Benchmarking in AIPS++

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Approach to benchmarking

• Develop easy-to-run tools

Populate the AIPS++ Benchmark module

Make same and repeatable measurements

- Develop conceptual understanding of the compute, I/O and other major costs of the algorithms
- Reduce (eliminate) extraneous parameters
- Understand performance curves (and not spot measurements) as a function of the problem size
- Start from the simplest, but realistic problem size

Work so far...

• Image re-gridder optimization

- Low level optimization (Brouw)
- Achieved a factor of ~10 improvement. Real-time image re-gridding possible
- <u>Interferometric imaging and calibration</u>: Study of the algorithms and the associated I/O costs
- Equivalent scripts for other popular packages (AIPS, Miriad)
- **Profiling:** Very useful and in progress using standard GNU tools
- Regular profiling builds of AIPS++
- **Dedicated computer** for benchmarking
 - Need isolation and controlled environment

Work so far...

<u>Developed benchmark.g</u>

Control on algorithms, problem size, data size, Table system parameters, memory model...

ALMA data size tests

- First cut at ALMA TI filler (Rusk)
- Iramcalibrater compared with — CLIC.
- Well known bottle-necks being optimized
- Profiling and optimization in progress

Common part of the benchmark code: 'CC-SF-VLAC-U1M-SP1-'							
Benchmark code	Dataset	N _{spw}	Stokes	Wgt.	N _{pixel}	N _{chan}	N _{clean}
I-UN-512-C1-1000	vlac1M	1	Ι	UN	512	1	1000
IQUV-UN-512-C1-1000	vlac1M	1	IQUV	UN	512	1	1000
I-UN-1024-C1-1000	vlac1M	1	Ι	UN	1024	1	1000
IQUV-UN-1024-C1-1000	vlac1M	1	IQUV	UN	1024	1	1000
I-UN-2048-C1-1000	vlac1M	1	I	UN	2048	1	1000
IQUV-UN-2048-C1-1000	vlac1M	1	IQUV	UN	2048	1	1000
I-NA-512-C1-1000	vlac1M	1	Ι	NA	512	1	1000
IQUV-NA-512-C1-1000	vlac1M	1	IQUV	NA	512	1	1000
I-NA-1024-C1-1000	vlac1M	1	I	NA	1024	1	1000
IQUV-NA-1024-C1-1000	vlac1M	1	IQUV	NA	1024	1	1000
I-NA-2048-C1-1000	vlac1M	1	Ι	NA	2048	1	1000

Table 1.1381: Table for continuum VLA 1M rows dataset benchmark.

Table 1.1384: Table for VLA calibrator benchmark (for 'G' and 'D' Jones).

1 IOUV

NA

2048

IQUV-NA-2048-C1-1000 vlac1M

1

1000

Benchmark code	Data	Compressed?	Jones	NAnt	SNR	NSolInt
CALVLAU-G-27-10-100	calvlac27s10.fits	Nope	G	27	10	100
CALVLAU-D-27-10-100	calvlac27s10.fits	Nope	D	27	10	100

Table 1.1385: Table for FITS UV read benchmark (import visibility data into AIPS++ from FITS files). Common part of the code is ' FUV-RD-VLA'.

Benchmark code	Dataset	Mode	N _{chans}	Compressed?	Data size
C-U125K-C1	vlac125k.fits	Continuum	1	False	125K
C-U1M-C1	vlac1m.fits	Continuum	1	False	1M
L-U125K-C64	vlal125K64Chan.fits	Line	64	False	125K

Results: General

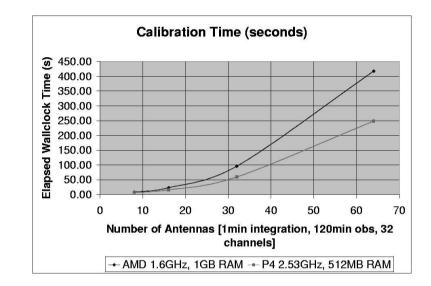
- AIPS performance curves were taken as the 'baseline' (Miriad is 2x faster than AIPS though!)
- AIPS++ default settings too conservative
 - Memory/CPU usage monitoring indicated memory model too conservative for present day computers
 - Default settings can be (should be?) made computer-resource-aware
- **Burnt-in assumptions for memory model** in some critical parts of the code
- Redundant (expensive) operations
- Profiling ultimately dug out a few real & 'dormant bugs'

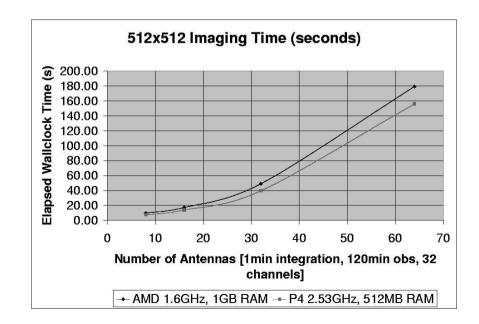
Results: Well-known mistakes

- **Tiling** in the construction of Tables is an important parameter
- Array access pattern: can be expensive if done the wrong way
- Inadvertent use of copy vs. reference semantics is expensive
 - Results into excessive memory copies
 - Use of objcopy() should be minimized
- Table creation may be expensive
 - Used by ArrayLattice in tiling mode
 - Minimize Table creation inside tight loops
- Strictly use only OSInfo() class for memory model

