

Lecture 7

Calibration:

Science Data Processing

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UNIVERSITY OF CAMBRIDGE

Syllabus Overview

- Introduction to Big Data Radio Astronomy and Key Science Projects
 - o Lecture 1: SKA Key Science Projects
 - o Lecture 2: Brief history of radio astronomy and the SKA telescope
 - Lecture 3: The modern "large-N" radio interferometers
- Instrument simulations and design tools
 - Lecture 4: Intro into numerical methods for electromagnetic modelling
 - o Lectures 5 and 6: Mutual coupling in antenna arrays
- Science Data Processing
 - Lecture 7: Calibration of radio observations
 - Lecture 8: Imaging techniques
 - Lecture 9: Advanced imaging techniques
 - Lecture 10: Time-domain radio astronomy
- Computing infrastructure
 - o Lecture 11: Federation and scaling approaches for exascale data
 - Lecture 12: Data centre challenges and opportunities
- Advanced ML and Bayesian methods for data analysis and science extraction
 - Lecture 13: Nested sampling and MCMC
 - Lecture 14: Applications of Bayesian analysis
 - Lecture 15: Signal emulation for astrophysics and cosmology
 - Lecture 16: Simulation-based inference in astrophysics and cosmology

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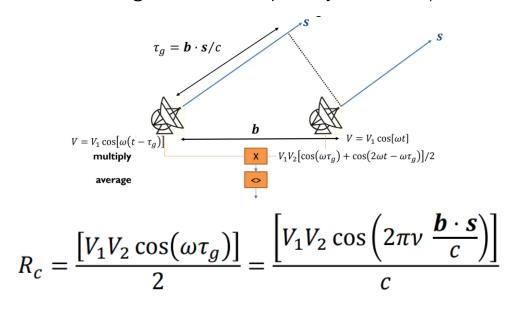
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Mapping the radio sky

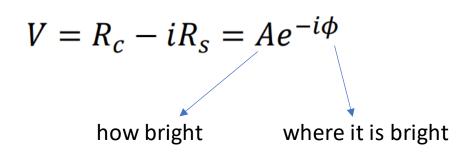
- In astronomy, we are interested in knowing the angular distribution of electromagnetic emission
- This means, we are interested in the *surface brightness* of the emission
 - Knowing how the brightness is distributed over the angular extent of the source (how bright and where it is bright)
- Modern radio interferometers measure complex visibilities (the interference pattern between pairs of antennas at a single unit of frequency and time):



$$V=R_c-iR_s=Ae^{-i\phi}$$
 how bright where it is bright



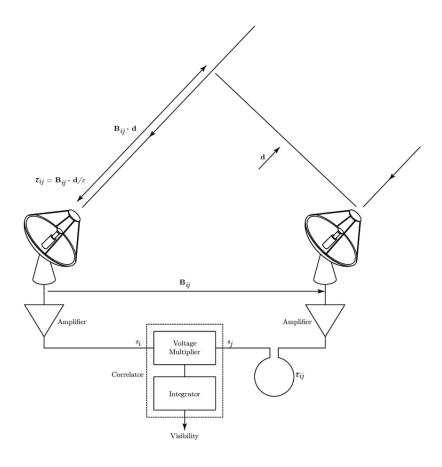
Mapping the radio sky



- The correlator outputs complex visibilities for each pair of antennas, for each unit of frequency, for each unit of time, for each polarization (most dishes measure 4 polarizations)
- These form rows of data (currently called Measurement Set) that are sent to the scientists

