An Initial Framework for Prototyping Radio-Interferometric **Imaging Pipelines**

<u>Sunrise Wang, Nicolas Gac, Hugo Miomandre, Jean-Francois Nezan,</u> Karol Desnos, Francois Orieux









[1] https://www.skao.int/

Introduction

Prototype framework to estimate resource usage of Radio-Interferometric imaging pipelines

Aimed at large radio-telescopes e.g. SKA

- Large amounts of data
- conflicting design restrictions

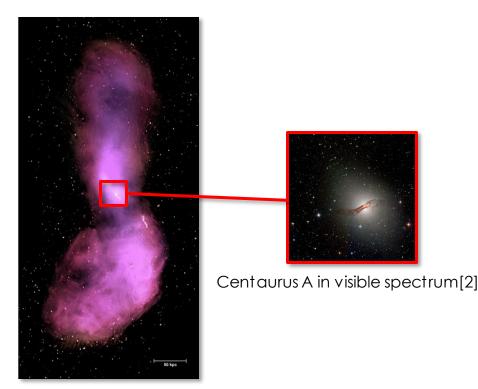


The square kilometer array in south africa (left) and australia (right)[1]

Aid in designing supercomputer hardware and software architecture

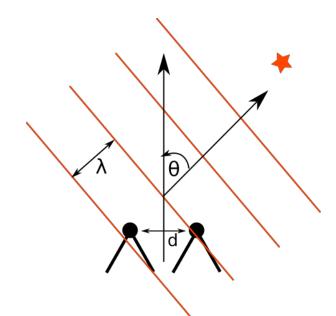
Radio-Interferometry

Seeing the universe in radio



Centaurus A @ ~1.4Ghz (z=~0.0018) [1]

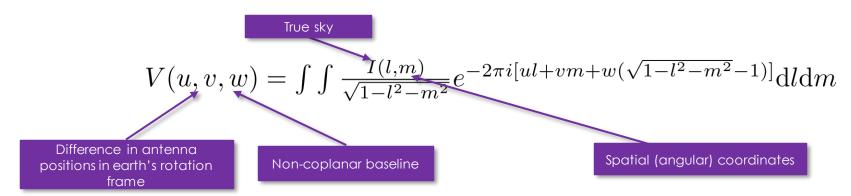
Antenna arrays + Radio-Interferometry allows better angular resolution and sensitivity compared to single dishes

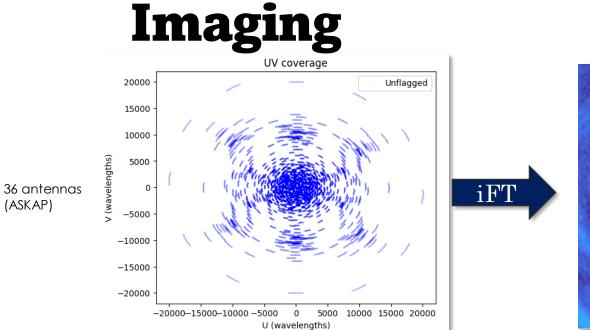


Sampling the Sky

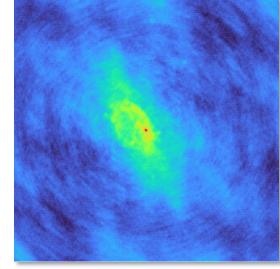
Sample by correlating antenna pairs ie. baselines

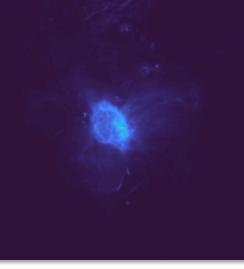
Each sample (ie. visibility) can expressed as:

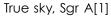




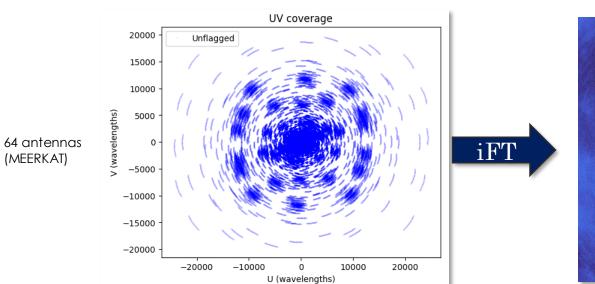
(ASKAP)







