Advances of the profiling & optimization of the LSST Pipeline

Antoine BERNARD (LPSC, Grenoble, France), Q. Le Boulc'h, J. Bregeon, D. Boutigny, F. Hernandez, D. Parello, G. Mainetti

Science Pipelines Team Meeting, Feb. 12th 2025

#### LSST Pipelines Optimization Summary)

- Context:
  - Key Motivations: Reducing resource use (CPU time, memory, storage) for finance, ecology, performance.
  - Code Context: Open-source code developed primarily by the Rubin project (US), many different computing nodes (USA / UK / France (CC-IN2P3))
  - **Profiler**: required for reliable **metrics**. Some requirements (Low overhead, native code compatibility, interactive visualizations, ...)

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  - **Mixed Language**: Difficulty profiling Python/C++ together (Python's memory management complicates analysis).
  - Overhead: Profiling tools significantly slow runtime.

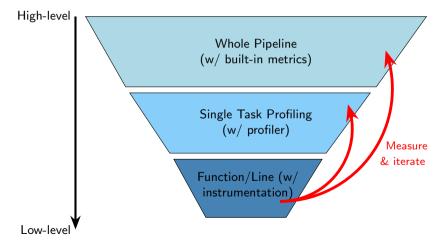
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- Next steps:
  - Set-up test case: Profile with realistic datasets.
  - Start optimization: Piff package or forcedPhotCoadd seems promising targets.

### LSST Pipelines Optimization

#### **Profiling Strategies**

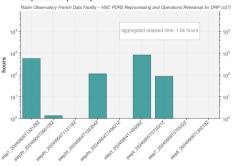
- Broad-to-specific profiling to isolate bottlenecks.
- Improvement shall be measured
- Process is iterative and empirical



- LSST pipelines has inbuilt metrics about every task, providing memory used and time elapsed (as well as input identifiers)
- already used to analyze ressource usage during DP0.2
- re-used on a partial HSC PDR2 run, for test purpose

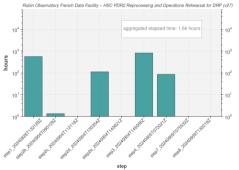
Elapsed time per step

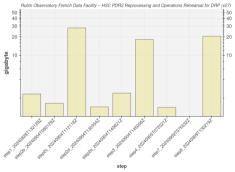
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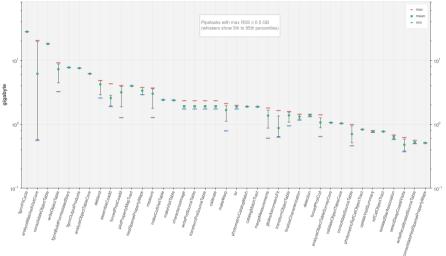




#### Maximum RSS per step

#### Memory consumption by pipetask

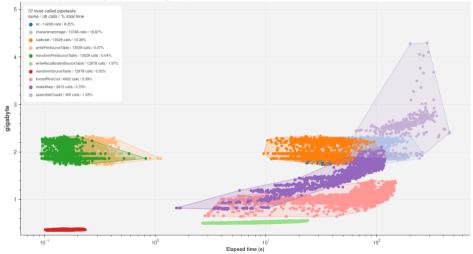
Rubin Observatory French Data Facility - HSC PDR2 Reprocessing and Operations Rehearsal for DRP (v27)



pipetask

#### Max memory used versus elapsed time, grouped by task,

Rubin Observatory French Data Facility – HSC PDR2 Reprocessing and Operations Rehearsal for DRP (v27)



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#### Elapsed Time $^{\odot}$ : 719.315s $\geq$

⊙ CPU Time <sup>⑦</sup> :	698.510s
Effective Time <sup>(2)</sup> :	696.870s
Spin Time <sup>®</sup> :	1.640s
Overhead Time <sup>(2)</sup> :	Os
Total Thread Count:	62
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This section lists the most active functions in your application. Optimizing these hotspot functions typically results in improving overall application performance.

