



Future options for SKA-scale astronomical computing

Attila Popping

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Second ngVLA Technical Workshop December 8-9 2015 Socorro, New Mexico, USA









Data is the new challenge



x100



Data is the new challenge





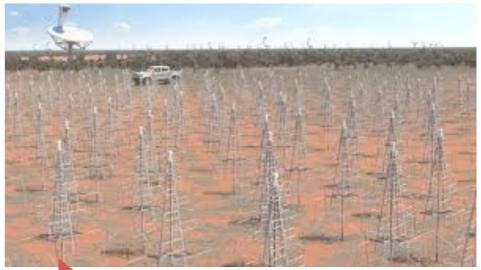


x10 VLASS

x10 ngVLA

x10





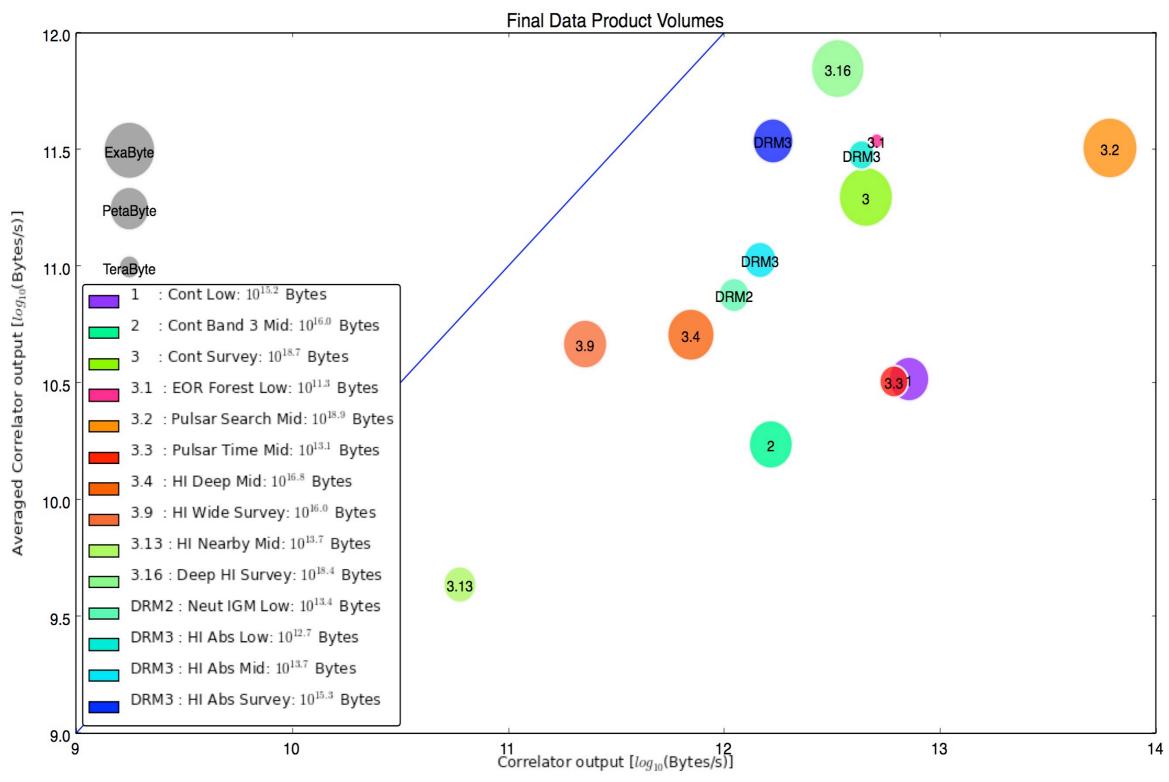
WALLABY/DINGO/ASKAP

SKA

x100



SKA Science Projects





Future Science

- Small teams or individuals are replaced by large international teams
- Requirements are outgrowing the capabilities of a single institute.
- The capability of storing and processing data will become equally important as the observations or the telescope.
- Manual data reduction has to be replaced by automated workflows



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- It will take time to develop