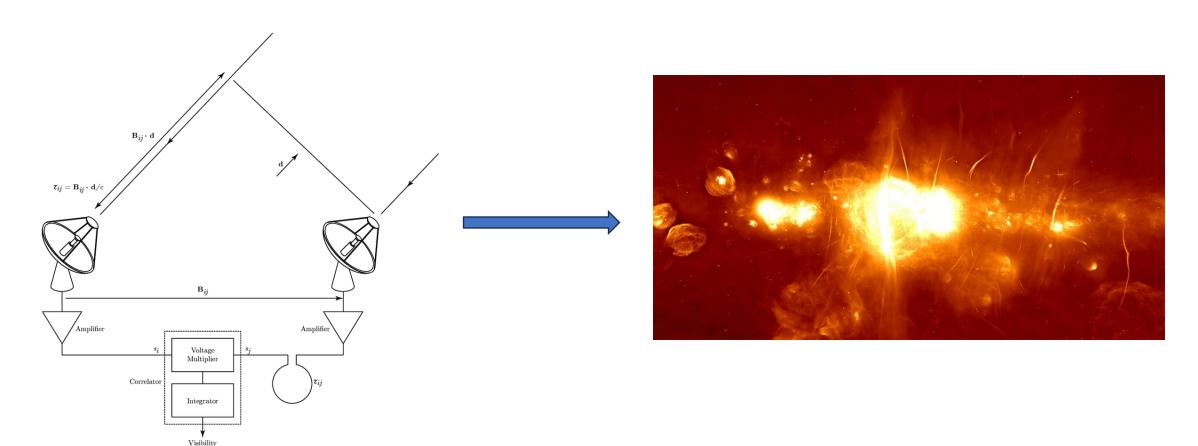


Modern radio telescopes



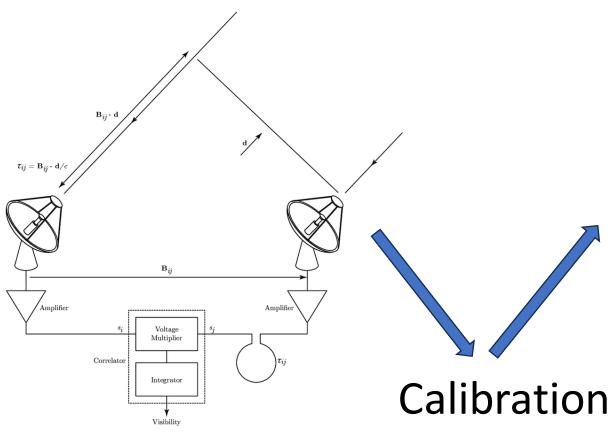


Modern radio telescopes





Modern radio telescopes







Calibration

What is wrong with taking radio interferometer visibilities and making a map of the sky?



Calibration

- What is wrong with taking radio interferometer visibilities and making a map of the sky?
 - Changing weather/atmosphere/ionosphere can distort radio signals
 - Instrumental differences between antennas (dish imperfections, cable lengths, electronics)
 - The signals after correlation do not have physical units
 - o Geometric delays between antennas not calculated perfectly
 - Bandpass filters not perfect
 - o Telescope 'clocks' may not be perfect
 - Telescope 'beam' changes with time/frequency/where it points

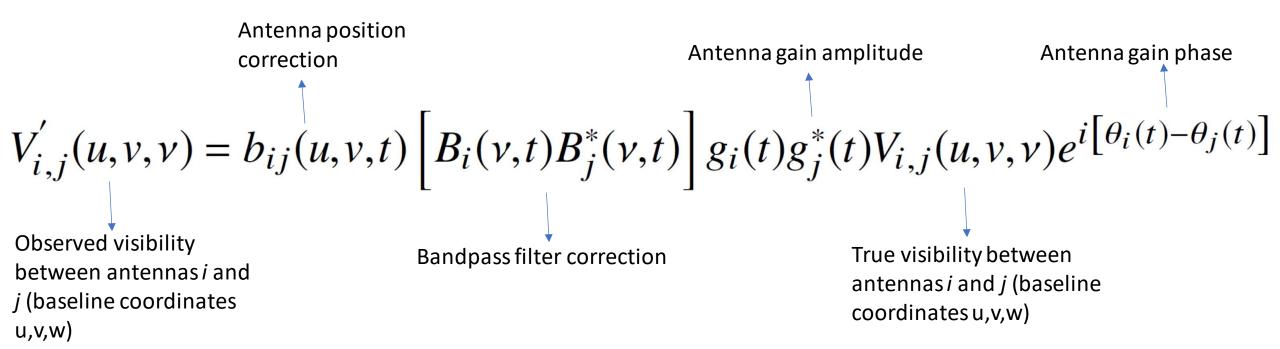


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 - Telescope 'beam' changes with time/frequency/where it points
- Calibration generally means to adjust the measurements of an instrument based on a standardized scale.
- For radio interferometers, this process tries to remove nasty effects on the measured visibilities, so that the radio observations become what we expected the interferometer to see