Results: ALMA Tests

- Performance studied as a function of data-rate
 (Rusk)
- Imaging appears to scale well
- Hardware/compilers factor in as significant parameters
- Understand the scaling as a function of various parameters
- More work needed to understand the differences in these curves



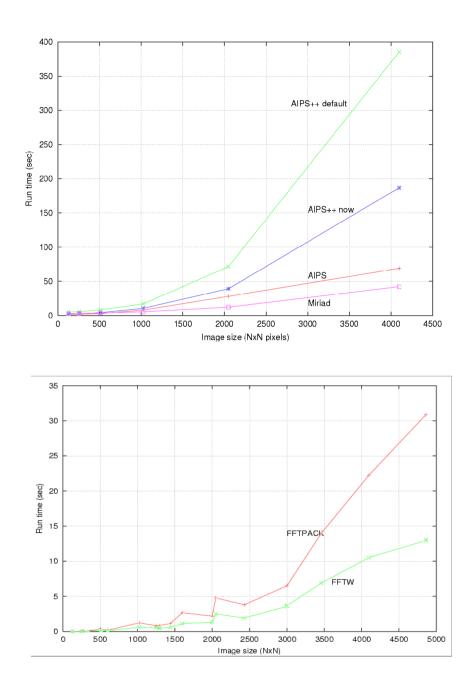


Results:IRAM Calibrater Tests

- *First cut at iramcalibrater benchmarking* (Rusk/Lucas/Golap) No. of antennas: 64 CLIC Calib. cycle AIPS++ AIPS++ (no plots) (sec) (sec) (sec) RF 1818 2187 2646 Phase 1404 804 942 2502 2076 708 Amp
- AIPS++ overheads insignificant
- AIPS++ code not yet fully optimized
- *RF* & *Phase solvers use the same engine in AIPS++ and CLIC: runtime within 10-20% of each other*
- Amp: same solver, but AIPS++ includes apply()

Results: Imager

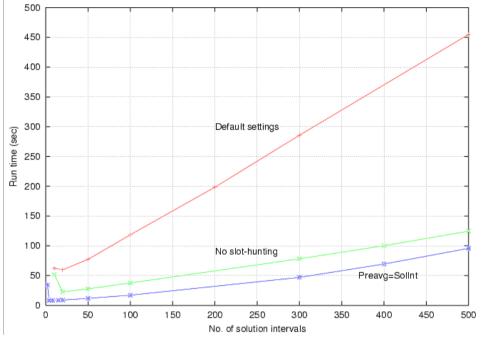
- Clark-CLEAN algorithm (one of the many in AIPS++)
- Recognize 'minor' /subtle differences in the implementation
 - AIPS uses Cotton-Schwab variant
- **FFTW**: optimize data locality and in-memory copies
- Threaded FFT: useful on ubiquitous multi-CPU machines



Results: Calibrater

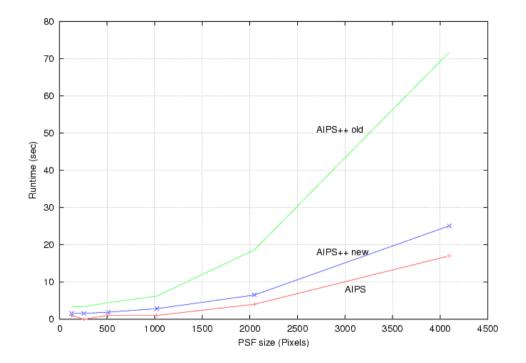
- Gain and band-pass calibration was not optimal (fixed)
- Slot hunting algorithm was non-optimal (fixed)
- Profile dominated by lower level calls
- More work needed once these major bottlenecks are removed

Gain and polarization leakage calibration comparable to AIPS till ~200 solution intervals.



Results: I/O test (not well understood)

- Measure I/O costs as seen by the AIPS++ 'algorithm layer'
- Table system's cache-hit rate was >99%
- Need more work



Lessons learned

- Benchmarks are a measure in a multi-dimentional space
 - Requires careful analysis and understanding of the software and the hardware platform
 - Various axis are not often orthogonal
- Benchmarking and optimization is better done closer to code development than much later.
- **Profiling** is the single most useful tool.
- No fundamental hot-spots (IMHO!)

...continued

- AIPS++ framework is highly configurable
 - Memory model is user configurable
 - Table I/O is configurable (need to bring it to the user level?)
- Smarter automatic (computer resource-aware) settings is an enticing possibility
- Better Technical documentation
 - Cost analysis for developers
 - Do's and Don't 's for developers