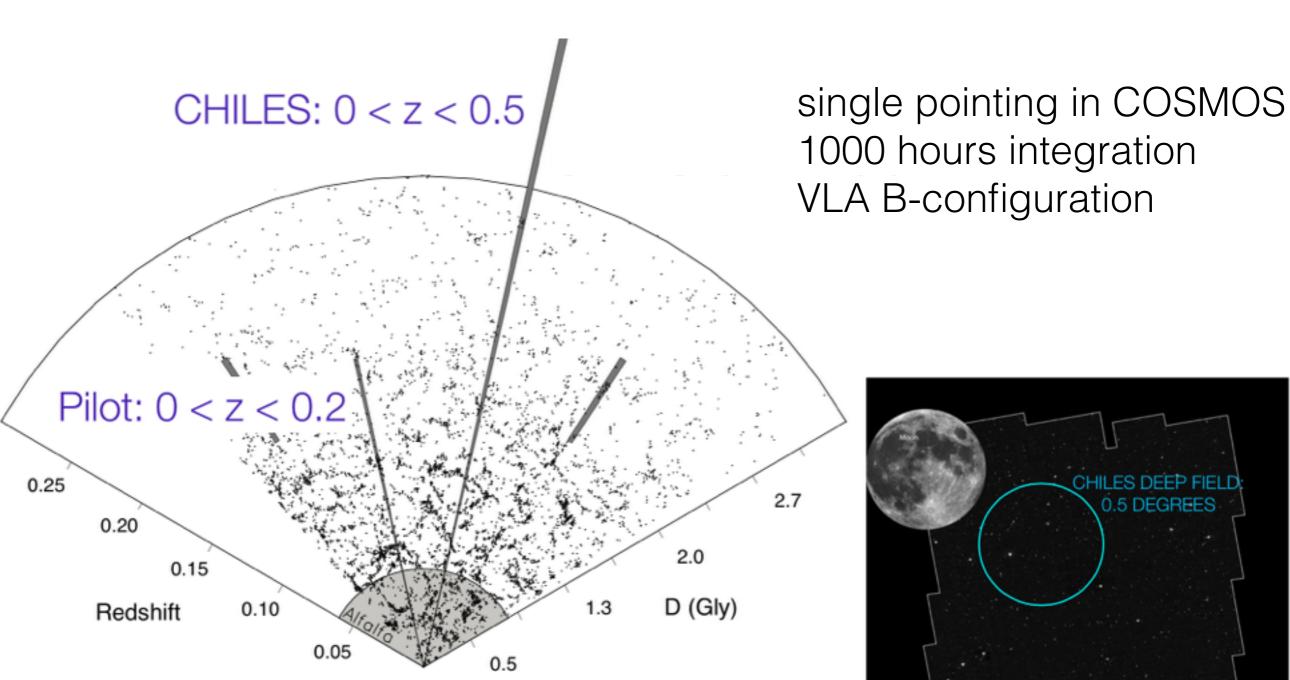
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CHILES is the perfect test particle!

CHILES

COSMOS HI Large Extragalactic Survey





Upgraded VLA

	OLD	PILOT	NEW
Bandwidth (MHz)	6.25	240	480
Channels	31	16384	30720
Velocity resolution (km/s)	40	3.5	3.5
Instantaneous z coverage	0 <z<0.004< td=""><td>0<z<0.193< td=""><td>0<z<0.5< td=""></z<0.5<></td></z<0.193<></td></z<0.004<>	0 <z<0.193< td=""><td>0<z<0.5< td=""></z<0.5<></td></z<0.193<>	0 <z<0.5< td=""></z<0.5<>





178 hours observed and reduced in Fall 2013

~200 hours observed in Spring 2015

~230 hours allocated for Spring 2016



A case study

Imaging SKA-Scale data in three different computing environments

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Abstract

We present the results of our investigations into options for the computing platform for the imaging pipeline in the CHILES project, an ultra-deep HI pathfinder for the era of the Square Kilometre Array. CHILES pushes the current computing infrastructure to its limits and understanding how to deliver the images from this project is clarifying the Science Data Processing requirements for the SKA. We have tested three platforms: a moderately sized cluster, a massive High Performance Computing (HPC) system, and the Amazon Web Services (AWS) cloud computing platform. We have used well-established tools for data reduction and performance measurement to investigate the behaviour of these platforms for the complicated access patterns of real-life Radio Astronomy data reduction. All of these platforms have strengths and weaknesses and the system tools allow us to identify and evaluate them in a quantitative manner. With the insights from these tests we are able to complete the imaging pipeline processing on both the HPC platform and also on the cloud computing platform, which paves the way for meeting big data challenges in the era of SKA in the field of Radio Astronomy. We discuss the implications that all similar projects will have to consider, in both performance and costs, to make recommendations for the planning of Radio Astronomy imaging workflows.