Imaging pipeline corrects for artefacts Limited by number of samples and sensitivity

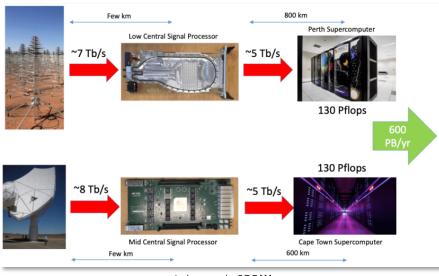


٠

Increasing the number of antennas

Increases amount of data quadratically!

$$n_{vis} = \frac{n_{ant}(n_{ant}+1)}{2}$$



AntennastoSDP[1]

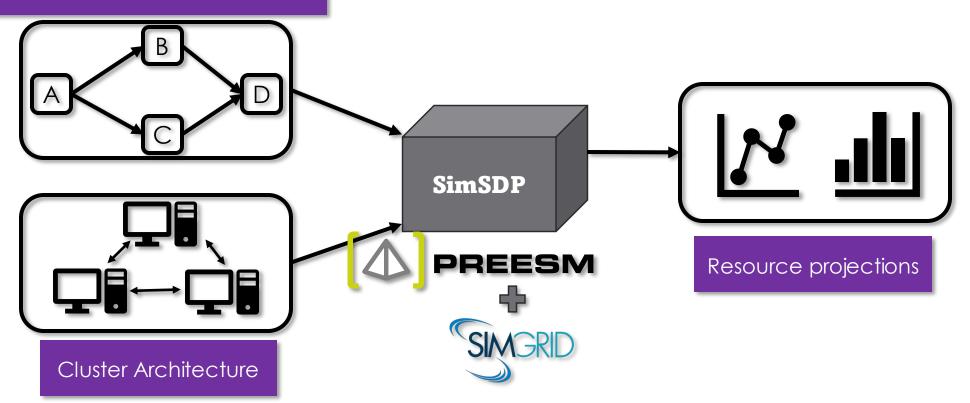
Science Data Processor ingest stream projected to be around 0.4 TB/s ≈ 34.5 PB/day

- Storage expensive, hard time constraints
- Introduces a lot of data-transfer overhead
- Energy costs
- Different pipelines for different types of science (e.g. continuum vs spectral line)

