

Delivering SKA Science:

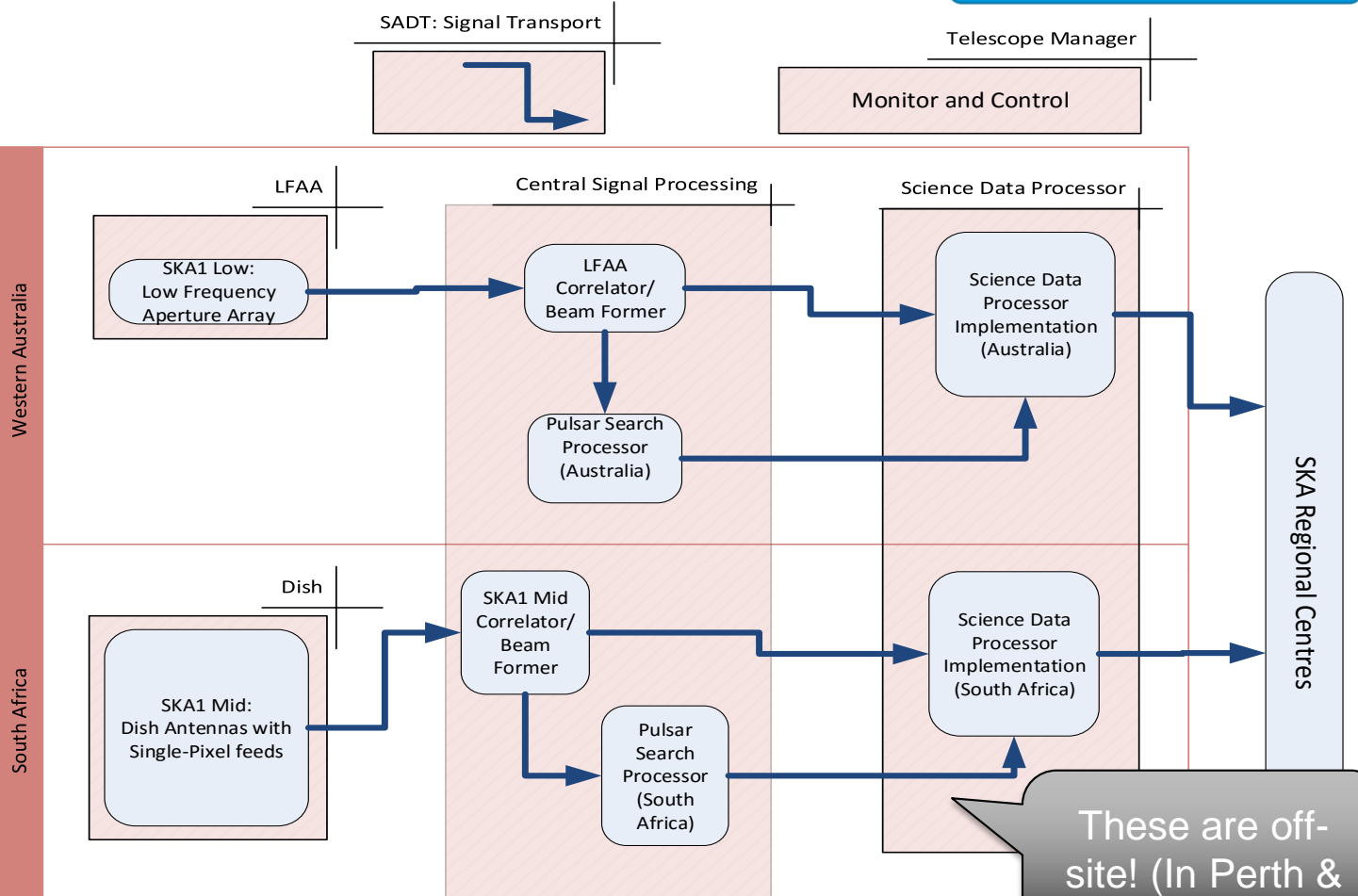
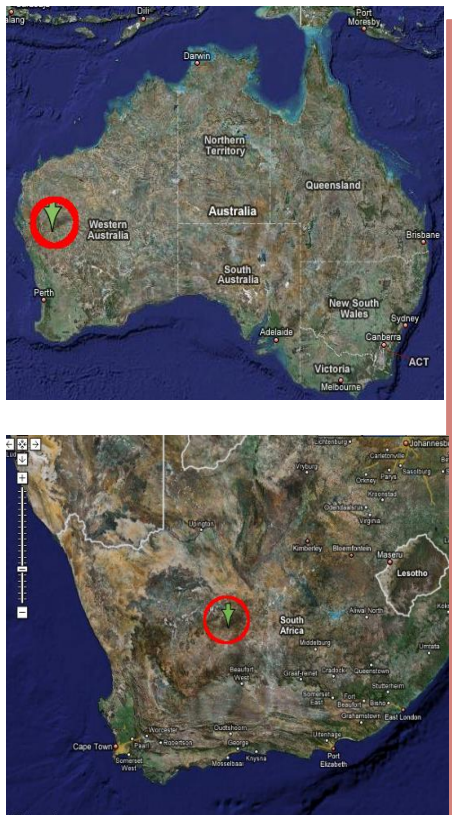
# THE SKA IMAGING AND CALIBRATION CHALLENGE

