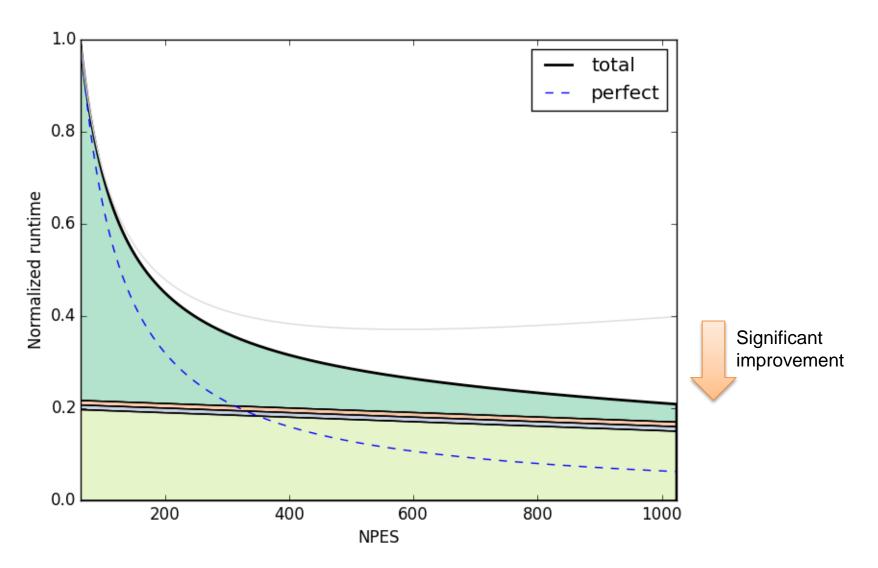














To improve the scalability of an application:

- 1. Identify the top few routines which don't scale well (or blow up)
- 2. Fix them.



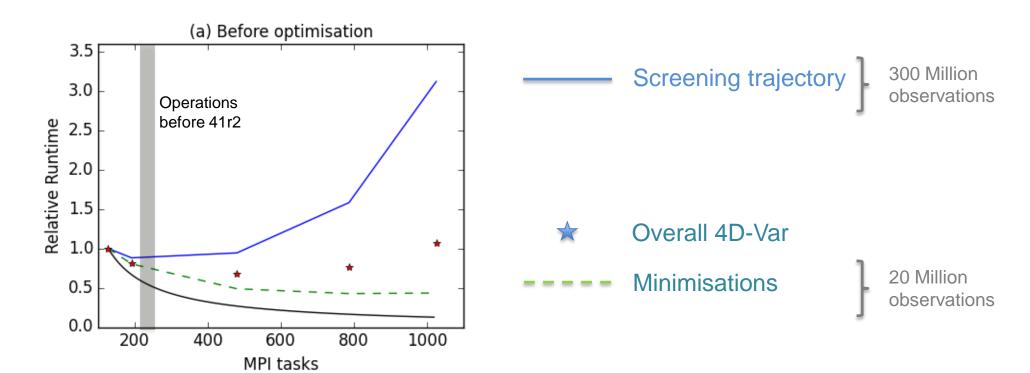


The hard part:

Fixing scalability of the components that blow up – a case study



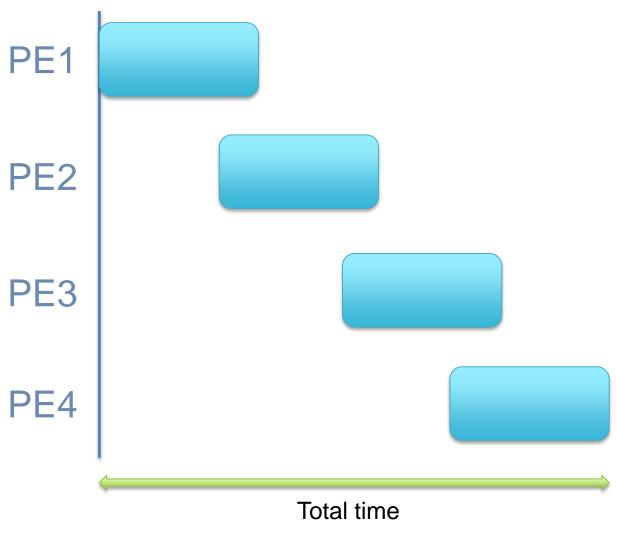
Problem with scalability of observation processing on Cray XC30