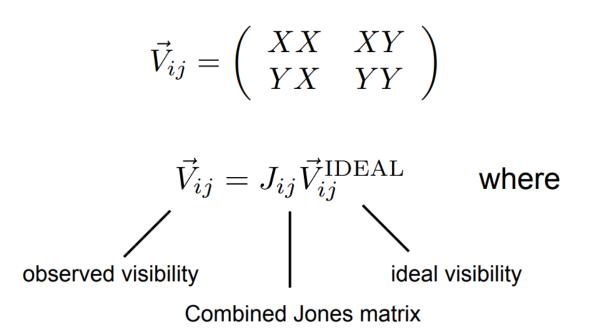


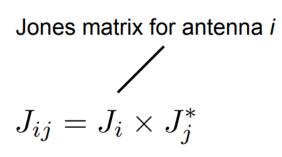
Calibration – general formalism





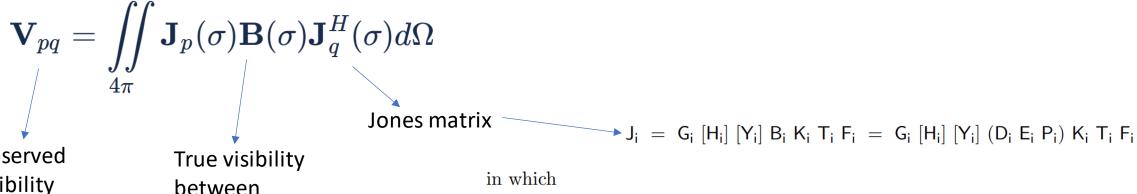
Calibration – Jones formalism







Calibration – Jones formalism



Observed visibility between antennas p and q

True visibility between antennas p and q

 $\mathsf{F}_{\mathsf{i}}(\vec{\rho}, \vec{r}_{\mathsf{i}})$ ionospheric Faraday rotation atmospheric complex gain $\mathsf{T}_{\mathsf{i}}(\vec{\rho}, \vec{r_{\mathsf{i}}})$ $K_i(\vec{\rho}.\vec{r_i})$ factored Fourier Transform kernel P_{i} projected receptor orientation(s) w.r.t. the sky $\mathsf{E}_{\mathsf{i}}(\vec{\rho})$ voltage primary beam position-independent receptor cross-leakage Di $[Y_i]$ commutation of IF-channels hybrid (conversion to circular polarisation coordinates) $[H_i]$ G_{i} electronic complex gain (feed-based contributions only)



Calibration – in practice

How do we determine the Jones matrix/correction factors to calibrate our interferometer data?



Calibration – in practice

- How do we determine the Jones matrix/correction factors to calibrate our interferometer data?
- Basic principle: have a model which is a well known representation of the radio source, and compare it to your observed visibilities least squares fitting



Calibration – in practice

- How do we determine the Jones matrix/correction factors to calibrate our interferometer data?
- Basic principle: have a model which is a well known representation of the radio source, and compare it to your observed visibilities least squares fitting
- Let's say you want to calibrate your data to solve for the complex antenna gains:

$$V_{i,j}' = g_i g_j^* e^{i[\theta_i - \theta_j]} V_{i,j}^{\text{model}}$$

- Have the observed visibilities, predict the model visibilities. Find which correction factors are required
 to minimize the difference one set of Jones matrices determined.
- Find all the Jones matrices, then do a single 2x2 matrix multiplication to the visibilities to calibrate.
- All about the model: need the most truth, but also should be representative of what the telescope would see (think about resolution)

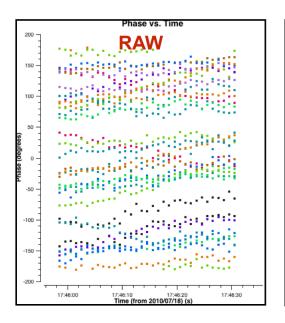
Calibration – where do we get the model

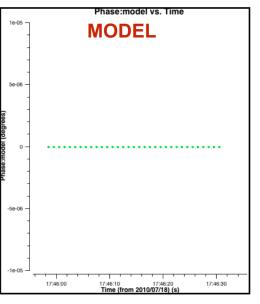
- A model is an astronomical source of known brightness and structure
- What type of object is the best model?