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University of Cambridge



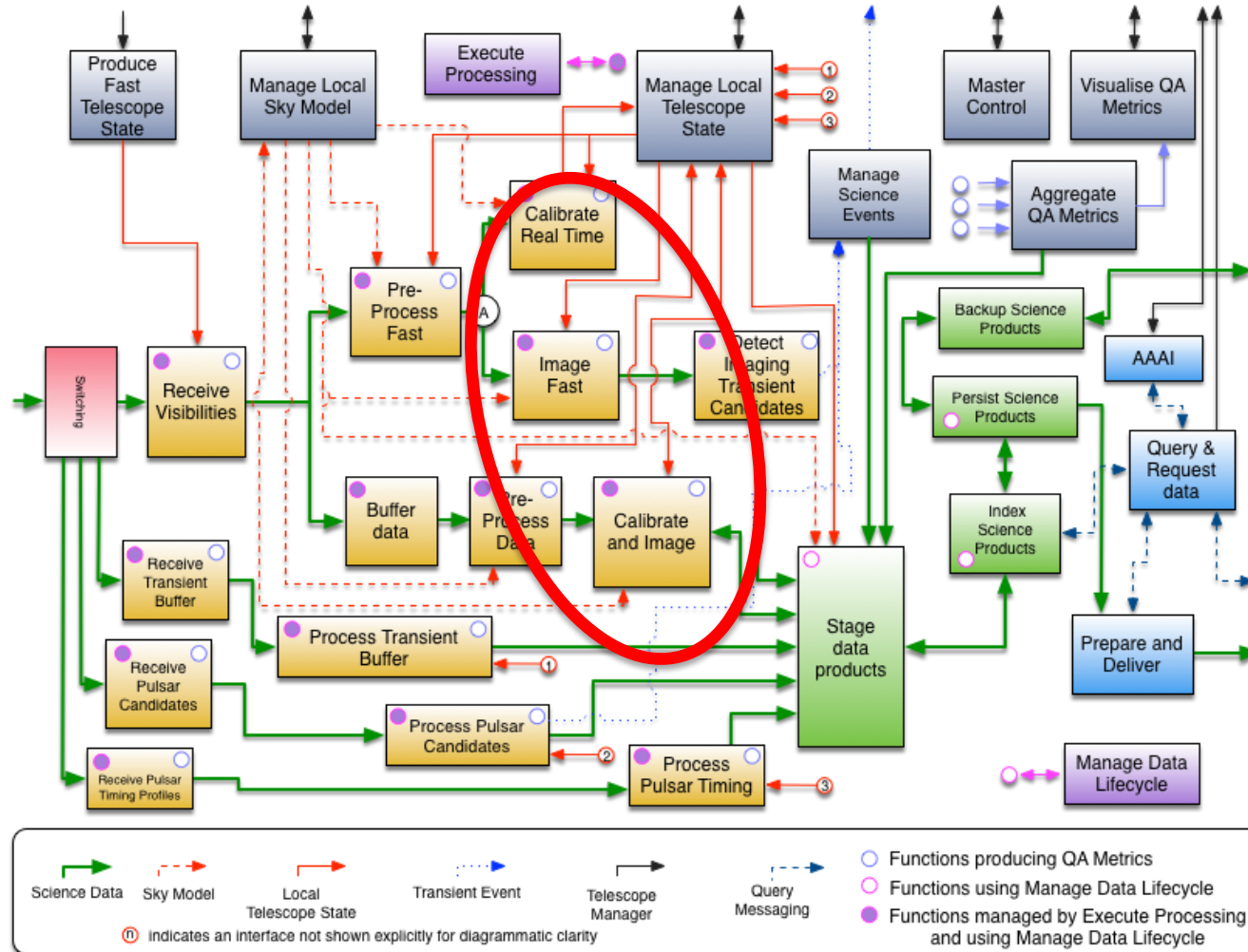
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# SKA Context Diagram