1. Parallel I/O in Observation Database



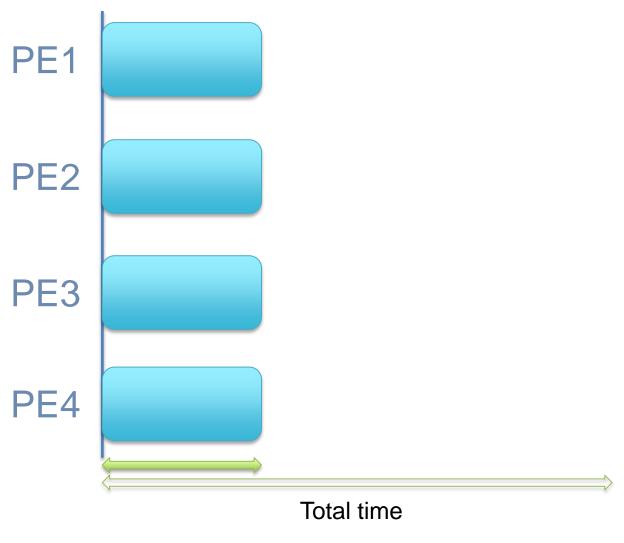
Parallel I/O for Observation Database



- Only subset of tasks do I/O in ODB to reduce strain on file system.
- Message passing sends the data to the PE doing I/O
- Ordering of loops in message passing and I/O code was inefficient
- Spent lots of time waiting at MPI barriers for messages to be received

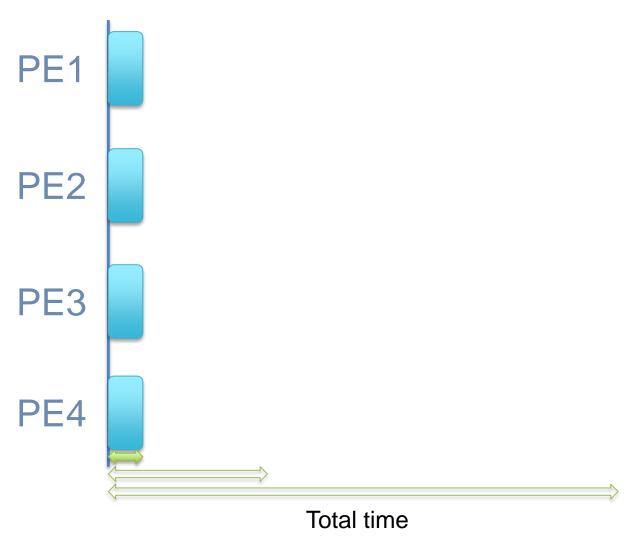


Parallel I/O for Observation Database



- Refactored the message passing and I/O loops to minimize time spent at barriers.
- Noticed that majority of time still spent on non I/O aspects.

Parallel I/O for Observation Database



Changed lookup table:

- was O(n) number of tasks
- now O(1)

Two issues:

