

*Julian Merten*

*Joint cluster reconstructions  
on GPUs*

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Zentrum für Astronomie  
Universität Heidelberg  
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with: Massimo Meneghetti (OABO)  
& Matthias Bartelmann (ITA)



# Great things to come: The CLASH

One of three HST/MCT programmes. Start September 2010 (3 cycles).



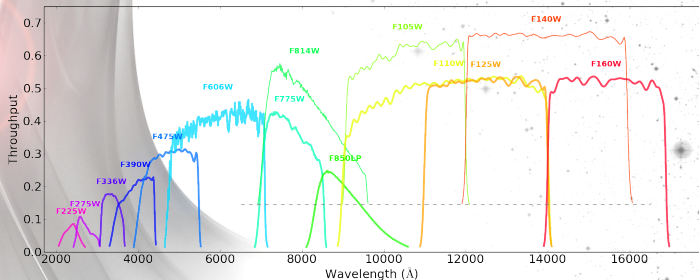
**The CLASH**

Cluster Lensing And Supernova survey with Hubble  
A Hubble Space Telescope Multi-Cycle Treasury Program  
P.I. Marc Postman (STScI) Co-P.I. Holland Ford (JHU)

Matthias Bartelmann • Narciso Benitez • Larry Bradley • Tom Broadhurst • Dan Coe • Megan Donahue • Rosa Gonzalez-Delgado  
Leopoldo Infante • Daniel Kelson • Ofer Lahav • Doron Lemze • Dan Maoz • Elisor Medezinski • Leonidas Moustakas • Eriko Nagao  
Adam Riess • Piero Rosati • Stella Seitz • Keichi Umetsu • Arjen van der Wel • Wei Zheng • Adi Zitrin

## Science Drivers

- To map the dark matter in galaxy clusters
- To detect SN out to redshifts  $z > 1.5$
- To detect and characterise  $z > 7$  galaxies
- To study the galaxies in and behind the clusters



## *CLASH Facts*

- 25 X-ray clusters
- 524 orbits
- ACS + WFC3 obs.
- 14 (16) wave bands
- wide follow-ups with SUBARU

## *Our contribution*

- Full pipeline calibration
- nonparametric DM profiles
- weak-lensing shapes + flexion
- magnification maps for the high-z guys (Maruša's talk)

# The basic idea of an “inverse” method (Bartelmann96)

## Cluster lensing in a box

$$\beta = \theta - \alpha(\theta)$$

$$\partial = \partial_1 + i\partial_2 \quad \partial^* = \partial_1 - i\partial_2$$

$$\psi \quad \alpha = \partial\psi$$

$$2\gamma = \partial\partial\psi \quad 2\kappa = \partial^*\partial\psi$$

$$2F = \partial^*\partial\partial\psi \quad 2G = \partial\partial\partial\psi$$

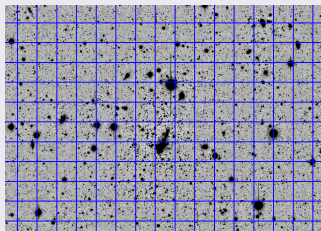
## Statistical approach

$$\chi^2(\psi) = \chi_1^2 + \chi_2^2 + \chi_3^2 + \dots$$

Possible constraints:

- Ellipticities of background sources
- Flexion (JM10 in prep.)
- Multiple image systems (Bradač05+)
- Critical curve estimates (JM09+)