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# Imaging and Calibration Context



# IMAGING AND CALIBRATION PROCESSING IS A MAJOR PART OF THE SKA BY DESIGN



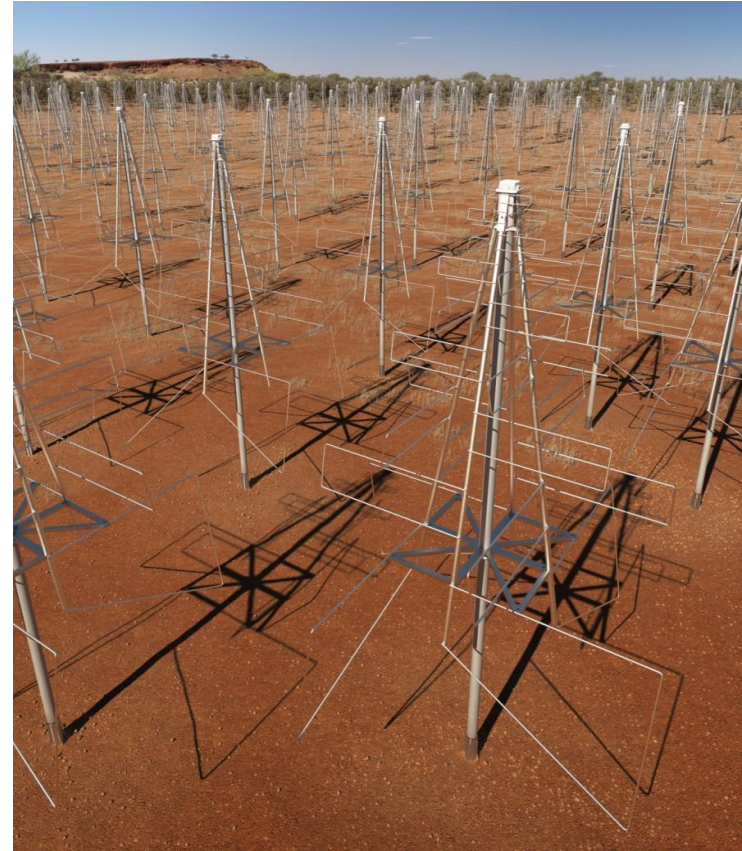
# Large “D” – vs – Large “N”



**GBT 100-m diameter telescope**



**SKA LFAA prototype array**



- No 1 aim: collect as many photons as possible -> high sensitivity
- No 2 aim: collect radiation from different directions -> high survey speed
- No 3 aim: maximum separation of collectors -> high angular resolution



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# Factors driving the SKA challenge



## Very high data rate in

- Unfeasible to permanently store
- Unfeasible to move off-continent
- Expensive to store even temporarily

## Optimal processing strategy, algorithms and parameters unknown:

- Will not be known until the telescope begins operations
- Will depend in part on science goals and demands of individual projects

## High computational requirements to process

- Capital and operational expense
- Hardware/software failures rare for individual computers become frequent



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Direct consequence of:

- Large field of view, fast survey speed (small D)
- High angular resolution (long B)
- High continuum sensitivity (large bandwidth)
- Good sampling (large N)
- Mechanical engineering constraints (SKA1-mid)

-> 0.5 TB/s for each of the telescopes  
100000x ALMA sustained data rate  
10000x ALMA maximum data rate  
1000x JVLA maximum data rate







- Receive, temporarily store incoming data
- Fairly demanding network but in principle can be done today
- Key challenge is:
  - Where to put the data, how to organise it
  - How to process the data