- 1. Time waiting at MPI barriers
- 2. Slow lookups

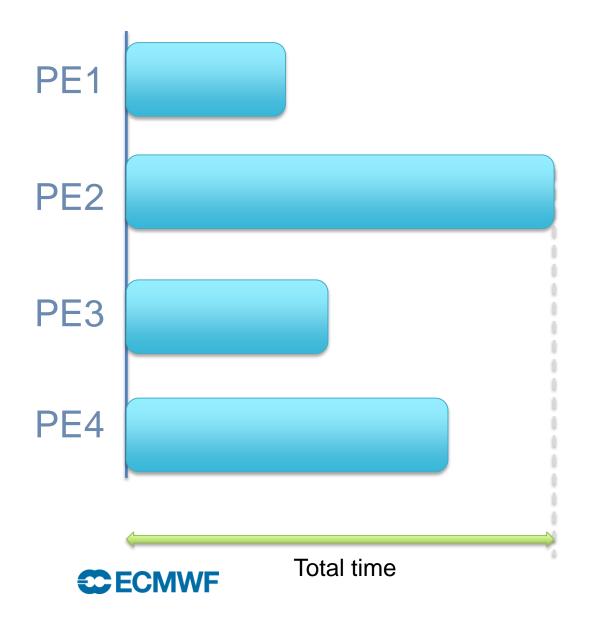
Amplification:

- barrier issue amplified the slow lookup issue
- lookups effectively became sequential.

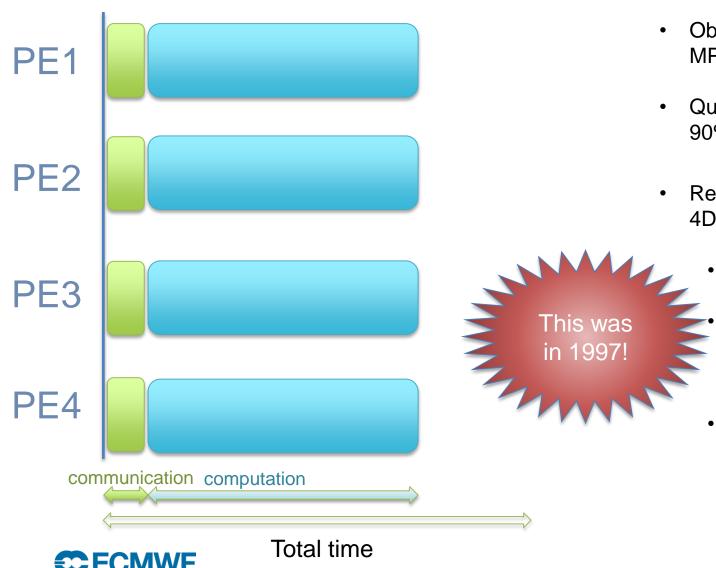
e.g. 0.1s per lookup x 1000 MPI tasks = 100s!



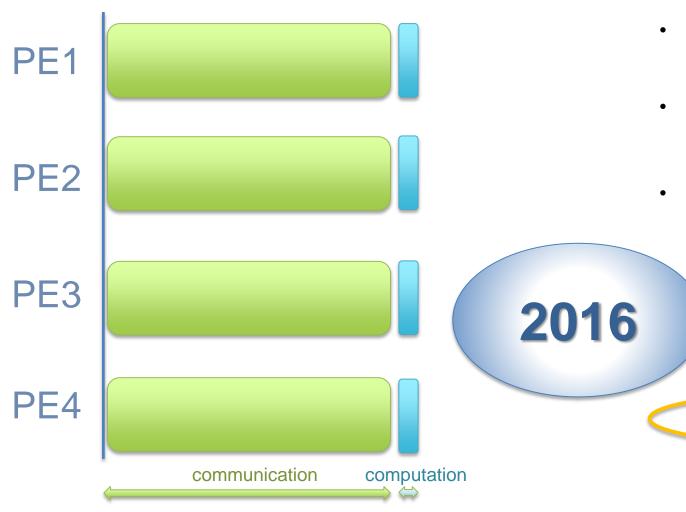




- Observations are distributed across MPI tasks in ODB / IFS.
- Quality control / screening removes 90% of the original observations.
- After Quality Control, the observations may no longer be well balanced.

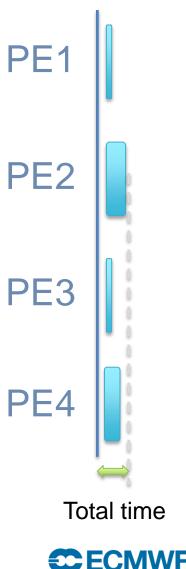


- Observations are distributed across MPI tasks in ODB / IFS.
- Quality control / screening removes
 90% of the original observations.
- Re-distribute the active observations in 4D-Var evenly across MPI tasks:
 - Communication required
 - Computational time on each task processing the observations is more even.
 - Overall time runtime is reduced.