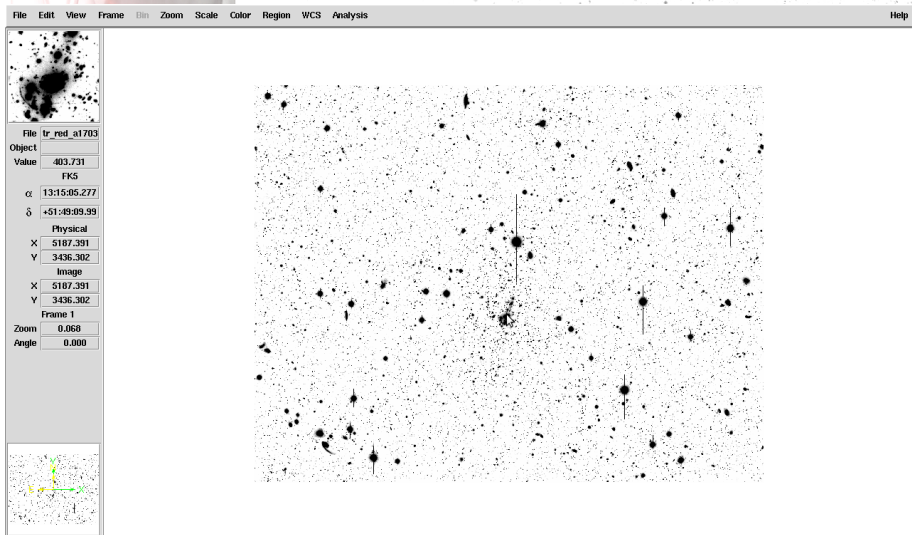
## The trick



$$\frac{\partial \chi^2(\psi_k)}{\partial \psi_l} \stackrel{!}{=} 0$$
$$\Rightarrow \mathcal{B}_{lk} \psi_k = \mathcal{V}_l$$

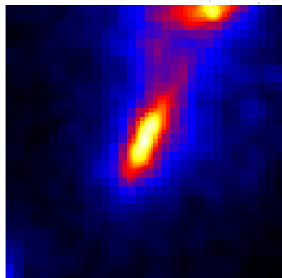
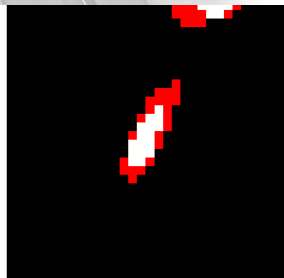
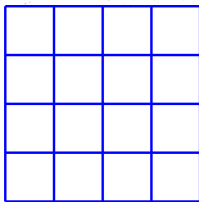
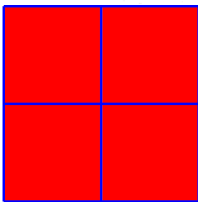


# A problem of different scales (JM10, Bradač09)



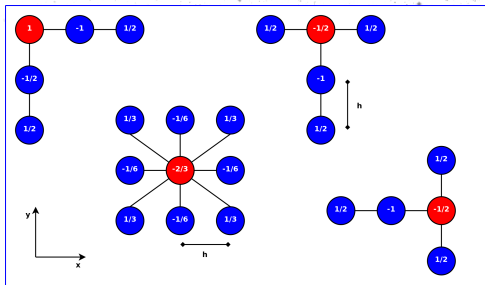
(Abell 1703 in SUBARU r-band)

*A problem of different scales (JM10, Bradač09)*



## Making it all work: Numerics (JM09)

- $\alpha$ ,  $\gamma$ ,  $\kappa$ ,  $F$  and  $G$  can be expressed by derivatives of  $\psi$  via finite differences.
- Linearisation of the problem.
- 2-level iteration scheme with simple regularisation (Bradač05).



