

Lenstool-HPC

A high performance computing approach to lens-modelling

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Collaboration between LASTRO and SCITAS

Lenstool

- ◆ Parametric mass-modelling of gravitational lenses using Bayesian MCMC sampler
- ◆ Multiply imaged sources serve as constraints:

$$\chi_i^2 = \sum_{j=1}^{n_i} [x_{obs}^j - x_{pred}^j(\theta)]^2 / \sigma_{ij}^2$$

- ◆ Parameter space is huge
- ◆ Process can take several weeks.

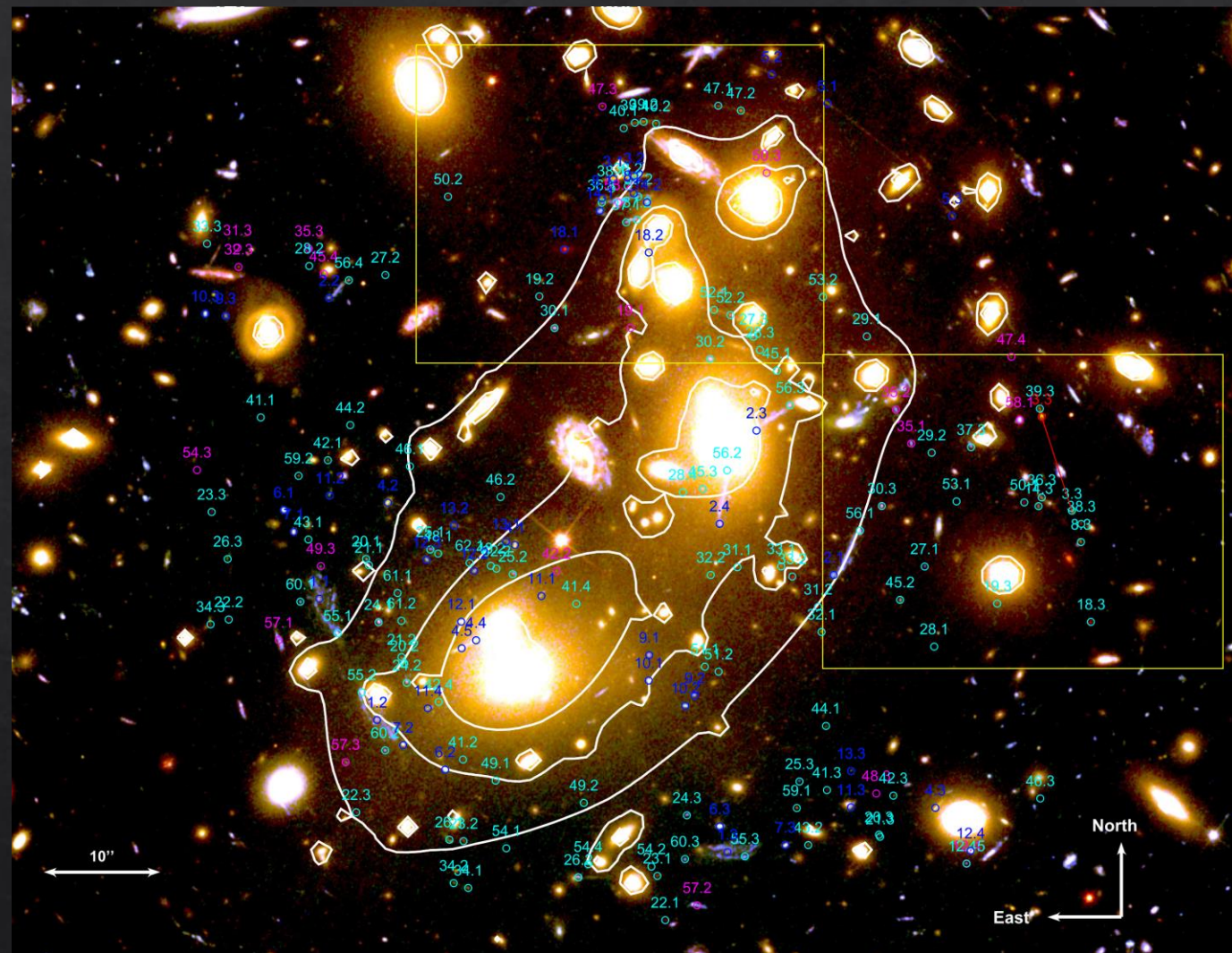


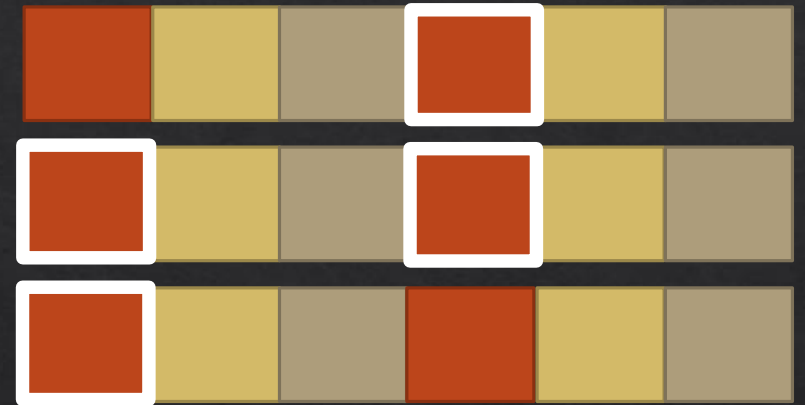
Fig. 1

Credit for Image: Jauzac et al. 2015b

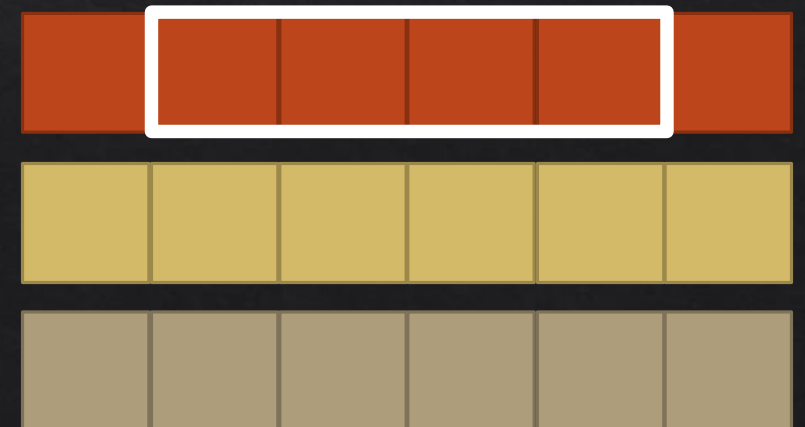
Lenstool-HPC

- ◆ Goal: Speed-up Lenstool by re-coding Lenstools mass-modelling capability using HPC techniques for HPC clusters.
- ◆ Challenge: Gradient computation of the deflection potential for each mass-model.
- ◆ A typical case of SIMD (Single Instruction, multiple data) operations.
- ◆ SOA layout VS AOS layout:
 - ◆ SOA components are stored contiguously in the memory, so only 1 memory access is needed to feed the CPU.
 - ◆ With AOS, memory jumps are needed, since CPUs do not load all the memory at once, and you need to "pack" your data in a vector.

Array of structure (AOS)



Structure of array (SOA)



Data Level Parallelism

- ◆ Modern processors are able to process multiple SIMD instructions simultaneously (DLP)
- ◆ Lenstool-HPCs SOA implementation shows a significant speed-up (1.4 at worst, 2 at best) already only due to easier memory access
- ◆ When allowing DLP, on AVX2 capable machines (vector length of 8) we get a speed-up of 10