Function	Module	CPU ③ Time	% of CPU ③ Time
func@0x63ea0	libm.so.6	41.466s	5.9%
Transform	libast.so.9	38.413s	5.5%
$boost::math::tools::detail::evaluate\_rational\_c\_imp$	libmeas_base.so	29.715s	4.3%
DOUBLE_onemultadd	_sigtools.cpython-311-x86_64-linux- gnu.so	21.842s	3.1%
func@0x6b950	libm.so.6	20.688s	3.0%
[Others]	N/A*	546.386s	78.2%

\*N/A is applied to non-summable metrics

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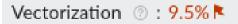
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  - tracked one consuming call stacks: sincosf64x and sqrt function used in boost::math::detail::bessel\_j1,
  - in meas\_base package, in src/SincCoeffs.cc, modified calcImageKSpaceCplx (see on Github),

// compute the k-space values and put them in the cing array	// compute the k-space values and put them in the cimg array
double const twoPiRad1 = geon::TWOPI * rad1;	double const twoPiRad1 = geom::TWOPI + rad1;
double const twoPiRad2 = geon::TwOPI + rad2;	double const twoPiRad2 = geom::TWOPI + rad2;
double const scale = (1.0 - ellipticity);	double const scale = (1.0 - ellipticity);
	// pre-compute rotation matrix
	double const cosAng = ::cos[posAng):
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	// pre-compute fftshift and kpss
	double const inv_wid = 1.0 / wid;
	std:/vectorsint> fftShifted/wid);
	std::vector=double= ksos(vid):
	for (int i = 0; i < wid; ++i) (
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	wheatil = (state_cast=cast= can) = tut_stat
	// pre-compute k
	std::vector:double> kValuesFlat(wid + wid);
for (int iY = 0: iY < wid: ++iY) {	for (int $Y = 0$ ; $Y < wid$ ; $++iY$ ) {
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<pre>std::pair<double, double=""> coo = rotate(kx, ky, posAng);</double,></pre>	double const kxr = cosAng + kx + sinAng + ky
double kxr = coo.first:	double const kyr = - sinAng * kx + cosAng * ky
double kyr = coo.second;	
// rescale	// rescale
<pre>double const k = [:sqrt[kxr + kxr + scale + scale + kyr + kyr];</pre>	<pre>kValuesFlat[fr + wid + fX] = std::sqrtikxr + ker + scale + scale + kyr + kyri;</pre>
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- Idea in this:
  - Refactor code for simplification and enable auto-vectorization
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- Expected results:
  - More memory usage because pre-computation
  - Less runtime thanks to vectorization
  - Biggest chunk of time unaffected yet, cause Bessel function not vectorizable.

### Optimization results Primary results

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Effective Time <sup>(*)</sup> :	696.870s
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Total Thread Count:	62
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Vectorization (2): 9.5%	<b>K</b>

Antoine BERNARD Profiling & optimization

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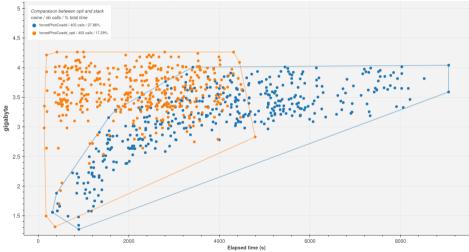
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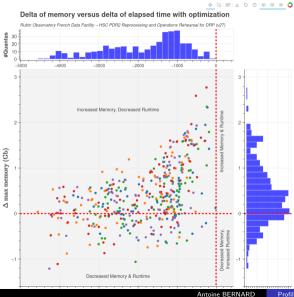
#### Optimization results Global results

#### Max memory used versus elapsed time, grouped by task,

Rubin Observatory French Data Facility – HSC PDR2 Reprocessing and Operations Rehearsal for DRP (v27)



#### Optimization results Global results



- Colors correponds to band,
- Every job runtime decreased,
- Approx 3/4 of the job memory increased,

	$\sum t_i$ (h)	$\sum m_i$ (Gb)	$\sum (t_i \cdot m_i)$ (Gb.h)
Whole Pipeline			
Base	1617.1	157045.7	4013.7
Opti	1410.5	157213.1	3367.1
Diff	206.6	-167.4	646.6
Relative (%)	-12.8%	0.1%	-16.1%
forcedPhotCoadd Task			
Base	450.5	1282.5	1525.5
Opti	243.9	1449.9	878.9
Diff	206.6	-167.4	646.6
Relative (%)	-45.9%	13.0%	-42.4%

- Primary results: 100s gain on a 700s run, seems great & code is more vectorized !
- Global results: -45.9% runtime for forcedPhotCoadd & -12.8% for the whole pipeline... for approx. 20 lines of codes

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Way too much performance gain ! That's suspicious !