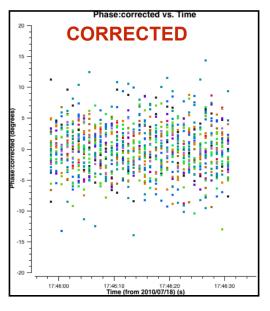


Calibration – choosing a model

- A model is an astronomical source of known brightness and structure
- What type of object is the best model?
 - Known brightness with the least amount of error
 - Point source a point source observed at the phase center (dish pointing directly at it) has visibilities with zero phase (waves completely in phase).









- During a radio telescope observation, point your telescope at:
 - Your target source of interest
 - o A well-known flux calibrator which has known brightness (direction-independent Jones matrices)
 - A phase calibrator a bright point source somewhere near the target source (directiondependent Jones matrices)



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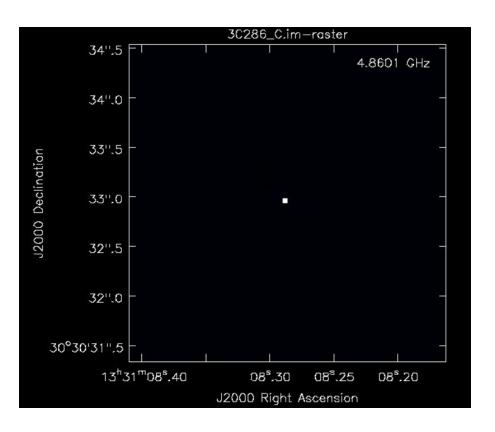
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MS Version 2
Observer: Prof. Julie Hlavacek-Larrondo Project: uid://evla/pdb/37594215
Observation: EVLA
Data records: 9219600
        Total elapsed time = 8970 seconds
Observed from 05-Jul-2020/04:45:55.0 to 05-Jul-2020/07:15:25.0 (UTC)
ObservationID = 0
        ArrayID = 0
                        SpwIds Average Interval(s) ScanIntent
   Timerange (UTC)
           Scan FldId FieldName
05-Jul-2020/04:45:55.0 - 04:51:20.0
              0 3C286
                     04:51:25.0 - 04:52:20.0
              0 3C286
                     04:52:25.0 - 04:58:50.0
              0 3C286
                     04:58:55.0 - 05:00:50.0
              1 J1407+2827
                     05:00:55.0 - 05:05:45.0
              2 J1330+2509
                     05:05:50.0 - 05:14:45.0
              3 NGC4869
   05:14:50.0 - 05:23:45.0
              3 NGC4869
                     05:23:50.0 - 05:32:45.0
   05:32:50.0 - 05:34:40.0
              2 J1330+2509
                     05:34:45.0 - 05:44:40.0
              3 NGC4869
              3 NGC4869
                     05:44:45.0 - 05:54:40.0
                     05:54:45.0 - 06:04:35.0
              3 NGC4869
                     06:04:40.0 - 06:06:35.0
              2 J1330+2509
   06:06:40.0 - 06:16:35.0
              3 NGC4869
                     06:16:40.0 - 06:26:35.0
              3 NGC4869
                     06:26:40.0 - 06:36:30.0
              3 NGC4869
                     06:36:35.0 - 06:38:30.0
              2 J1330+2509
   06:38:35.0 - 06:48:30.0
              3 NGC4869
                     06:48:35.0 - 06:58:30.0
              3 NGC4869
                     06:58:35.0 - 07:08:25.0
              3 NGC4869
                     07:08:30.0 - 07:10:25.0
              2 J1330+2509
                     0 3C286
   07:10:30.0 - 07:15:25.0
   (nRows = Total number of rows per scan)
Fields: 4
ID Code Name
                  Epoch SrcId
                        nRows
 NONE 3C286
         13:31:08.287984 +30.30.32.95886 J2000
                       1102400
 NONE J1407+2827
         14:07:00.394410 +28.27.14.68997 J2000
                       119600
 NONE J1330+2509
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                       774800
 NONE NGC4869
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12:59:23.330000 +27.54.41.79999 J2000