SimSDP

Aid in design of software and hardware architecture by simulating resource usage

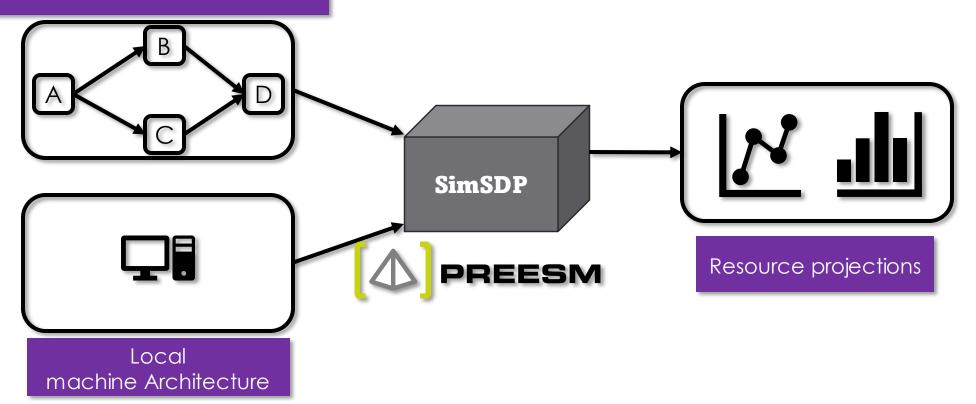
Dataflow Pipeline Descriptions



SimSDP

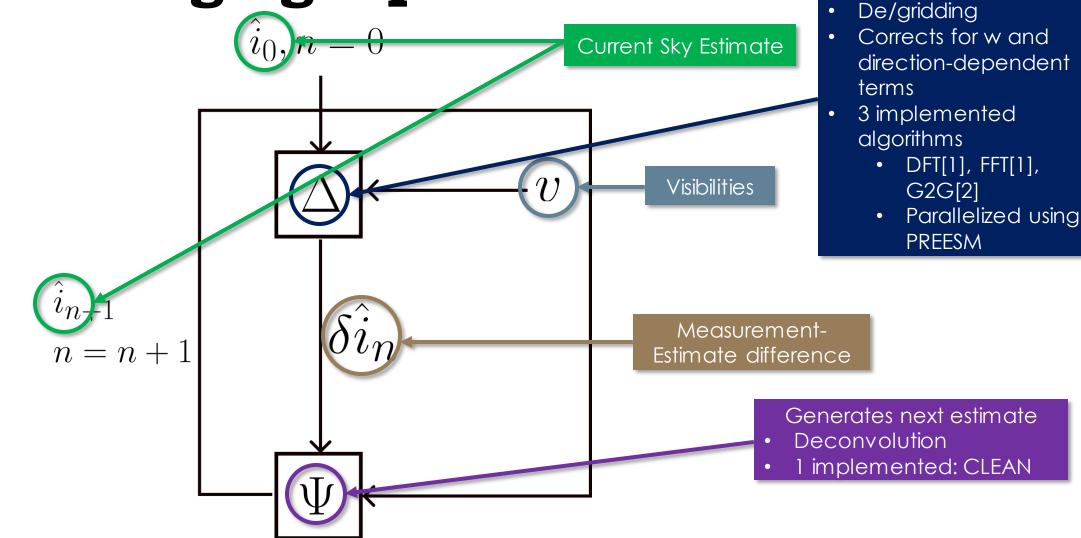
Aid in design of software and hardware architecture by simulating resource usage

Dataflow Pipeline Descriptions



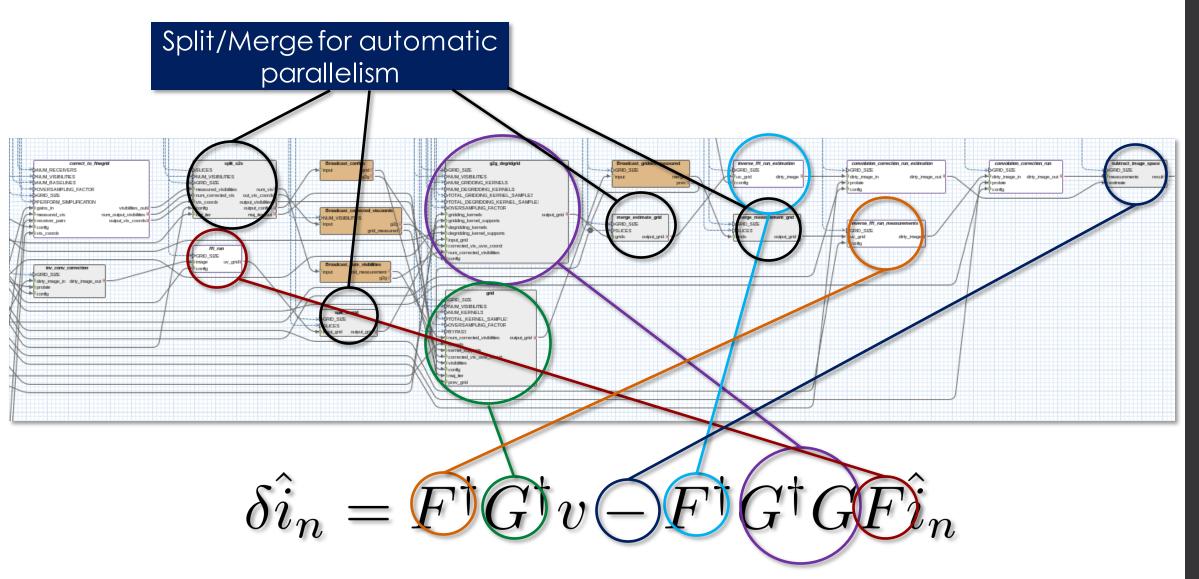
 \mathbf{Q}

The Radio-Interferometric Imaging Pipeline



9

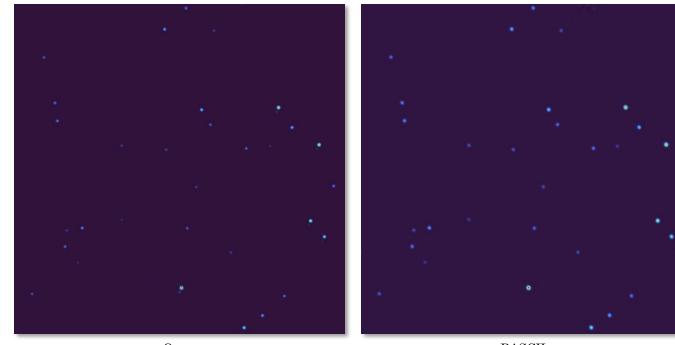
Concrete Example: Grid 2 Grid



Evaluation

Implement actors as C functions, generate entire pipeline code with PREESM.

