Instituto de Informática - UFRGS Programa de Pós Graduação em Computação

Performance Improvements
Applied in an Electromagnetic
Inversion Application Focused on
Homogeneous and Heterogeneous
Computational Environments

Jessica Imlau Dagostini Advisor: Lucas Mello Schnorr

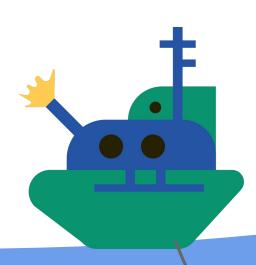








Introduction



Emitter

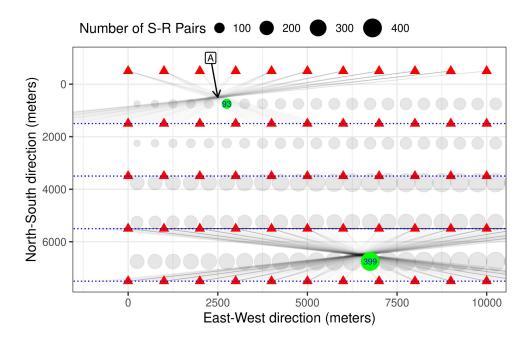
- Marine Controlled-Source Electromagnetic Method (mCSEM)
 - It uses receivers to collect underwater electromagnetic data
 - Electromagnetic waves traverse the subsurface under the seabed and reflect
- Oil and gas have known signatures

Introduction



- The data collected from receivers is not "understandable"
- Needs numerical methods to recover data
- Big quantity of data
- Inversion process is computationally expensive
- Run inversion in parallel
- High Performance Cluster

The mCSEM Inversion Application

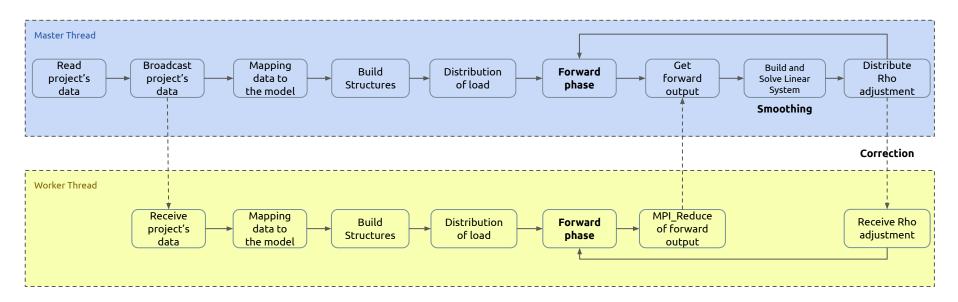


Survey data (red triangles are receivers, small blue squares are emitters), combined with the 2D grid of CMPs (circles).

- Application uses the Common Midpoint Technique
 - Map the collected data to the model
- It uses geographical midpoint between the receiver and the emitter
- Mapping will assign different quantities of Source-Receiver pairs to each CMP cell
- The CMP cell can or cannot match a receiver position

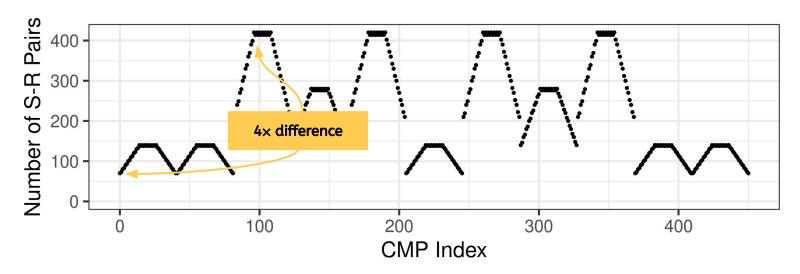
The mCSEM Inversion Application

- Based on the numerical inversion process
- 1D inversion using the CMP domain for 3D data
- Application parallelize two phases: the forward and smoothing