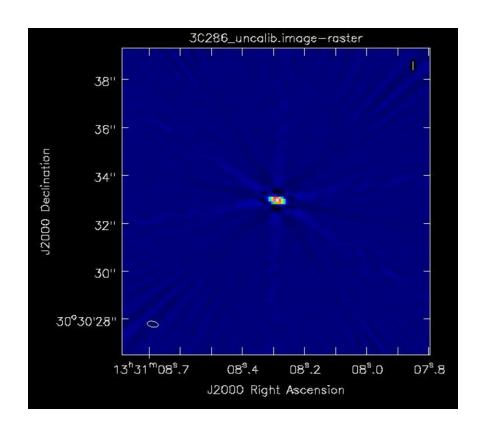


- 3 types (done in order)
 - Direction-independent calibration
 - Time-independent quantities use a well-known astronomical object with known brightness
 - Direction-dependent calibration
 - Time-dependent quantities use a bright point source nearby to your target
 - Self-calibration
- Determine all these corrections and store them in tables before making an image apply all corrections to visibilities



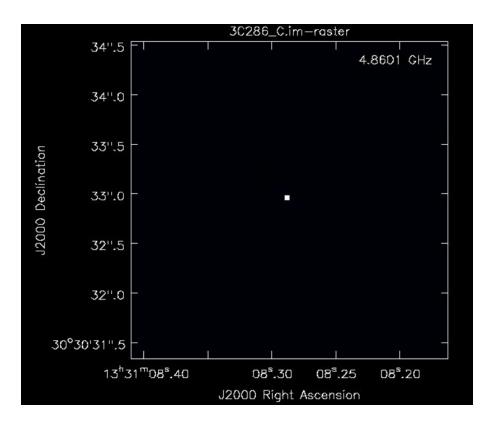


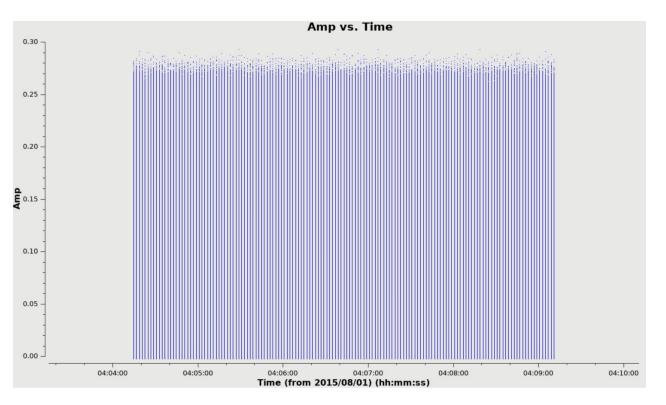
3C286 5 GHz model



3C286 5 GHz VLA observation



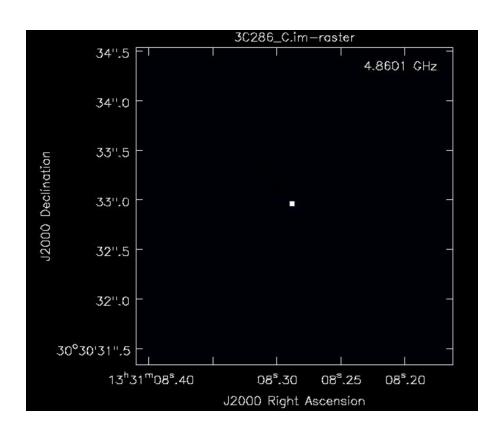


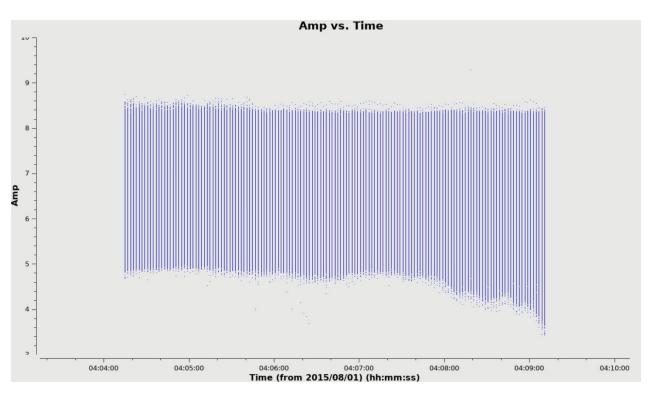


3C286 5 GHz model

3C286 5 GHz VLA observation



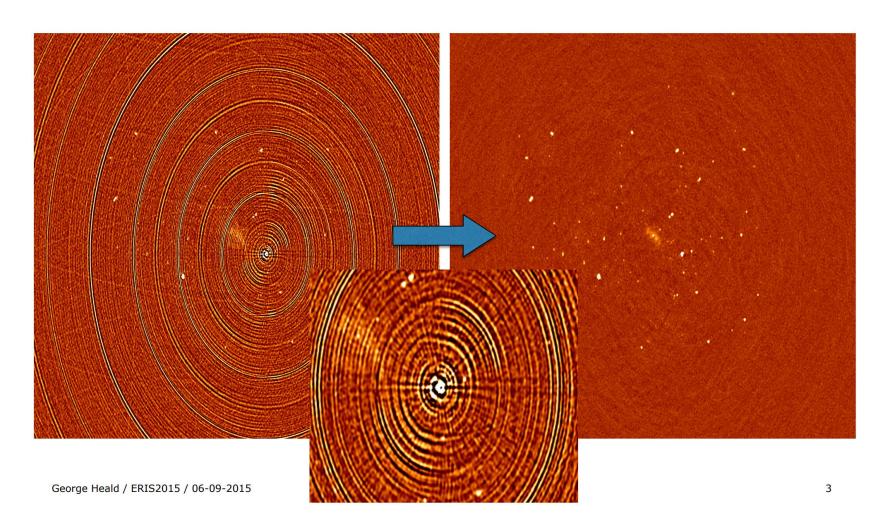




3C286 5 GHz model