Compare output against another imaging system (RASCIL[1])



Ours

RASCIL

Compared measured against estimated

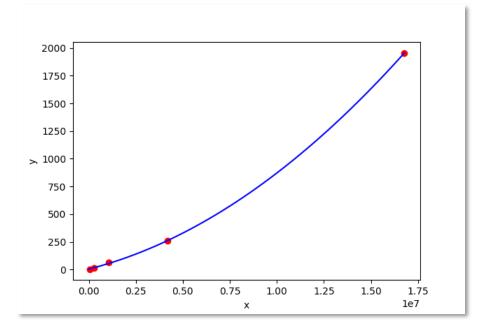
- Memory
- Computation Time

Estimating Computation Time

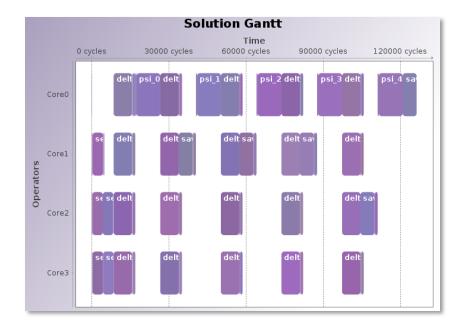
Ran benchmarks for each actor, either with our or with optimized code

• Varied number of visibilities, grid-size, and number of minor cycles

• Fitted polynomial of appropriate degree

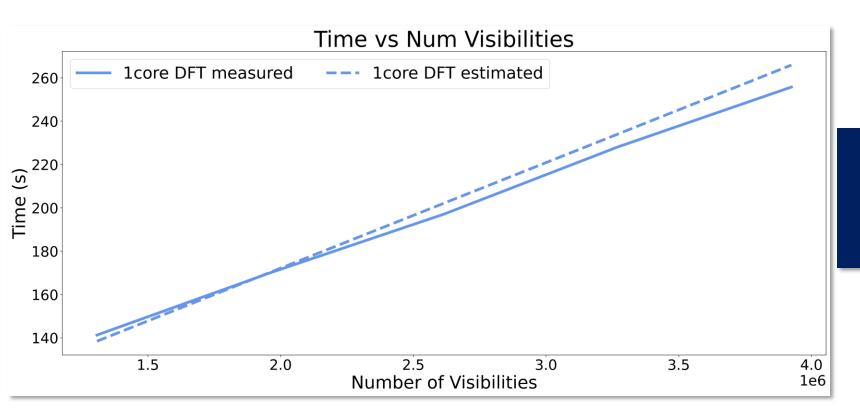


Estimated time obtained via PREESM gantt chart



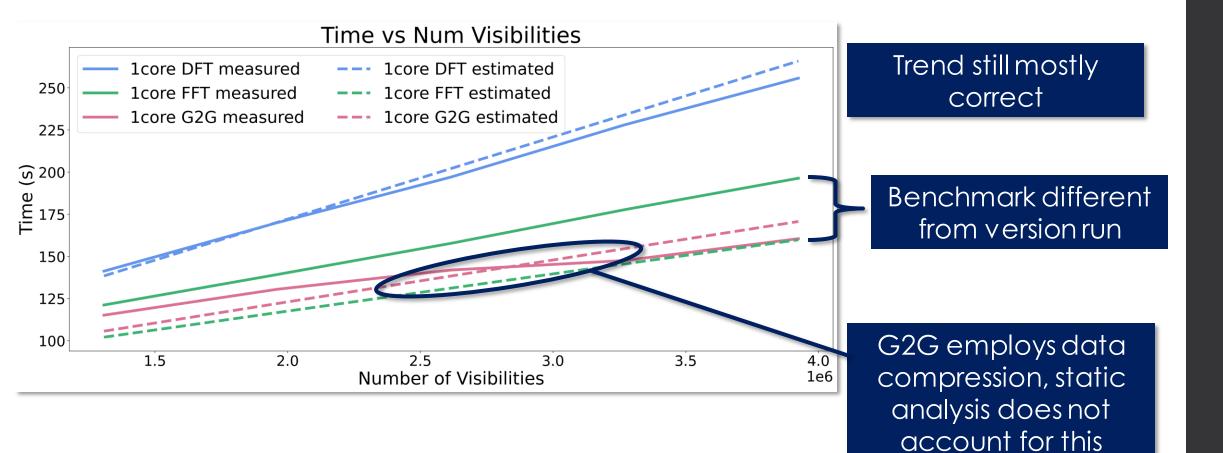
12

Estimated vs Measured Computation times



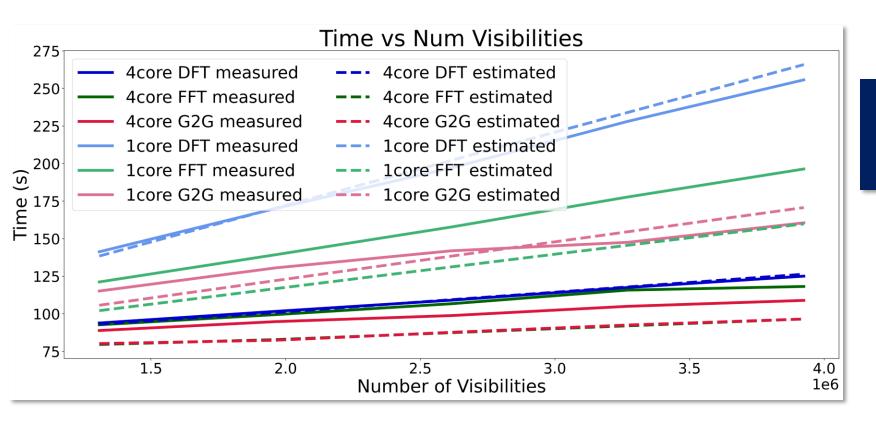
Estimation mostly correctly, difference primarily due to error in fitting model