Keywords: methods: data analysis, Parallel Architectures: Multicore architectures, Distributed architectures: Cloud computing, CHILES

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Dodson et al. 2015

Imaging SKA-Scale data in three different computing environments

arXiv: 1511.0040

- 1) Local Cluster (Pleiades)
- 2) Super Computer (Magnus)
- 3) Cloud computing (AWS)

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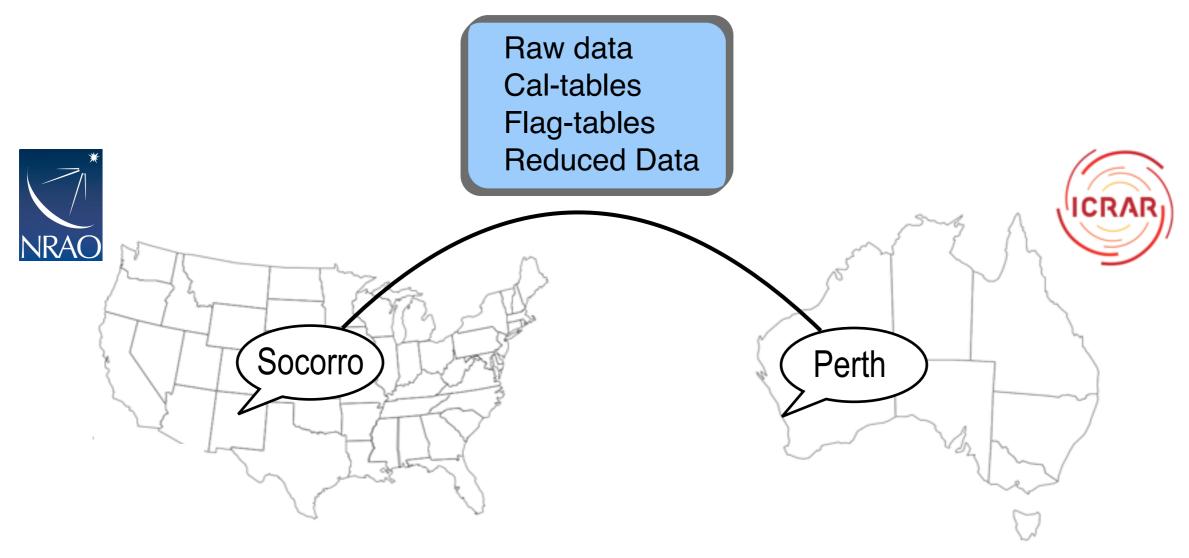
CHILES workflow



