Perspectives of GPU computing in Science, Rome 2016



ASTRI data reduction software on low-power and parallel architectures

Michele Mastropietro, S. Lombardi, D. Bastieri, L. A. Antonelli INAF - Osservatorio Astronomico di Roma & UniPD

for the ASTRI Collaboration and the CTA Consortium



- Imaging Atmospheric Cherenkov Telescopes (IACTs)
 - the Cherenkov Telescope Array (CTA)
 - the ASTRI project in the framework of CTA
- The ASTRI data reconstruction and scientific analysis software
- Algorithms parallelization
- GPU and low-power platform benchmarks using different metrics

Top of the atmosphere

- Very High Energy primary cosmic rays (p, α , e[±], γ , ...) (CTA range: 20 GeV - 300 TeV)
- Extended Air Showers (EAS)

- Hadronic showers (>99%)
- **Electromagnetic showers**
- Cherenkov radiation emission
 - optical and near-UV light
 - Flash duration of few ns
- Light collected by the mirrors
- Light focalized in the cameras
- EAS images in the focal planes
- Image calibration, cleaning, and parameterization
- Stereoscopic reconstruction and background suppression
- Scientific analysis
- Science products
- Observatory products



IACT

light pool



- Next generation ground based Gamma-ray Observatory
- Open observatory
- Two sites with total > 100 telescopes (LSTs+MSTs+SSTs)
 - Southern Site: Near Paranal in Chile (selected for negotiations)
 - Northern Site: La Palma, Canary Islands (selected for negotiations)
- 32 countries, ~400M€ project





The **ASTRI Project** (led by INAF) has two main goals:

- an end-to-end prototype of the CTA small-size telescope in a dual mirror configuration (ASTRI SST-2M), inaugurated on 2014 Sept. 24th at the INAF observing station on Mt. Etna (Sicily) and undergoing the scientific and performance validation phase by the end of 2016;
- an **ASTRI mini-array** composed of **9 ASTRI telescopes** proposed to be installed at the chosen CTA Southern site likely in 2018





The ASTRI Mini-array



- Proposed to be installed at the southern site of the CTA, as an initial seed of the entire observatory
- In such a location power consumption, data bandwidth, heat dissipation become critical concerns



The ASTRI Mini-array



- The capability of each detector to process, at least partially, its own data before sending them to the central data acquisition could provide a key advantage
- Carrying out preliminary data reduction on the telescope would greatly decrease the bandwidth required by the array installation



The ASTRI Scientific Software

- Two main goals:
 - Having something working and fully functional ready to analyze real data from ASTRI camera in fall 2016
 - Having the low level part of the analysis running on low-power architectures (i.e. ARM+GPUs)

A-SciSoft is organized in four distinct functional breakdown stages:

- Calibration
- Reconstruction
 - Telescope-wise
 - Array-wise
- Analysis
- Science





ASTRI Scientific Software implementation



Breakdown stages; Basic components; Executable modules; I/O Data level.

The ASTRI SST-2M Prototype and Mini Array Data Pipeline

- Manages FITS data (from DL0 to DL3) adopting CFITSIO/CCFITS libraries;
- Operates on chuncks of events
- It is written in C++, CUDA, Python
- It is developed in independent software modules linked by pipelines written in Python



Pixel-level algorithms easily express parallelism

Calibration

Essentially an *embarrassingly parallel* Fused Multiply-Add (FMA) operation:

 $0 = 0 \times 1 + 2$

Cleaning

Two pass cleaning (two thresholds comparison) Well suited to parallelism







Calibration

Calibration formula (for each pixel):

PHE = ADC * CC + PED

Where:

PHE: photo-electron equivalentADC: ADC counts from camera pixelCC: calibration coefficientPED: pedestal value

Extraction of CAL1 from CAL0

- Application of CAL1 to EVT0/MC0
- NSB level extraction
- Data Quality Checks
- ✤ Production of EVT1a/MC1a

(TEL.-WISE CALIBRATED DATA)