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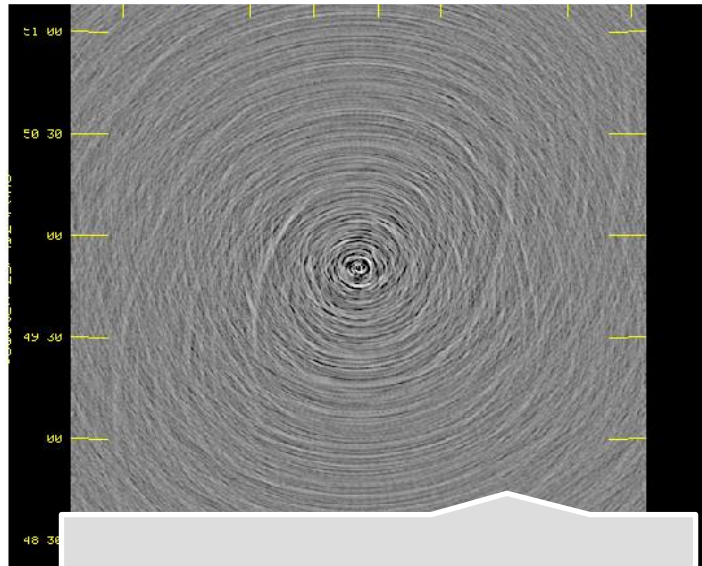


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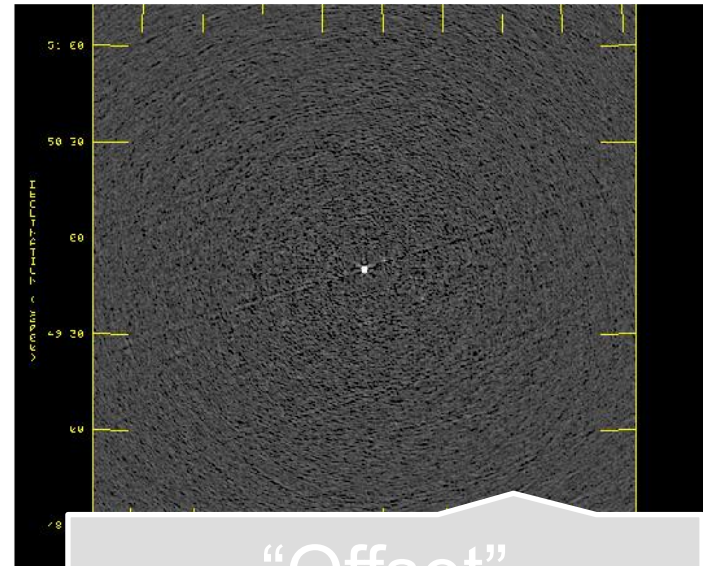
- Time- and frequency-variable corruption of the incoming signal:
  - Atmospheric, mechanical & electronic causes
  - Requires iteratively solving for Sky and the corrupting effects – “Self Calibration”
- Irregular, non-uniform sampling of measurements
  - Requires (typically iterative) de-convolution – CLEAN, Wavelets, compressed sensing, etc
- Non-planar distribution of measurements
  - Approximate correction to the plane required if want to use 2D FFTs



Measurements are imperfect – corrupted by slowly changing mechanical, electrical & atmospheric effects



Uncalibrated



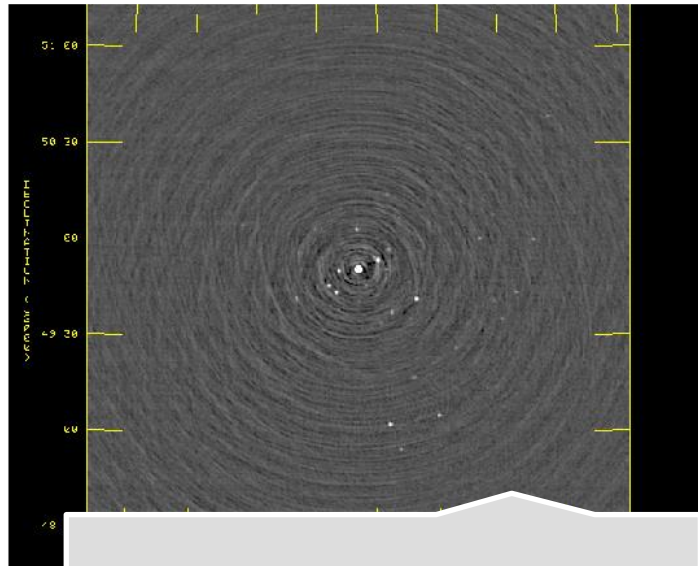
“Offset”  
Calibration

Rick Perley & Oleg Smirnov: “High Dynamic Range Imaging”,  
[www.astron.nl/gerfeest/presentations/perley.pdf](http://www.astron.nl/gerfeest/presentations/perley.pdf)