Observing
Calibration
Flagging
Quality Control



CHILES workflow

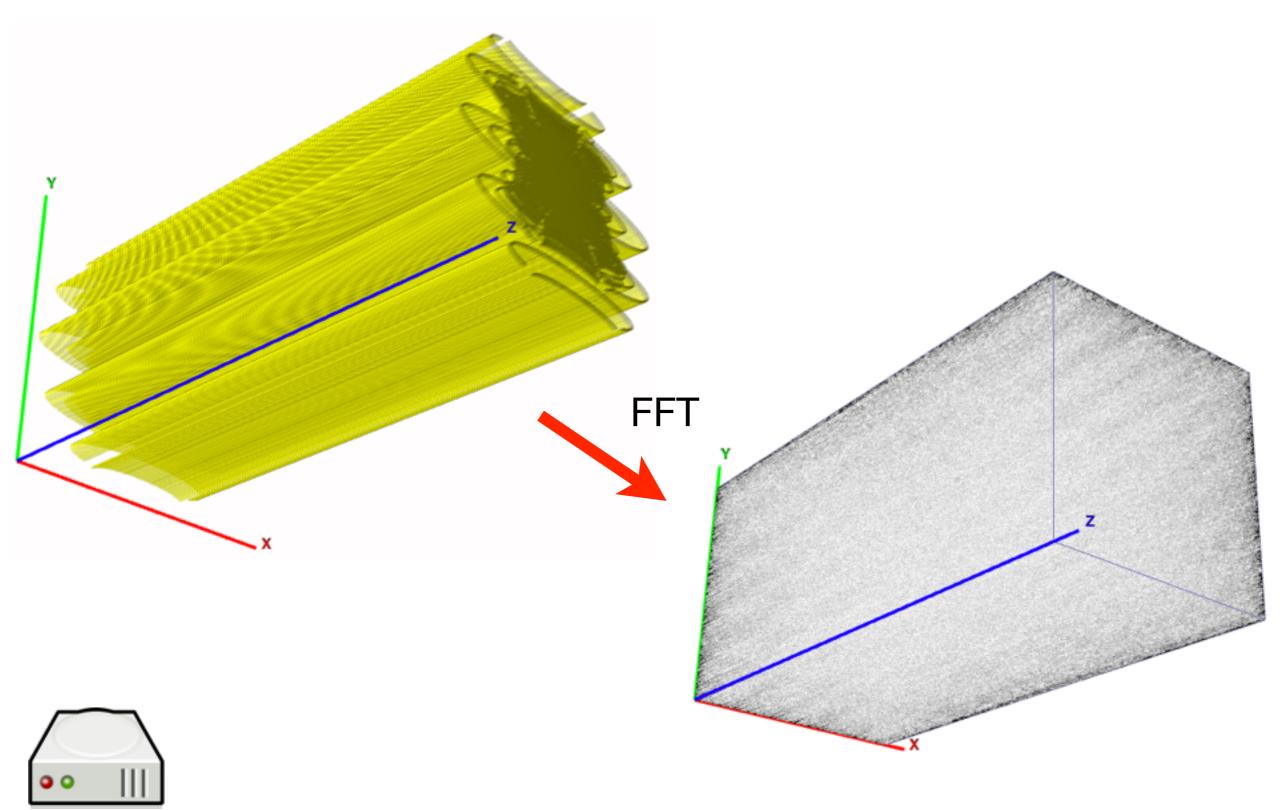


Observing
Calibration
Flagging
Quality Control

Backup