### Problem Runtime

$$\mathcal{B}_{lk}\psi_k = \mathcal{V}_l$$

$$\mathcal{B}_{lk} \sim a_i b_j C_{ij} D_{il} E_{jk}$$

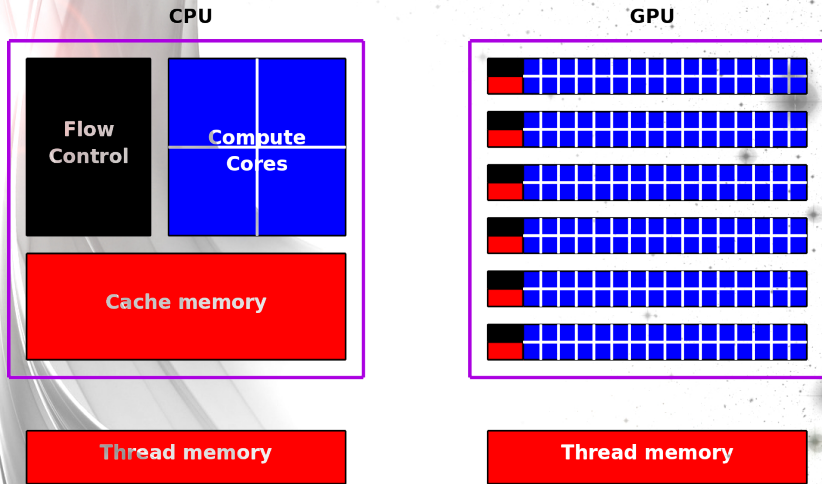
$$\mathcal{V}_l \sim a_i b_j C_{ij} E_{il}$$

$$l, k, i, j \sim \mathcal{O}(\text{grid\_dim}^2)$$

### Geeky implementation facts

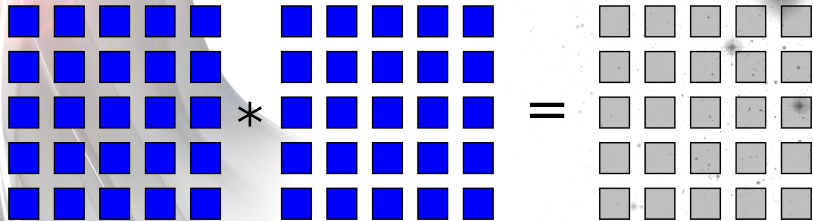
- Parallel C++ code
- medium sized  $\sim 12000$  lines
- Uses GSL, LAPACK, ATLAS, MPI
- Fully documented, including user manual
- and...CUDA...

# Data-parallel single node GPU Parallelisation



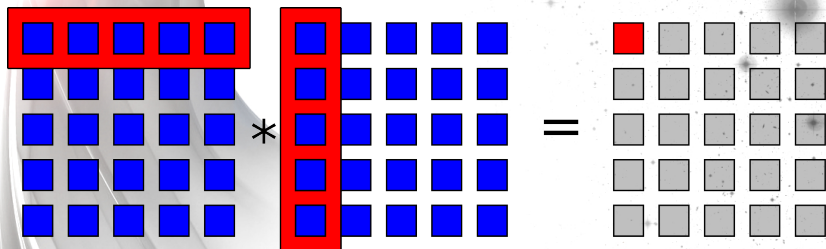
- One single GPU allows for massive parallelisation at a fraction of the cost of a CPU cluster, if problem is suited for ⇒ **Data-parallel**.

## Data-parallel single node GPU Parallelisation



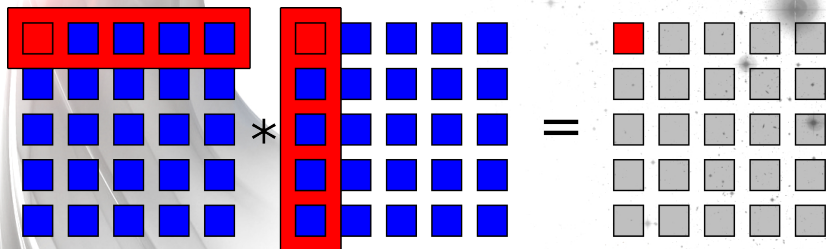
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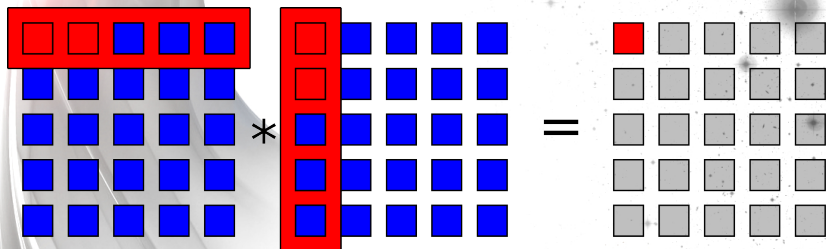
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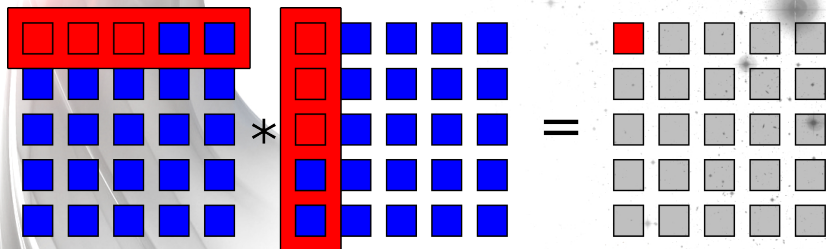
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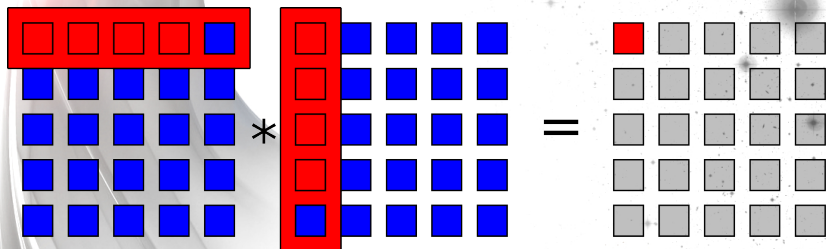