- Observations are distributed across MPI tasks in IFS.
- Screening / quality control removes 90% of the original observations.
- Re-distribute the active observations in 4D-Var evenly across MPI tasks:
 - Communication required
 - Computational time on each task processing the observations is more even.
 - Overall time runtime is reduced.

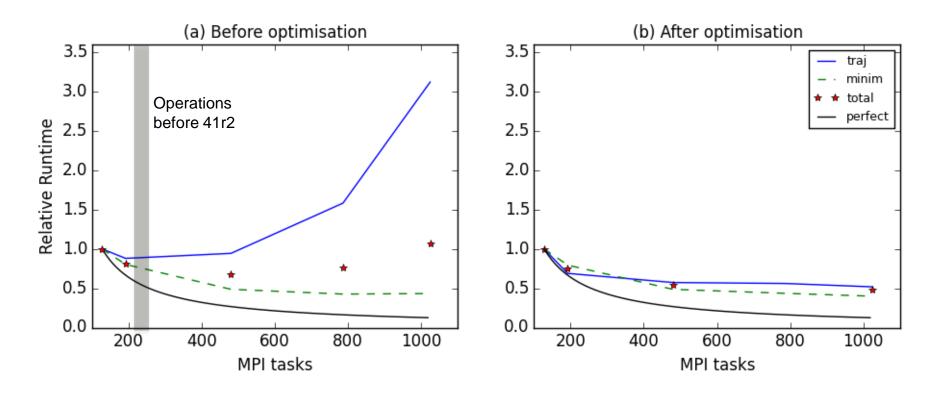




- In 2016, the relative cost of computation has gone down (scaled away as running on far more PEs).
- The load balancing costs more than it saves.

Optimisations which gave improvements 10 years ago may no longer be giving the same benefits today!

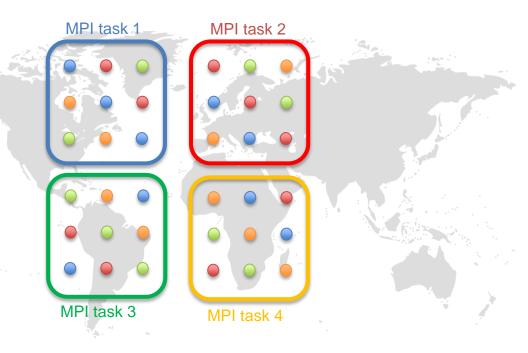
Improved scalability:



- Scalability of first and final trajectories significantly improved
- Running on 1000 MPI tasks, 4D-Var sped up x2
- But scalability of 4D-Var still not perfect...

Future directions: data locality → reduced communications

- Data locality: reduce costly data transfer by doing the calculations where the data resides.
- 4D-Var compares each
 observation against the model equivalent
 - Two "Big" datasets
 - Generally not co-located.



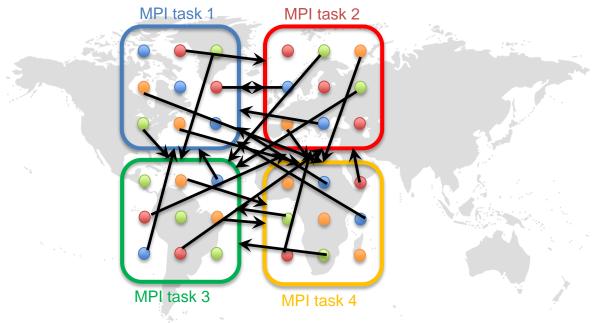
Model domain decomposed geographically across MPI tasks

ODB distributes observations randomly across MPI tasks



Future directions: data locality → reduced communications

Comparison between observation and model usually requires communication



	Runtime [s]
ODB I/O	3s
Interpolation + communication of model profiles at observation locations	34s
Observation operator (calculation of observation equivalents)	3s