Finding the Imbalance

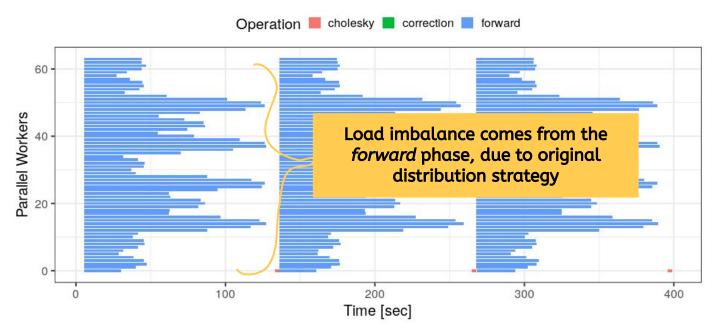
- First instrumentations results give us the organization of the cells
- Outputs with quantity of S-R pairs per CMP



Number of S-R pairs associated for each CMP of the studied model.

Finding the Imbalance

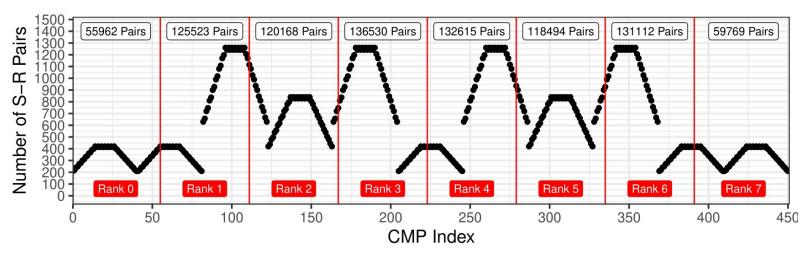
We ran the application with 64 MPI workers, disabling OpenMP threads



Execution behavior of the application executing only with MPI, presenting the timespace of the three main regions of the inversion algorithm.

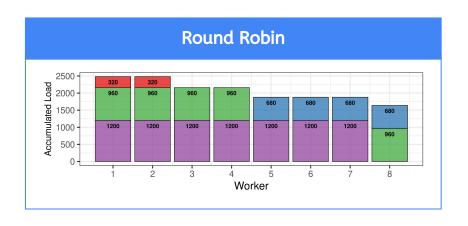
Finding the Imbalance

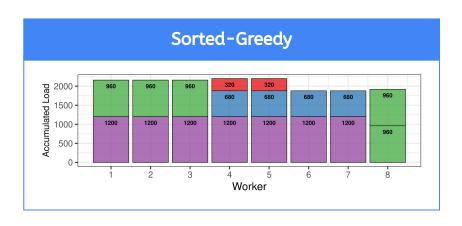
- Original distribution:
 - Considers the total quantity of CMP cells
 - Disregards the heterogeneous load of them

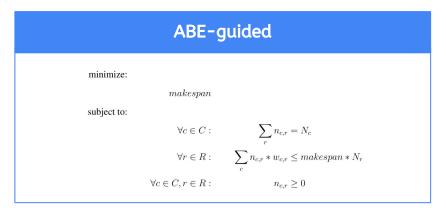


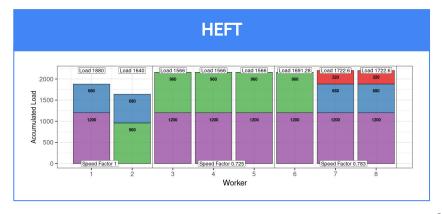
Behavior of the original distribution – it divide the cells among workers, disregarding its own heterogeneous weight.