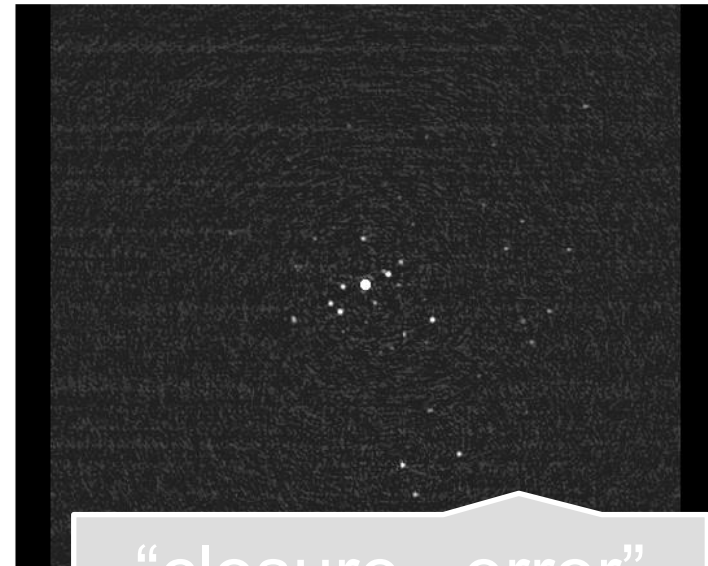


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# Iterative & joint solving for the image of the Sky & Calibration



“Self-Calibration”



“closure –error”  
calibration

Rick Perley & Oleg Smirnov: “High Dynamic Range Imaging”,  
[www.astron.nl/gerfeest/presentations/perley.pdf](http://www.astron.nl/gerfeest/presentations/perley.pdf)



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## SKA/SDP Design:

- To support current best-practice algorithms:
  - Multi-frequency multi-scale CLEAN
  - Self-calibration
  - Direction dependent correction using “A” terms
- Flexibility to update and improve in future

Important role for ongoing current research and future optimisation and commissioning



## SKA/SDP Design:

- To support current research
  - Multi-frequency
  - Self-calibration
  - Direction dependent correction using “A” terms
- Flexibility to update and adapt

Challenge: Can these algorithms be expressed scalably?

Need >1000x improvement from current proven scales

Challenge: too much flexibility – nothing ever works

Important role for ongoing current research and future optimisation and commissioning



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- ~100 PetaFLOPS total achieved
- ~200 PetaByte/s aggregate BW to fast working memory
- ~50 PetaByte Storage
- ~1 TeraByte/s sustained write to storage
- ~10 TeraByte/s sustained read from storage
  - ~ 10000 FLOPS/byte read from storage