Combination
Imaging

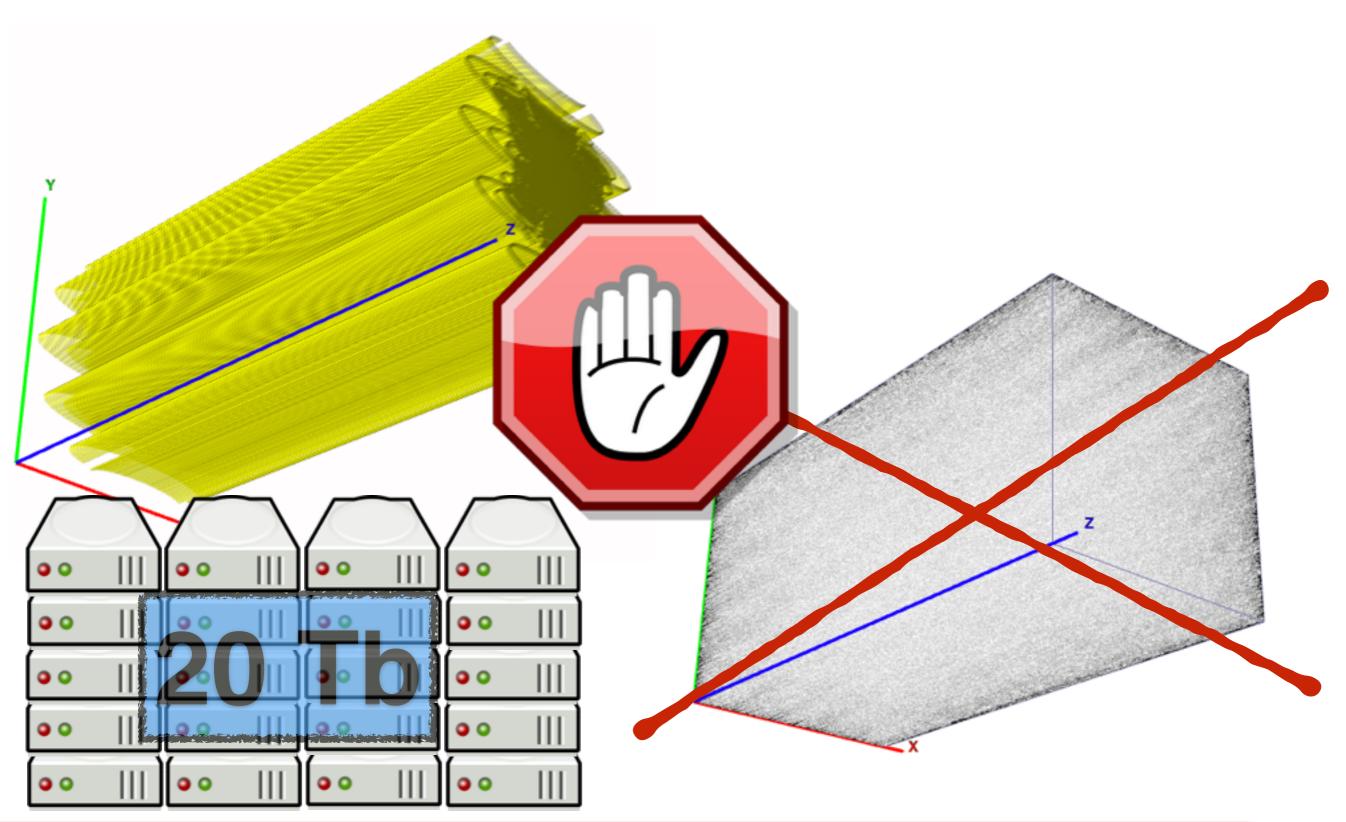


Imaging



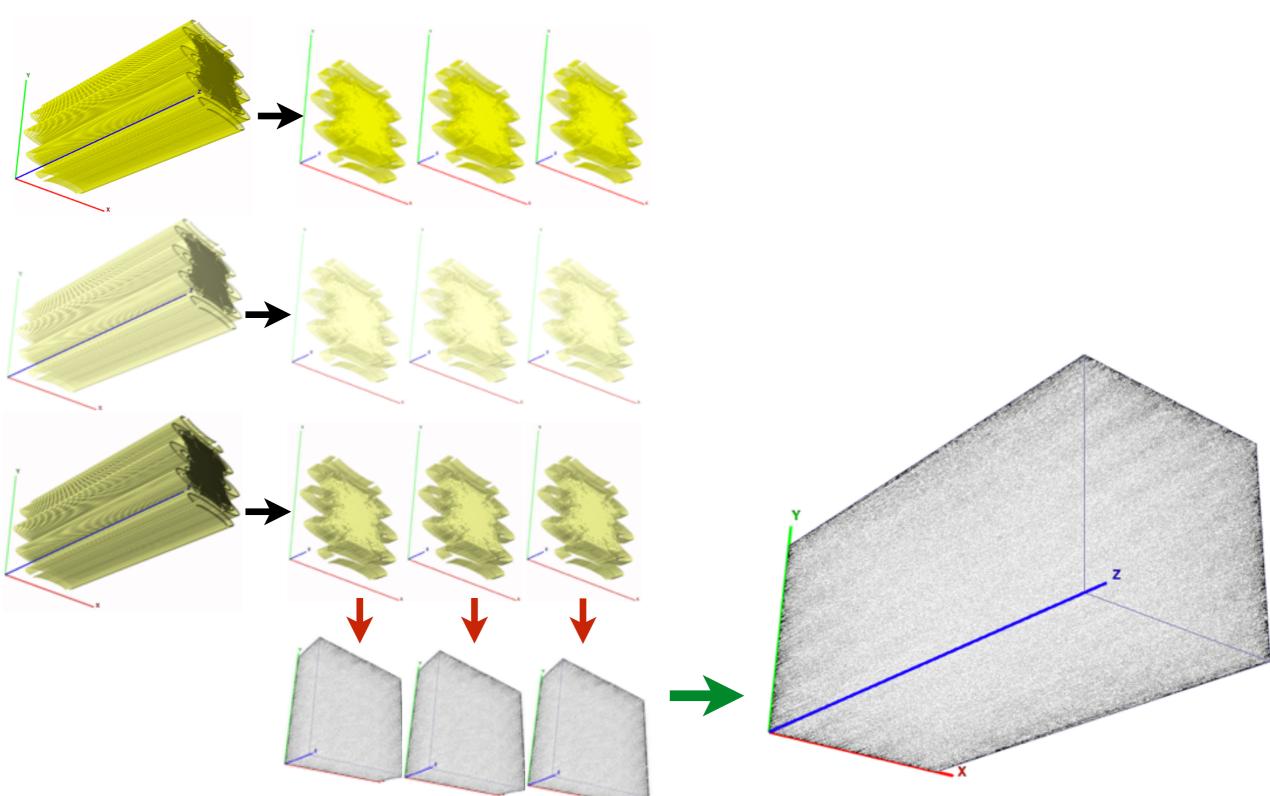


Imaging





Imaging





Data reduction (imaging)