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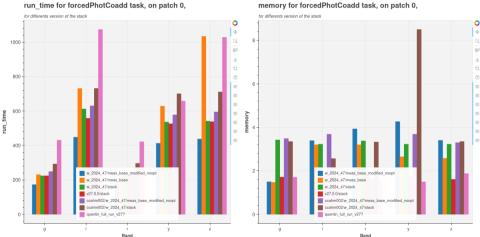
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What could have gone wrong ?

- Code improved between v27 and w\_2024\_47 ?
- Difference between interactive server and bacth-server ?
- Huge variability in runtime inherent to the processing ?

- w\_2024\_47/meas\_base\_modified\_noopt: Using w\_2024\_47 of the stack, with package meas\_base modified with base compilation option, run with sbatch.
- w\_2024\_47/meas\_base: Using w\_2024\_47 of the stack, with package meas\_base with base compilation option, run with sbatch.
- w\_2024\_47/stack: Using w\_2024\_47 version of the stack, run with sbatch.
- v27.0.0/stack: Using v27.0.0 version of the stack (dated approx. June 2024), run with sbatch.
- ccahm002/w\_2024\_47/meas\_base\_modified\_noopt: Using w\_2024\_47 of the stack, with package meas\_base modified with base compilation option, run on ccahm002 interactive server.
- ccahm002/w\_2024\_47/stack: Using the w\_2024\_47 version of the stack, run on ccahm002 interactive server.
- quentin\_full\_run\_v27?: Results obtained from HSC-PDR2 run made by Quentin in early September 2024, using version v27.0.0. No info on the run's specifics.

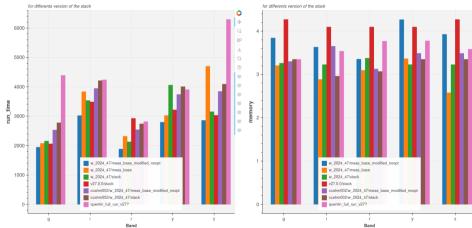
#### Patch 0



memory for forcedPhotCoadd task, on patch 0,

#### Machine impact Interactive vs batch, & multiple stacks version

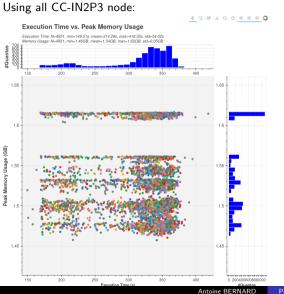
#### Patch 40



run\_time for forcedPhotCoadd task, on patch 40,

memory for forcedPhotCoadd task, on patch 40,

#### Machine impact Multiple run using slurm



- Some memory threshold
- Runtime variability does not seems correlated to a particular server

1.5

1.45

130

140

150

160

Run time (e)

170



#### Using 2 reserved nodes, 1 job at a time:

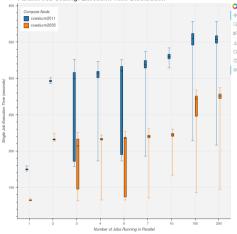
1.5

1.45

180

#### Using 2 reserved nodes:

Parallel Job Scaling: Execution Time Distribution



- Difference between two nodes match nodes specs !
- Runtime bump even at 2 jobs in parallel seems to be slurm putting multiple thread on the same physical core,
- Hyperthreading ?
- Time per job in increase,

#### Notebooks here & more results here.

### Summary

- Need to control the environnement to properly measure performance gain ...
- ... but we need to ensure our gains translate into a production environnement !

or

- $\bullet\,$  We could just run  $\tilde{1}00$  of runs to get a distribution and statistics each improvement,
- Coud work for short runtime tasks ...

### Summary

- Need to control the environnement to properly measure performance gain ...
- ... but we need to ensure our gains translate into a production environnement !

or

- We could just run  $\tilde{1}00$  of runs to get a distribution and statistics each improvement,
- Coud work for short runtime tasks ...

Strategy:

- Dedicate a machine to pursue testing on, simple fast case
- Test again with inputs thats take longer to process,
- Once improvements observed, tests how it scale on a loaded machine,