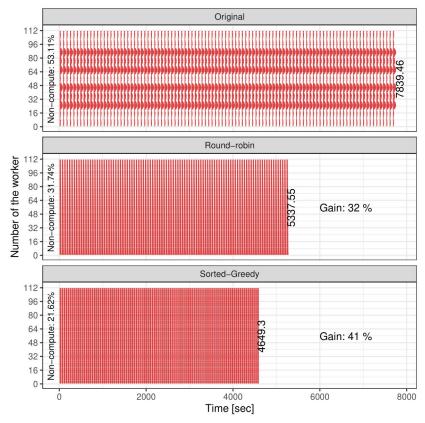
Load Balancing Heuristics











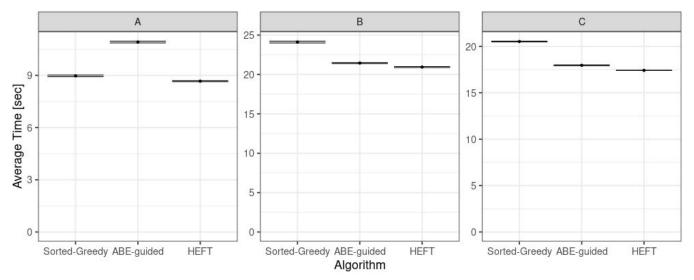
Space/time view of the forward step of 100 iterations for the Original, Round-robin, and Greedy allocation policies.

- Makespan
- 3 homogeneous heuristics
- 100 iterations
- Local cluster Draco
- 7 nodes, 16 cores per node
- Total of 112 MPI ranks
- Improvements up to 41%

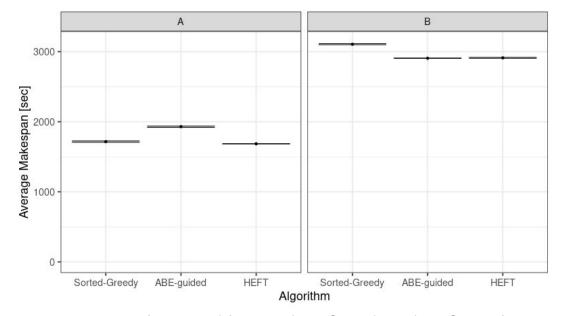
- Previous experiment motivates to extend solutions
- Load balancing using heterogeneous machines
- No changes on processor's frequency
- Two heuristics: ABE-guided and HEFT
- We selected three different machine combinations scenarios

| Study case | draco | hype | cei | blaise | Total Cores |
|------------|-------|------|-----|--------|-------------|
| Α | 7 | 5 | 2 | 1 | 304 |
| В | 3 | 2 | 1 | 0 | 112 |
| C | 0 | 1 | 2 | 1 | 112 |

- Compare iteration time in each scenario with each heuristic
- 10 repetitions with 10 iterations each



Average time of execution of one iteration in each scheduler running on different heterogeneous configurations.



Average makespan with execution of 100 iterations for each scheduler in two different heterogeneous scenarios.

- Makespan reduction with heterogeneous solutions
- 100 iterations
- Achieved a performance gain of 78% against the Original scheduler

Conclusion

- We investigate the computing performance of an MPI-based inversion application of mCSEM
- Identified issue regarding load balancing and solved it
- Performance improvement
 - 41% on homogeneous scenarios
 - 78% on heterogeneous scenarios
- Capacity planning tool to further predict best execution scenario
 - Available at bit.ly/interactive-capacity-plan

ACKNOWLEDGMENTS

This research was financed by Petrobras (2018/00263-5) and in part by the Coordenação de Aperfeiçoamento de Pessoal de Nível Superior - Brasil (CAPES) - Finance Code 001. Experiments executed in INF/UFRGS's PCAD, http://gppd-hpc.inf.ufrgs.br.