Cleaning

- Two pass cleaning (two thresholds comparison)
 - 1. Identify seeds of core pixels
 - 2. Find neighboring pixels constituting the image boundary
- Well suited to parallelism
 - parallelize both per pixel and per event





- Image Cleaning and Parameterization
- Telescope Pointing Reconstruction
- Production of EVT1(b)/MC1(b)



Numerical Precision Validation

• FMA instruction matters! (Available since Haswell)

round(X * *Y* + *Z*)

VS.

round(round(X * Y) + Z)

FMA makes computations faster and more accurate!*

Calibration: phe = cc*adc+ped double precision math

In [3]: 0.0444444455*971-43.1333351
Out[3]: 0.022221480500000723
In [4]: 0.0439560451*968-42.6923103
Out[4]: -0.1428586432000003

In [5]: 0.0439560451*970-42.6483536
Out[5]: -0.010989852999998106



*Whitehead, N.; Fit-Florea, A. - NVIDIA (2011). "Precision & Performance: Floating Point and IEEE 754 Compliance for NVIDIA GPUs".

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Seamless and flexible integration of GPU code

- Build phase does not depend on CUDA toolkit
- Software modules detect GPUs in the system and act accordingly (*thanks C++ polymorphism!*)
- CPU/GPU execution switchable on user request
- Important to enable GPU *Persistence daemon* to minimize driver load latency



Reference Test Case



- 55000 (≈ 500 MB) events of simulated "real data"
- \approx 110s of nominal acquisition rate (500Hz)
- ≈ 80.5% of events survives cleaning with default settings
- Compliant with format and size agreed with camera hardware team



Development system

Workstation

Dual-processor Intel Sandy Bridge @ 2GHz with 16 physical cores and 128GB of RAM (8GB per core)

GPU gen3 READY and n.1 installed (up to 3 GPU drives)

8 disk slots of 4TB each (to export 2 different redundant drives of ~12TB each)

direct link and share with the storage system



- Installed @ OAR
 Monte Porzio Catone
- Accelerator:
 One NVIDIA Tesla
 K20c (20-30% slower than K40)





Unified module (up to DL1b)



- Direct processing from DL0 to DL1b
- 73x reduction in size:

500 MB input 6.9 MB output

 CPU parallel processing with 32 OpenMP threads



Low-power vs Speed



BUT

CTA requirement:

• The observatory should be able to send scientific alert within **30s** after the acquisition of the last event participating to the alert.



Low-power vs Speed



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cta cherenkov telescope array

NVIDIA Jetson TK1

- Heterogeneous System-on-Chip
- CPU: Quad-core ARM A15
- GPU: Kepler architecture 1 Multiprocessor
- RAM: 2GB (memory shared by CPU & GPU)
- OS: Ubuntu 14.04 Linux for Tegra (L4T)
- CUDA 6.5
- I/O: SATA 3Gb/s HDD





Average power consumption: < 10 W



Low-power Unified module



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cta



Reconstruction on Jetson TK1



astrireco

- Implements random forest application
- Loads pre-trained models (look up tables LUTs)
- Energy, direction and hadronness reconstruction

Execution time: 10s - 4 ARM core (using OpenMP)



{Time,Energy} to solution

- Our computing node: 225W
- Jetson: 10 W peak
- Rough (conservative) estimation:

	Workstation	Jetson	Improvement
Time to solution	6,5 s	22,5 s	0,28 x
Energy to solution	1430 J	225 J	6,3 x
Energy delay product	9295 Js	5062 Js	
Event/s	7690 evt/s	2200 evt/s	
Energy/event	29,6 mJ/evt	4,5 mJ /evt	



Challenges

Soon tests on real ASTRI SST-2M prototype data
 @ Mt. Etna (fall 2016)

- Finish porting of clustering on GPU
 - Less time on PCIe transfers
- Try the new Jetson hardware: NVIDIA Jetson TX1
 - More effective usage of Sistem-On-Chip memory
 - Improvements in time foreseen within the same power consumption (10 W)