- Exclusively done using CASA
- A lot of decisions we made were to get around limitations features of CASA
- We trialled 3 ways of splitting out the data
 - SPLIT
 - CVEL
 - MSTRANSFORM
- Tried to keep a common code base in GitHub
 - PBS Pleiades
 - SLURM Magnus
 - Python/Boto AWS



Computing efforts

Conventional Cluster (pleiades)
5 nodes each node has 2x Intel
Xeon X5650
2.66GHz CPUs (6 cores / 12 HTs)

Enough computing power, however it would take weeks

Super computer (MAGNUS)
Cray XC40 - 24 cores per node
2.6GHz Intel Xeon E5-2690V3
64GB per Node
35,712 cores available
3PB of storage #58 in the world



AWS
Whatever we wanted
r3.xlarge 16 cores 122GB Ram

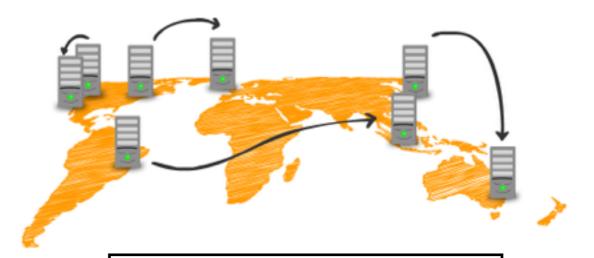




AWS pricing



	On demand	Spot Price
r3.4xlarge	\$1.68	\$0.20
r3.2xlarge	\$0.840	\$0.09
m3.xlarge	\$0.392	\$0.04
m3.medium	\$0.098	\$0.01

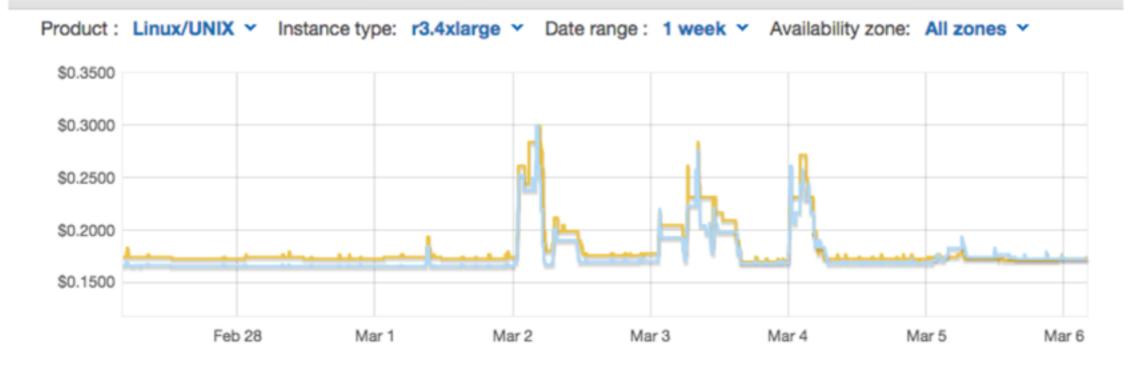


Works!