Estimated vs Measured Computation times



14

Estimated vs Measured Computation times



Predicts correctly the 1.5-3x speedup when parallelizing

Estimated vs Measured Computation times

Framework predicts well:

- General algorithmic complexity
- Performance gain from parallelization

Estimation limitations:

- Benchmark different to measured
- Fitting error
- Only static analysis

Resource Estimation – Max Memory

Use PREESM's reported allocated inter-node memory as estimation

Starting allocation with RestFitAllocator(LARGEST_EIRST)

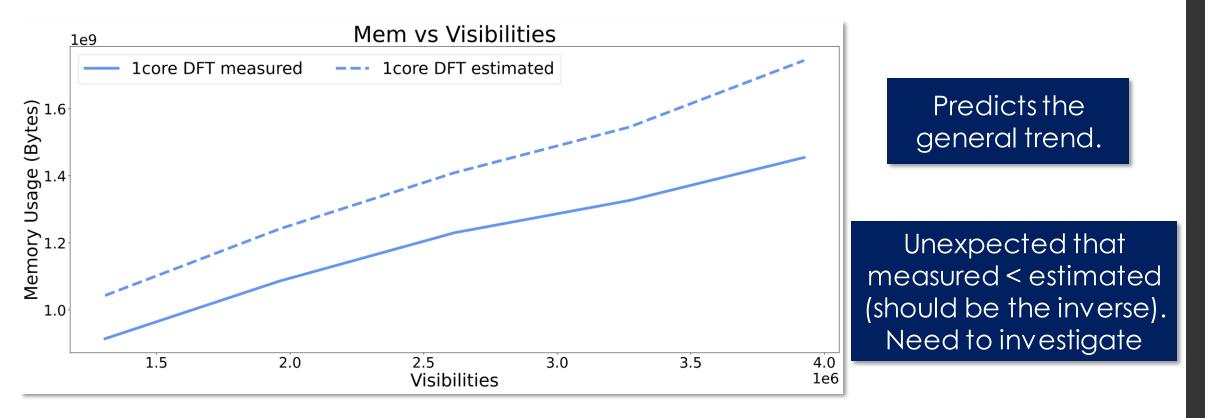
BestFitAllocator(LARGEST_FIRST) allocates 3.2940004020929337 GBytes in 171 ms.

