

Data Movement within the Square Kilometre Array Regional Centre network

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SKAO Mission

"The SKAO's mission is to build and operate cutting-edge radio telescopes to transform our understanding of the Universe, and deliver benefits to society through global collaboration and innovation."



Two telescopes, one observatory, Global Collaboration

SKA Low

131,000 antennas in the Western Australian

SKA mid 197 dishes in South Africa







Global SKA Regional Centre collaboration (SRCNet)



How do users "control" data products?



- Need to satisfy users whilst retaining control of SKAO resources
- Science Data Processor is fundamentally part of each SKA Telescope
- Users will specify required data products in Observing Proposals
- All user interaction with data products will be in SRCs

SKA Regional Centre Capabilities



Dissemination of Data to SRCs and Distributed Data Storage

The Role of SRCs: Collaboration platform

SRCs will bridge the gap between the highly data intensive **pre-defined workflows generating SKA data products** in the SDP, and the **iterative flexible, user-led data analysis** required to produce scientific results

SRCs will provide collaborative tools backed up by powerful compute and data management

Users will <u>not</u> have access to the SDP or to Raw SKA data!



The Role of SRCs: Support data product (re-)use Why

- All SKA Data Products will (in time) become public - this is likely to be the biggest science generator (see Hubble)
 - Build SKA science archive around IVOA standards
 - Ensure interoperability with other archives and other experiments



SKA Regional Centers: Data management

Storing SKAO data growing at up to 700 PBytes each year will be a challenge (plus usergenerated data too).

Several million dollars per year in new data, for one copy

Global data management within SRCNet should enable best possible use to be made of available storage resources

Avoid (reduce) unnecessary duplication

Support mirroring of popular data products to enhance user experience



SRC Network global capabilities



Collectively meet the needs of the global community of SKA users Anticipate heterogeneous SRCs, with different strengths



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Type 1: Planned, Co-located data scenario

Allocation for compute resources made for projects user group at target SRC

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Plan data placement ahead

of time

IENICE DATA PROCESSOE

SKA Data

products

generated

from our live Rucio instance



AFRICA

Products transferred to target SRC

ASIA

Users can be allocated to different SRCs based on data location, HW mapping to compute needs; to even out demand, or other reasons



SOUTH

AMERICA

(Aside) We are prototyping Rucio (continuing our ESCAPE work) for this data placement scenario



Estimates of SKA Data Output... why this is hard!

Co-located scenario is the easiest to predict, but range of SKA science is vast - huge differences in the data volumes required

Examples here are taken from the parametric model used to predict computing needs *within* SKAO, but they also include output data rates

The specific examples are quire out of date - take home message is the range - from <1 to 100 Gbit/s data generation rate, and the fractions of time assumed

SKA1_Low:											
HPSO	Time [%]	Tobs [h]	Npix (side)	Channels (DPrepB)	Channels (DPrepC)	Image size [GB]	Non-Vis Rate [Gbit/s]	Visibility Size [TB]	Visibility Rate [Gbit/s]	Total Rate [Gbit/s]	
hpso01	15.6	5.00	18344	500	1500	2.7	8.5	205.8	91.4	99.9	
hpso02a	15.6	5.00	18344	500	1500	2.7	8.5	205.8	91.4	99.9	
hpso02b	15.6	5.00	18344	500	1500	2.7	8.5	205.8	91.4	99.9	
hpso04a	39.8	0.67	-	-	-	-	0.7	-	-	0.7	
hpso05a	13.4	0.67	-	-	-	-	2.6	-	-	2.6	
Average	-	-	-	-	-	-	4.6	-	42.8	47.4	

HPSO	Time [%]	Tobs [h]	Npix (side)	Channels (DPrepB)	Channels (DPrepC)	Image size [GB]	Non-Vis Rate [Gbit/s]	Visibility Size [TB]	Visibility Rate [Gbit/s]	Total Rate [Gbit/s]
npso04b	1.0	0.17	-	-	-	-	2.3	-	-	2.3
npso04c	3.1	0.17	-	-	-	-	2.3	-	-	2.3
1pso05b	2.1	0.25	-	-	-	-	6.9	-	-	6.9
npso13	6.5	8.00	25339	160	3200	5.1	4.2	-	-	4.2
npso14	2.6	8.00	18814	300	5000	2.8	2.8	-	-	2.8
npso15	16.5	4.40	10837	260	2500	0.9	0.8	-	-	0.8
npso18	13.1	0.02	-	-	-	-	0.1	-	-	0.1
npso22	7.9	8.00	110601	1000	0	97.9	48.1	-	-	48.1
npso27and33	13.1	0.12	23549	700	0	4.4	99.3	-	-	99.3
npso32	13.1	2.20	-	-	-	-	1.3	-	-	1.3
npso37a	13.1	3.80	94195	700	0	71.0	60.6	-	-	60.6
npso37b	2.6	8.00	94195	700	0	71.0	28.8	-	-	28.8
npso37c	2.6	8.00	94195	700	0	71.0	28.8	-	-	28.8
npso38a	1.3	8.00	113204	1000	0	102.5	50.4	-	-	50.4
npso38b	1.3	8.00	113204	1000	0	102.5	50.4		-	50.4
Average	-	-	-	-	-	-	28.4	-	0.0	28.4

https://ska-telescope.gitlab.io/sdp/ska-sdp-par-model/notebooks/SKA1_System_Sizing.html

Type 2: Distributed data scenario, server side actions

Moving compute to the data

Server-side actions

Likely public data

Not SKA Programme Users

Hard to predict

Single workflow running at primary site launching multiple secondary server-side actions at sites holding data

e.g. return image statistics for small area around source location (e.g. redo source parameter analysis using location from a catalogue), just return the parameters



Type 2: Distributed data scenario, server side actions

jupyter

Questions:

How can we predict the network load of this?

Expectation management: can this ever work for interactive jobs (prioritisation??)

Are there similar examples in HEP?

How would this sit alongside the bulk data distribution (planned transfers, known targets, rule-based async)?



Type 3: Remote Data Visualisation

Sé

sion



Questions for SIG

- SKA full operations ~2028, but data transfers and live system from 2026
- How do we collectively prepare our networks to support SKA and HL-LHC data transfers?
- Balance between high-level replication (e.g rucio based, "scientific" data transfers) vs detailed tuning (protocols, DTNs, specific paths)?
- What's the hardest problem? (Type 1, 2, 3)
- Can we (do we need to) mimic / run tests with server-side data operations and results/ visualisation data transfer?
- Mapping to Optical Private Network vs Open Network Environment? What are the advantages of making an SKAONE vs effort required?
- Which models for data distribution are sensible / need testing? (Degree of coordination of transfers vs asynchronous queueing?)
- So far we working assumption of 100Gbit/s (x2 sites) because looks "big enough" (Science ~25-50Gbit/s) - are we aiming too low? or is