Future directions: data locality → reduced communications

MPI task 1

MPI task 2

Geographical partitioning of the observations in ODB to match the model domain decomposition should reduce required communication

MPI task 3

MPI task 4

⇒ Faster, more scalable 4D-Var (hopefully)



Improving scalability: short v long term perspectives

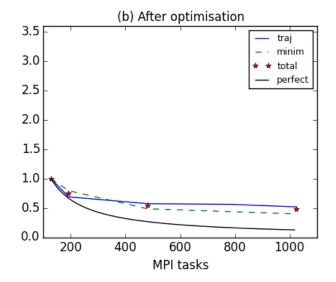
Short/medium term: 1-2 years

- Optimising existing code : spend ⇒ short term payoff
- Can give us several years headroom before scalability becomes a problem and can deliver significant performance improvements today.
- But it won't solve the major scalability challenges we face.

Long term: 2 years +

- **Invest** in major re-writes / restructuring:
 - ⇒ major scalability improvements (but maybe not for another 5 years)

e.g. OOPS: weak constraint time parallel 4D-Var





Take home messages:

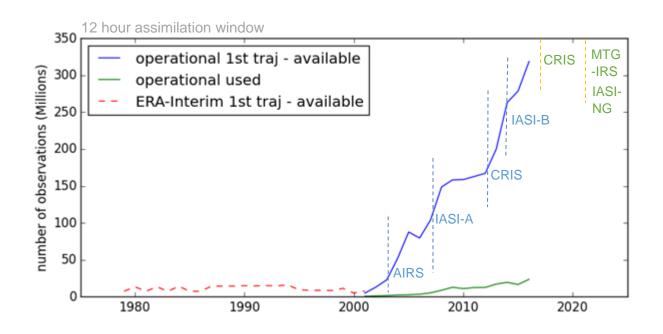
- Sometimes a few poorly scaling components have a disproportionate impact on overall scalability of an application.
- In this case, minimising unnecessary communications improved scalability.
- Optimisations which gave improvements 10 years ago may no longer be giving the same benefits today.
- Optimisation isn't a one off; it should be part of business as usual.



Thanks for listening



Context: Trends in number of observations



Other factors:

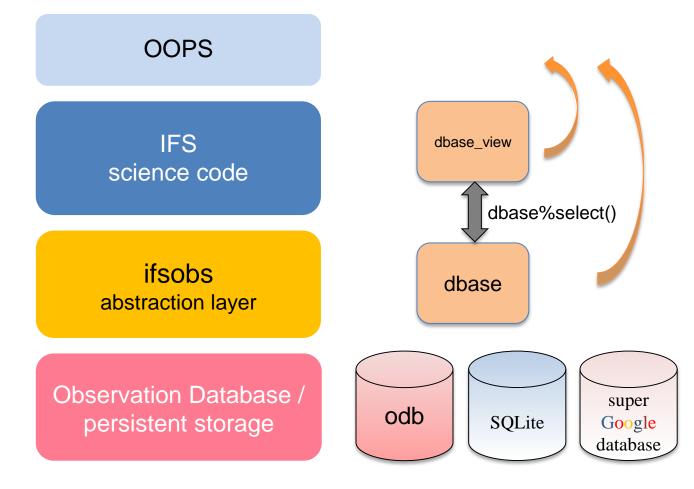
- Long window 4D-Var? 5 day = x10
- Spatial error correlations? Less thinning.
- Clear sky only → all-sky.

- Explosion in availability of satellite data since 2001.
- Mostly from infra-red sounders.
- Only around 5-10% are actively assimilated:
 - most are cloud-screened, thinned or blacklisted.
- Trend expected to continue with new CRIS, MTG-IRS, IASI-NG
- Projection for next 10 years:
 - x2 (current usage trends)
 - x10 (more aggressive usage)



ifsobs

- a data acces layer for observations in IFS:



Decouples IFS from the underlying database / file format; improved **modularity** of IFS.

IFS only interacts with dbase and dbase_view objects, **not** with the underlying database directly.