3C286 5 GHz VLA observation Calibrated, with correct brightness units





Proof that you can't simply image (take a Fourier transform of) visibilities straight after an observation – calibration is necessary for science quality



- Is there more data processing to be done after calibration?
- The visibility amplitudes and phases should now be correct, no?



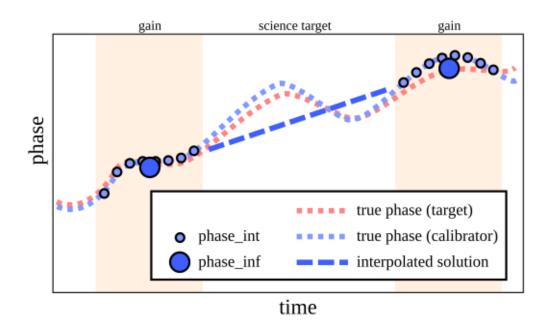
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• Why is calibration not enough?

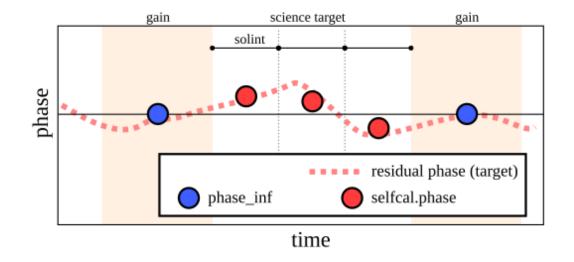


- Why is calibration not enough?
 - Calibrators observed at a different time than the target – linear interpolation may be incorrect assumption
 - Calibrator models may not be perfect
 - Calibrators might not have enough S/N to follow tiny phase variations with time