## Data-parallel single node GPU Parallelisation



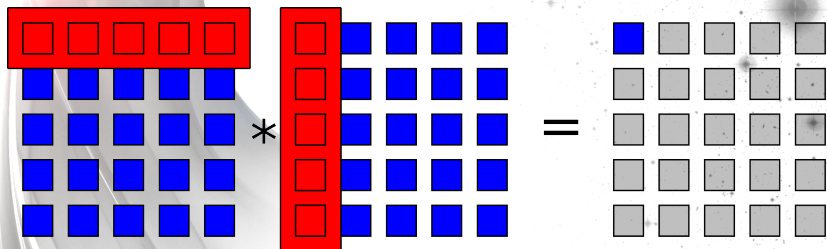
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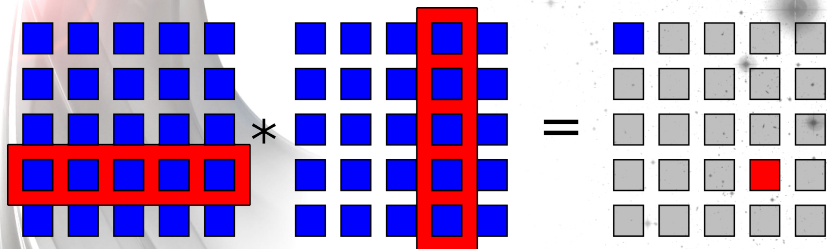
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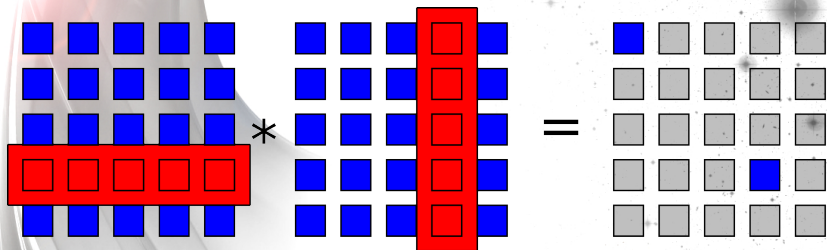
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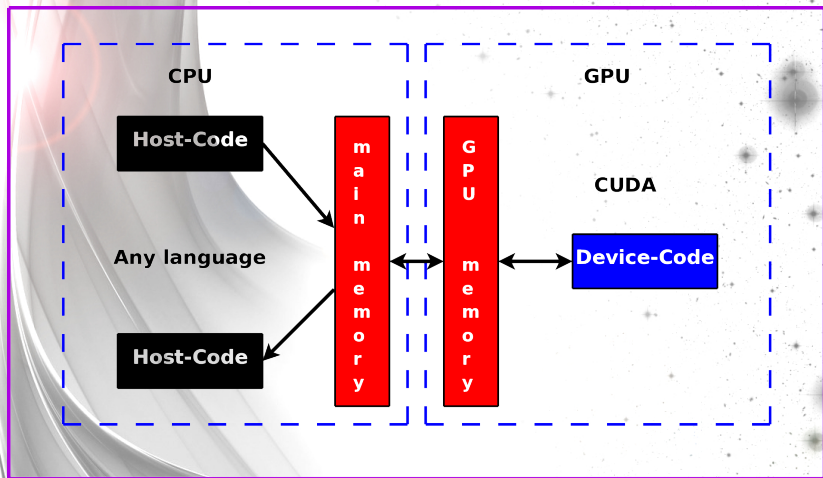
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## GPUs in practice (JM10 in prep.)