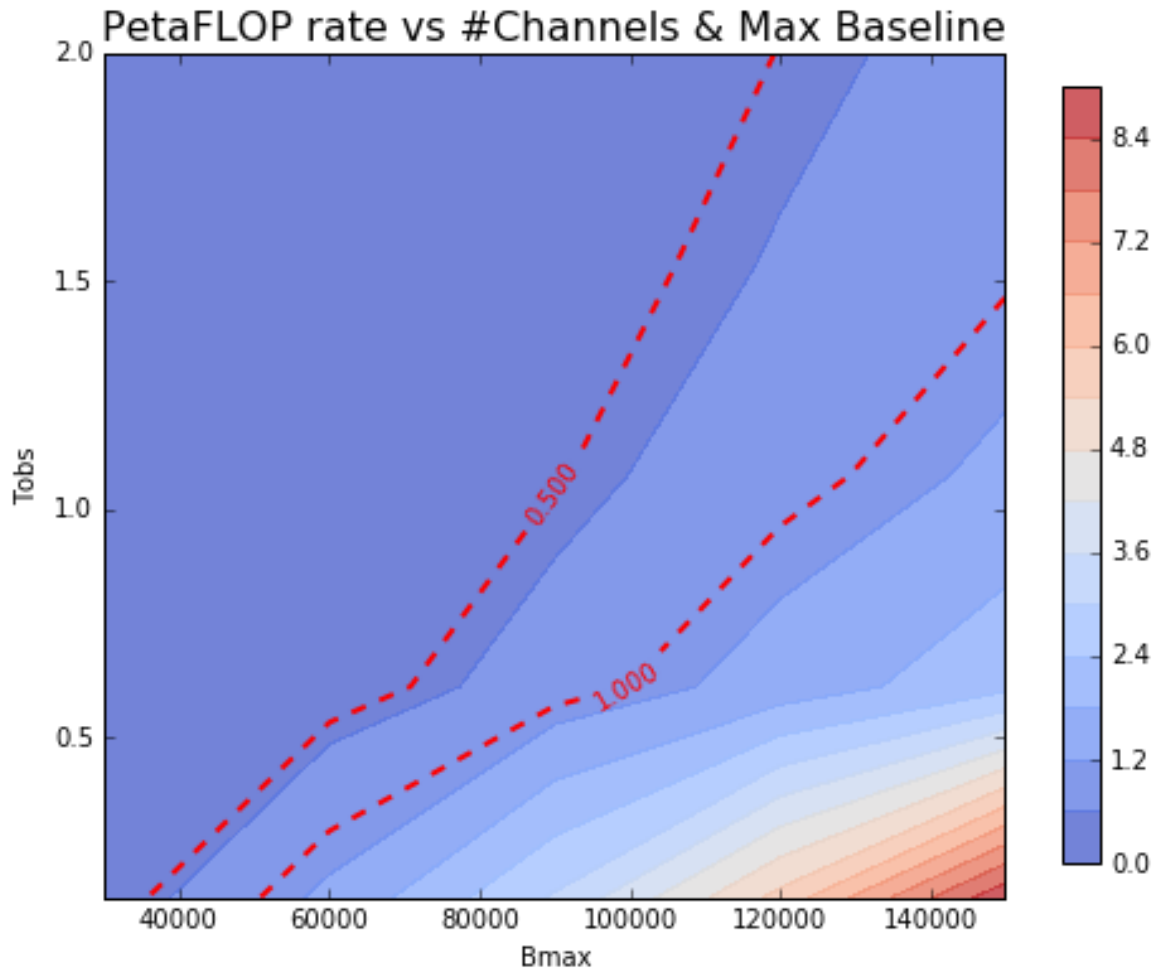
# Illustrative Computing Requirements



- ~100 PetaFLOPS total achieved Likely to be achievable ~ 2020
- ~200 PetaByte/s aggregate BW One of the big challenges  
working memory
- ~50 PetaByte Storage Also likely to be achievable well ahead of SDP roll out
- ~1 TeraByte/s sustained write One of the big challenges
- ~10 TeraByte/s sustained read from storage
  - ~ 10000 FLOPS/byte read from storage