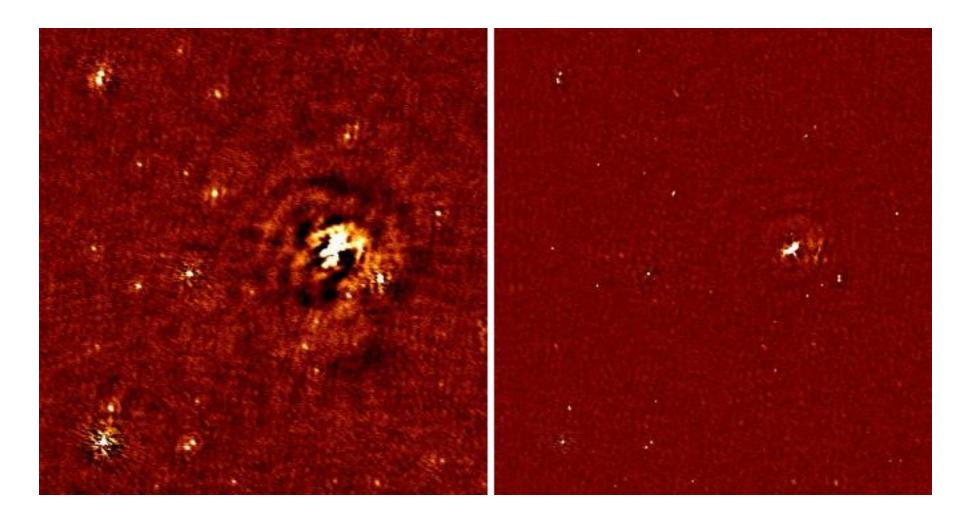




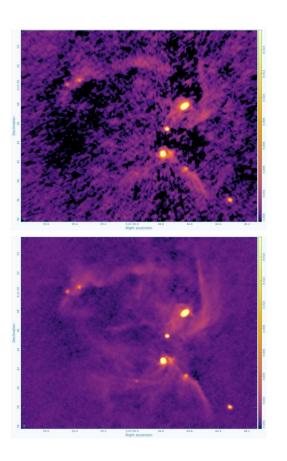
- General idea
 - Image your science target after calibration
 - 2. Take the brightest parts of those radio sources to generate a model
 - 3. Calibrate your data with that model (phase, amplitude)
 - 4. Image the new calibrated data
 - Repeat from step 1
- Idea is that the model improves with every cycle