### NVIDIA Tesla C1060

- 240 streaming cores
- 4 GB DDR3 GPU memory
- 933 GFLOPS peak performance
- Fermi cards out by now, update planned



### Speed-up

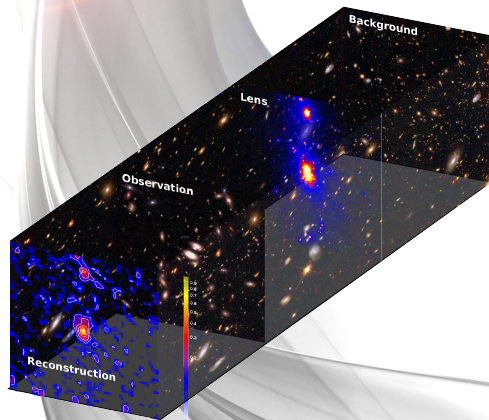
- Calculate:

$$B_{lk} = a_i b_j c_{ij} d_{ik} e_{jl}$$

- one-core CPU: 82.3 s
- 240 core GPU: 1.03 s

# Realistic lensing simulations: *SkyLens* (Meneghetti, JM 08/10)

Developers: Massimo Meneghetti, Peter Melchior, Fabio Bellagamba, JM

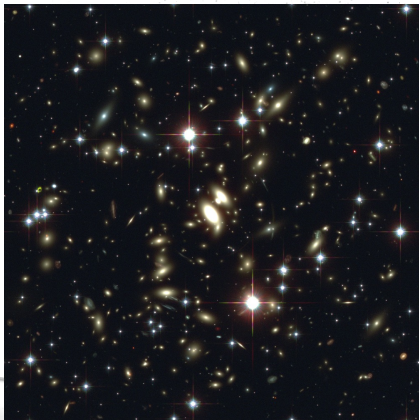


Name	Description
$D$	aperture diameter
$g$	detector gain
$A_{\text{pix}}$	pixel area
$F(\lambda)$	used filter
$M(\lambda)$	mirror filter curve
$O(\lambda)$	optics filter curve
$C(\lambda)$	CCD filter curve
FoV	total field-of-view
RON	detector readout-noise
$f$	flat-field accuracy
$a$	residual flat-field error
PSF	PSF model
$t_{\text{exp}}$	exposure time
$A(\lambda)$	atmospheric extinction
$m_a$	airmass
$\text{SED}_{\text{sky}}$	sky-background emission
$\text{SED}_{\text{gal}}$	background population
$\alpha$	deflection angle map