# Parametric Model Example

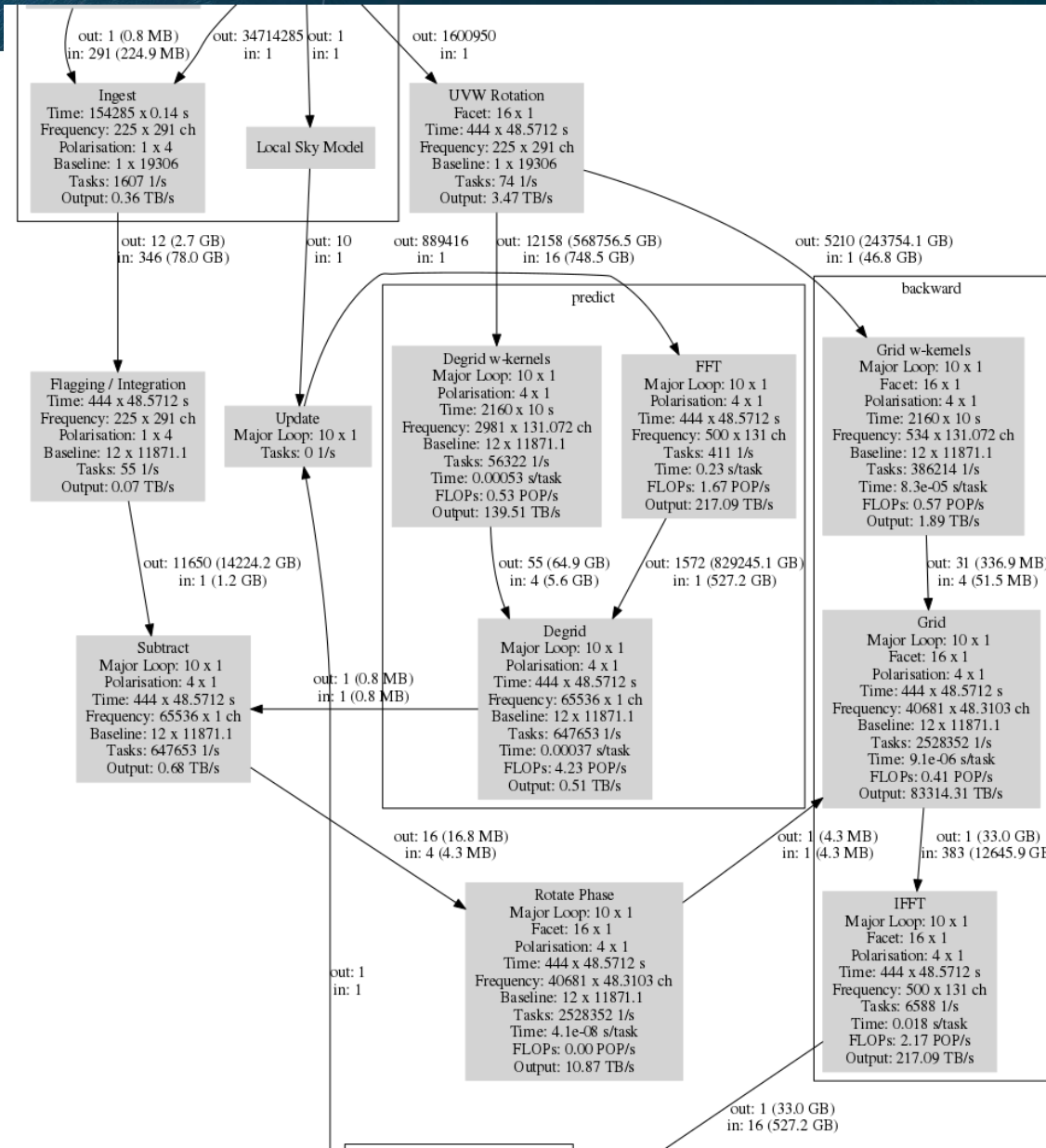


Computational cost of a transient survey as a function of integration of each pointing and maximum baseline length that is used



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# Computational requirements breakdown





- Document the computational requirements, their relationship to the SDP requirements
- Document the roadmap for likely evolution of computing systems
- Ensure the SDP software architecture can make reasonably efficient use of likely future computing system
- Ensure the maintenance of software is tractable, especially across changes in future computing system architectures
- Prototyping to provide evidential support to the above, demonstrate appropriate technical readiness of potential solutions



# Factors driving the SKA challenge



Critical learning period during commissioning and early operations

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High Degree of Parallelism, automatic unsupervised pipelines

**SDP**

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# SDP Top-level Components & Key Performance Requirements -- SKA Phase 1



Telescope Manager

Science Data Processor

SDP Local Monitor & Control

Data Processor

High Performance

- ~100 PetaFLOPS

Data Intensive

- ~100 PetaBytes/observation (job)

Partially real-time

- ~10s response time

Partially iterative

- ~10 iterations/job (~3 hour)

Data Preservation

High Volume & High Growth Rate

- ~100 PetaByte/year

Infrequent Access

- ~few times/year max

Delivery System

Data Distribution

- ~100 PetaByte/year from Cape Town & Perth to rest of World

Data Discovery

- Visualisation of 100k by 100k by 100k voxel cubes

CSP

1 Tera Byte/s

Regional Centres & Astronomers



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Data Preservation

**Delivery System**

- Data Distribution
  - ~100 PetaByte/year from Cape Town & Perth to rest of World
- Data Discovery
  - Visualisation of 100k by 100k by 100k voxel cubes

Goal is to extract information from data and then discard the data

CSP

1 Tera Byte/s

Regional Centres & Astronomers





- Hybrid programming model:
  - Dataflow at **coarse-grained** level:
    - About 1 million tasks/s max over the whole processor (-> ~10s – 100s milli second tasks), consuming ~100 MegaByte each
    - **Static scheduling** at coarsest-level (down to “data-island”)
      - Static partitioning of the large-volume input data
    - **Dynamic scheduling** within data island:
      - Failure recovery, dynamic load-balancing
    - *Data driven* (all data will be used)
  - Shared memory model at **fine-grained** level e.g.: threads/OpenMP/SIMT-like
    - ~100s active threads per shared memory space
    - Allows manageable working memory size, computational efficiency