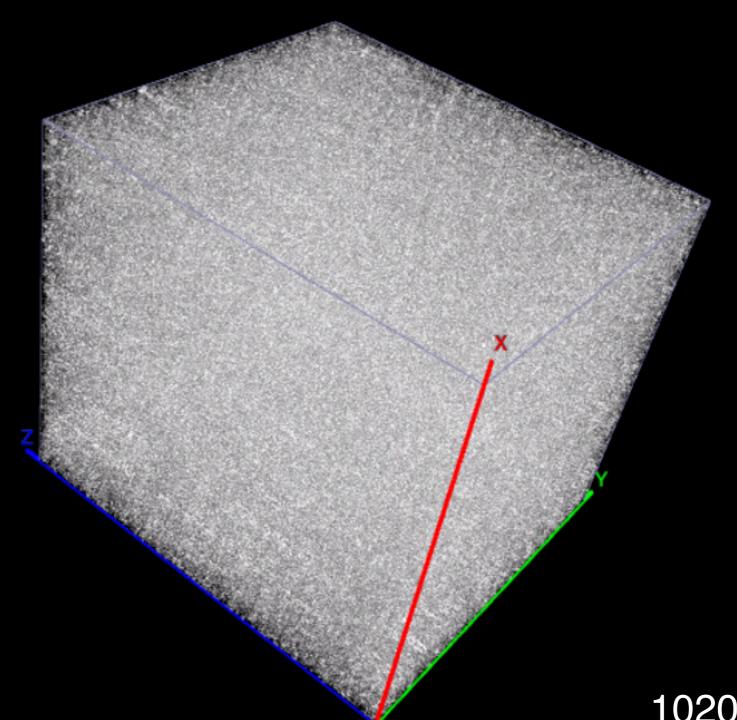
costs so far: ~\$2000

Spot Instance Pricing History

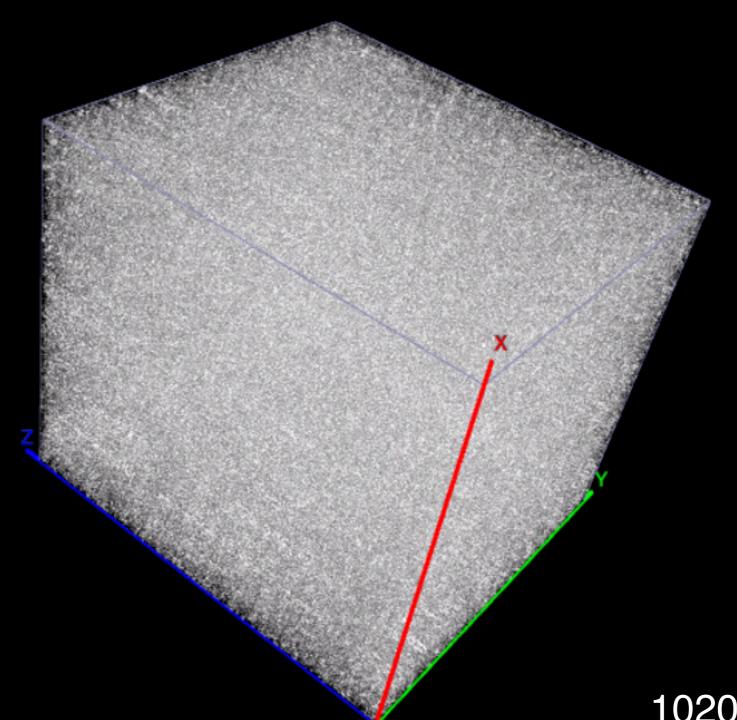




We have now combined 42 observing runs (\sim 20 Tb) and imaged 2048*2048*31.000 pixels (\sim 500 Gb), covering the redshift range z=0 \sim 0.5



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Results - I/O

Operation	Platform	Peak Memory	I/O Throughput	CPU Usage	I/O Characterics	
SPLIT	AWS (EBS)	420MB	<10MB/s	40%	Sequential read/write dominate	
	Magnus	545MB	40 ~ 100 MB/ s	100%		
	Pleiades	390MB	60 ~ 100 MB/ s	100%		
INVERT	AWS (SSD)	60GB	70 ~ 500MB/ s	400%	Random	
	Magnus	30GB	50 ~ 400MB/ s	100%	writes and sequential reads dominate	
	Pleiades	35GB	50 ~ 400MB/ s	400%		



Results

	AWS	Magnus (HPC)	Pleiades
Completion Time	96hr	110hr	1,060hr (est)
Capital Costs	AUD\$0	AUD\$12,000,000	AUD\$50,000
Operational Costs	AUD\$2,000	AUD\$3,240 (free)	_
Control	Root	Limited	Root
Usability	Complex	Good	Good

What next



CHILES - The return of the Jalapeño. (SKA-AWS project)

- Using more AWS working with NRAO
 - NRAO to upload Measurement Sets to AWS Oregon
 - Code run at the Oregon data centre from Perth
 - Scaling Groups will fire off the instances to split the code
 - Code rather than "a person" looking at the spot prices
- Using SKA-SDP Prototype code
 - Docker
 - DROPS
 - CasaPy
- JPEG2000 Visualisation

ICRAR

Conclusions

- Department Cluster
 - Not very satisfactory
- HPC
 - Very fast
 - You can't have root access
 - Installing new software is done by the admin's
 - In WA it is effectively free
- Cloud
 - You can do what you like (a good and a bad thing)
 - EBS volumes are slow
 - Directed attached SSDs are very quick
 - You pay for what you use... And if you forget to turn it off you are still paying



Conclusions