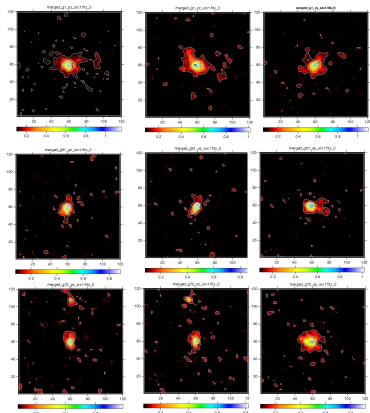
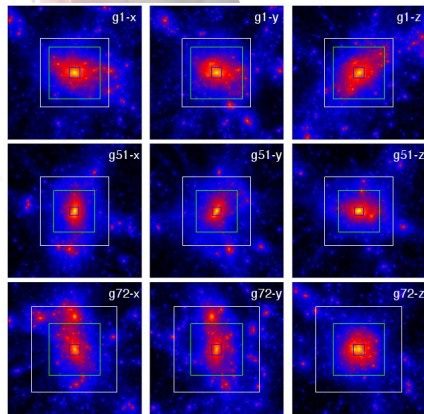
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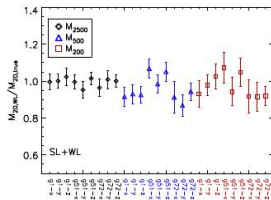
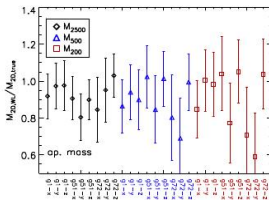
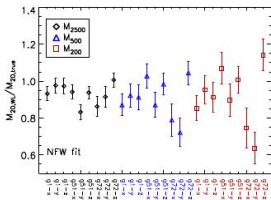
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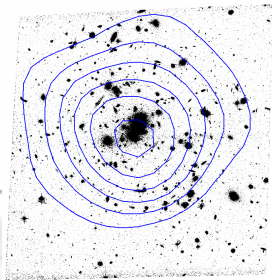
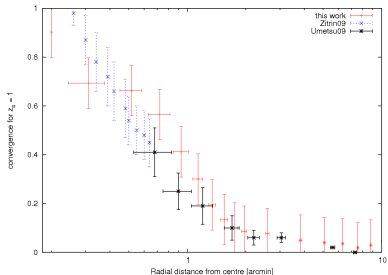
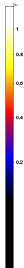
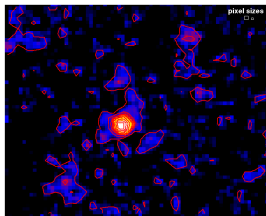
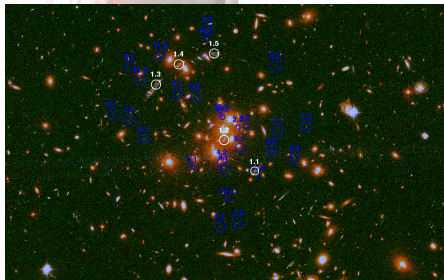


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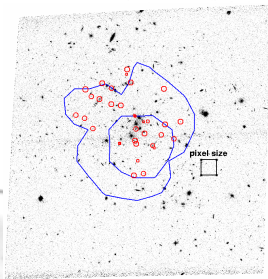
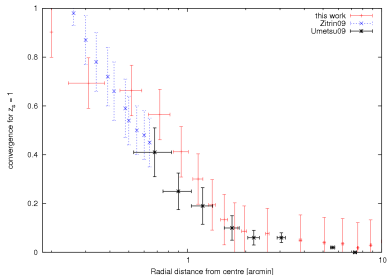
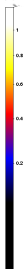
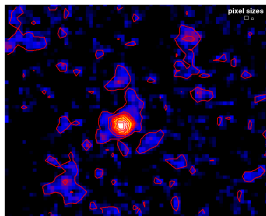
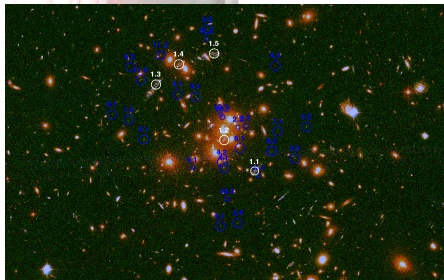
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# CL0024 (with: T. Broadhurst, A. Zitrin, K. Umetsu)



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# Conclusions

- 1 Looking forward to see the CLASH:
  - ▶ Cluster DM profiles
  - ▶ Strong lensing peculiarities
  - ▶ High-z universe
- 2 Joint methods, combining several lensing constraints are able to reliably reconstruct clusters in a nonparametric way.
- 3 GPUs are able to outperform “classical” compute clusters at a fraction of the cost, if your code is suited for such an implementation.