- Extremely challenging to deliver early in operations
- Very challenging to deliver for a diverse set of science programmes and goals
- Unsatisfactory performance will lead to low observatory efficiencies



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SDP  
Design Phase

SDP

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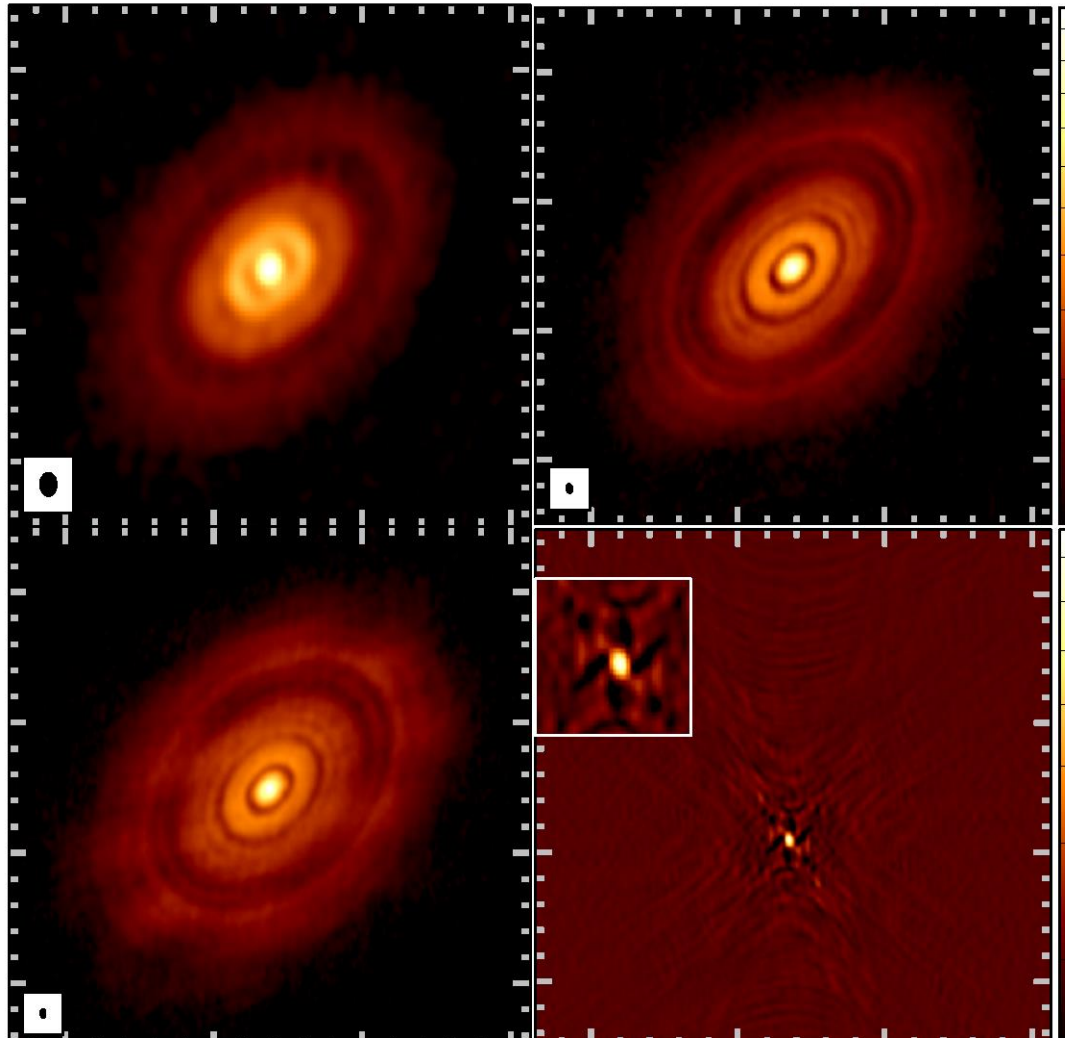
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# Long term commissioning and optimisation

## Example from ALMA



**First Results from High Angular Resolution ALMA Observations Toward the HL Tau Region, ALMA Partnership, [2015ApJ...808L...3A](#)**  
**Result of collaboration of observatory staff, institutes and universities to characterise and commission ALMA long baselines**



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