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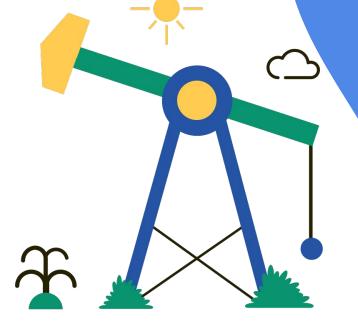
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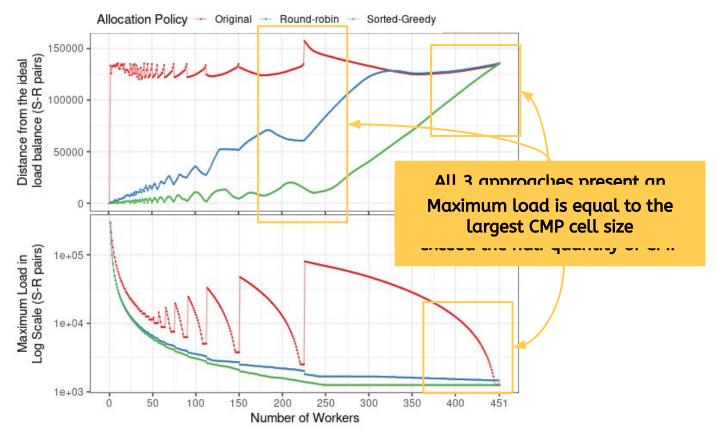
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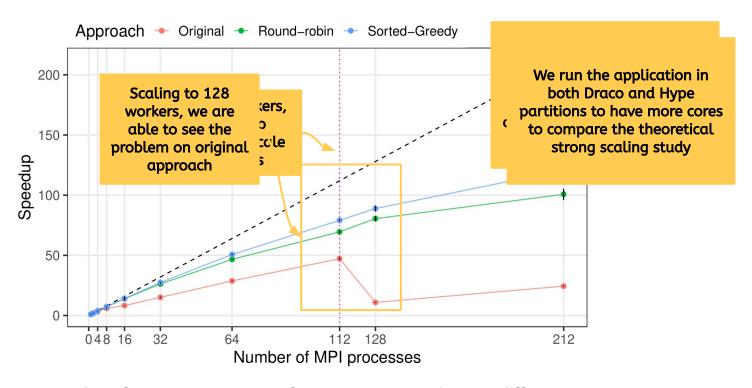








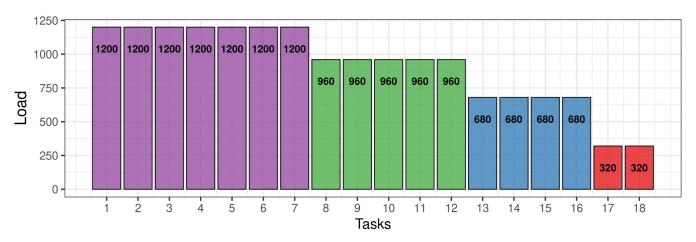




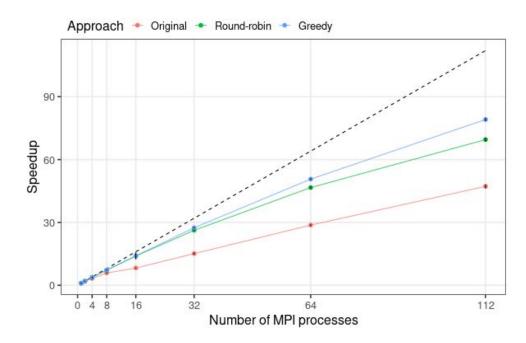
Speedup of many MPI processes from 1 up to 212 using two different partitions with the same processor's frequency.

Load Balancing Heuristics

- The characteristics of the load give us a number partitioning problem
- We will use the following synthetic load
- Searched for state-of-art heuristics to solve the problem



A synthetic load characterizing number partitioning.



Speedup of 1 up to 112 MPI processes. We run three iterations of the application. The baseline is Original with one process.

The dashed black line represents the ideal speedup.

- Speedup of the application
- We compared the two heuristics against the Original using 1 up to 112 MPI processes
- Compute the mean speedup with following equations

$$\begin{split} \mathbf{E}(\text{speedup}) &= \mathbf{E}(\text{Original 1 proc}) \times \mathbf{E}(\frac{1}{\text{new_distribution } n \text{ proc}}); \text{ and} \\ \mathbf{Var}(\text{speedup}) &= \mathbf{E}(\text{Original 1 proc}^2) \times \mathbf{E}(\frac{1}{\text{new_distribution } n \text{ proc}^2}) \\ &- \mathbf{E^2}(\text{original 1 proc}) \times \mathbf{E^2}(\frac{1}{\text{new_distribution } n \text{ proc}}). \end{split}$$