workitow Step: code Generation (id org.ietr.preesm.codegen.xtend.task.codege

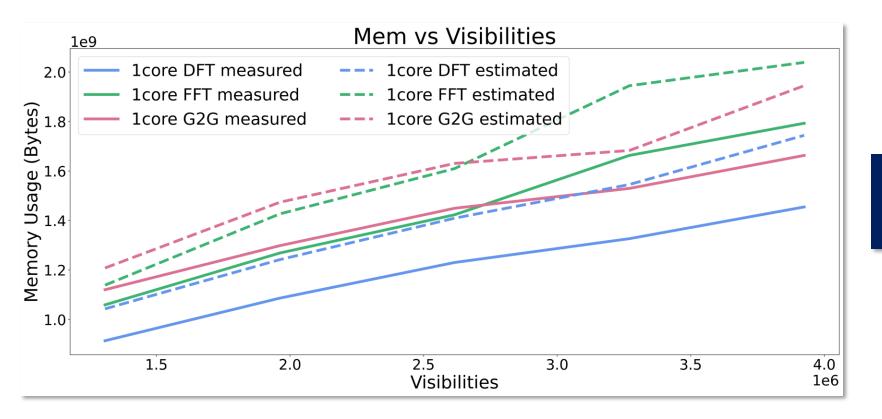
Use GNU time tool's reported maximum resident set size for measured

Average total size (kbytes). 0 Maximum resident set size (kbytes): 914240 Average resident set size (kbytes): 0 Major (requiring I/O) page faults: 56

Resource Estimation – Memory: Results

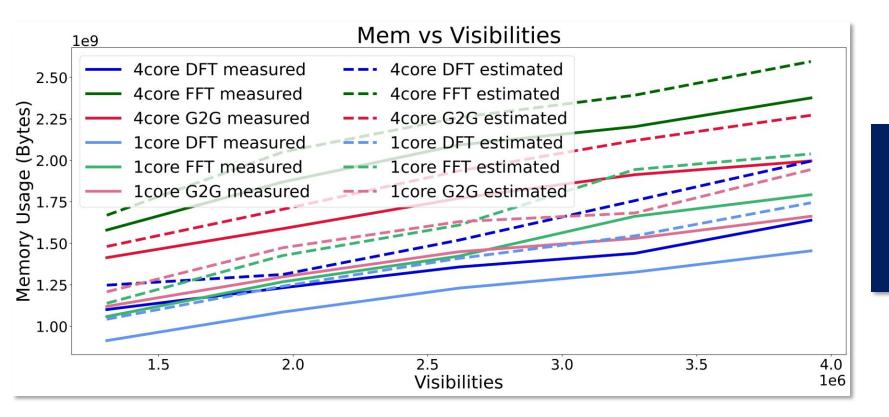


Resource Estimation – Memory: Results



Similar results for FFT and G2G pipelines

Resource Estimation – Memory: Results



Accurately predicts 1.1-1.5x increase in memory when increasing parallelization

Resource Estimation – Memory: Conclusions

Framework does a good job in predicting memory increase both as algorithm parameters, and parallelization increases

> Measured always a bit less than estimated. Should not be the case and need to test with valgrind which should give a more thorough profiling

Conclusions

Introduced an initial framework to prototype radio-interferometric algorithms

- Estimates computation time and memory
- Promising results, estimations similar trend to measured

Aspects to improve

- Improve models and benchmark data for better estimations
- Optimize implemented pipelines

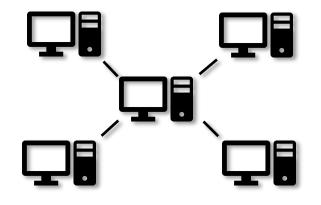
Main drawback is that only static analysis is supported currently

- Runtime/data dependent aspects e.g. compression can change runtime
- SPIDER can solve this

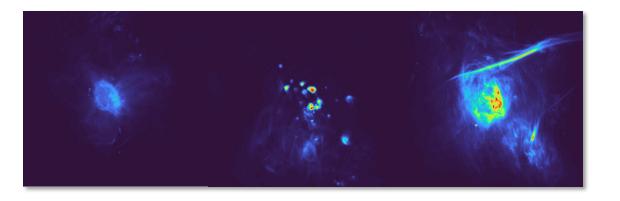


future work

Integrate support for different hardware architectures, evaluate on clusters, and in cases where we cannot obtain measurements (e.g. massive datasets)



Add support for more pipelines and datasets



Questions?