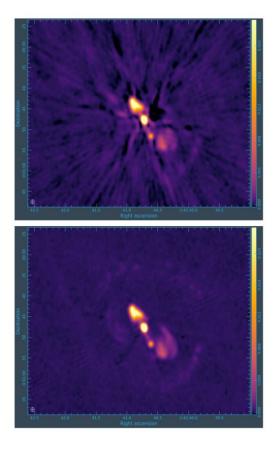


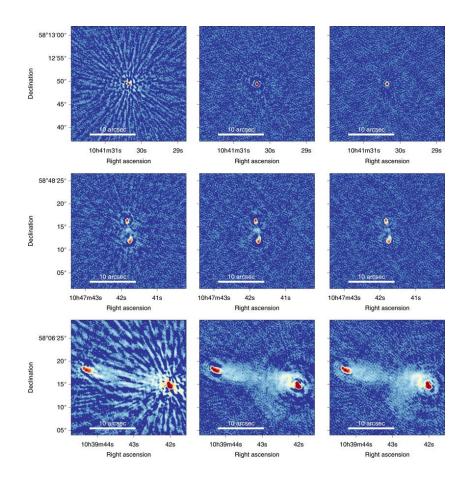








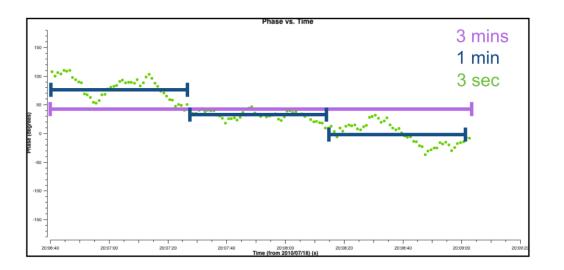






General remarks

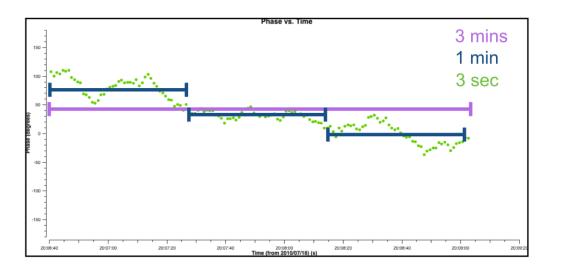
- Jones matrix solutions are not static several effects (e.g. the ionosphere are time, frequency, baseline, polarization dependent)
- We don't have one solution, but many, for each type
- Determining solution intervals in time, frequency etc depends on many things – how bright your calibrator is, how well behaved the ionosphere is during the observation.
- Errors at any step (e.g. incorrect model, poor choice of solution interval) will lead to the snowball effect and a poor image. Rubbish in=rubbish out





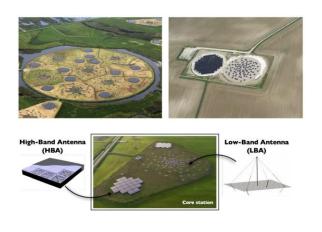
General remarks

- Calibration is an over-determined problem
- We have N antennas to calibrate, but there are N(N-1)/2 unique baselines (data points)
- When N>3, we have a lot more constraints than unknowns: improved accuracy and reliability
- But this increases computational complexity
 - o Imagine doing regression fitting (getting all antenna visibility phases and amplitudes from all its baselines) over 10 solution intervals in time and 10 solution intervals in frequency for each antenna, multiplying all the Jones matrices and applying them to a 10 TB dataset

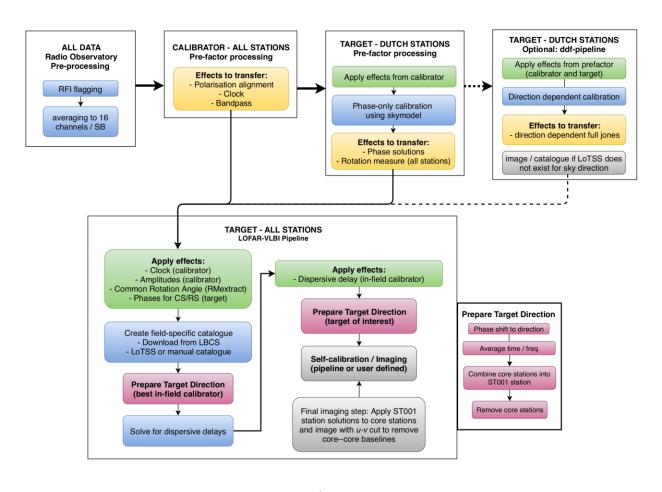




Modern calibration pipelines



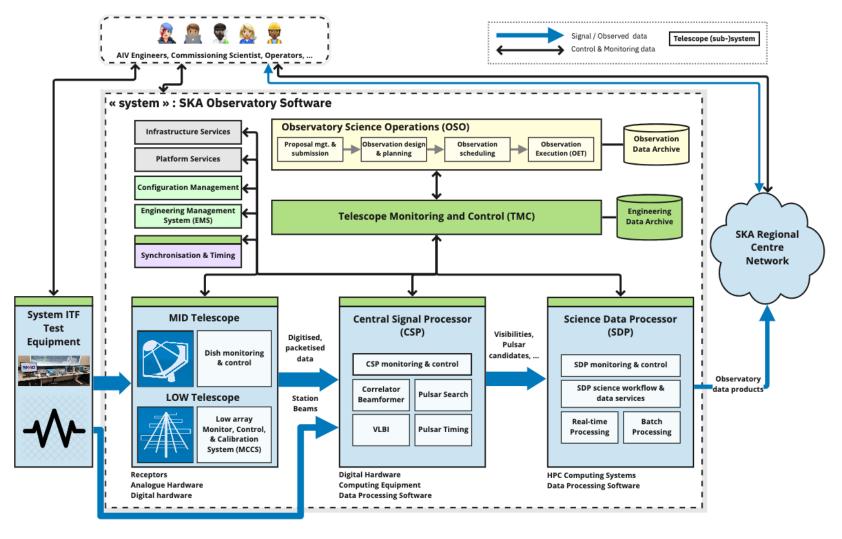




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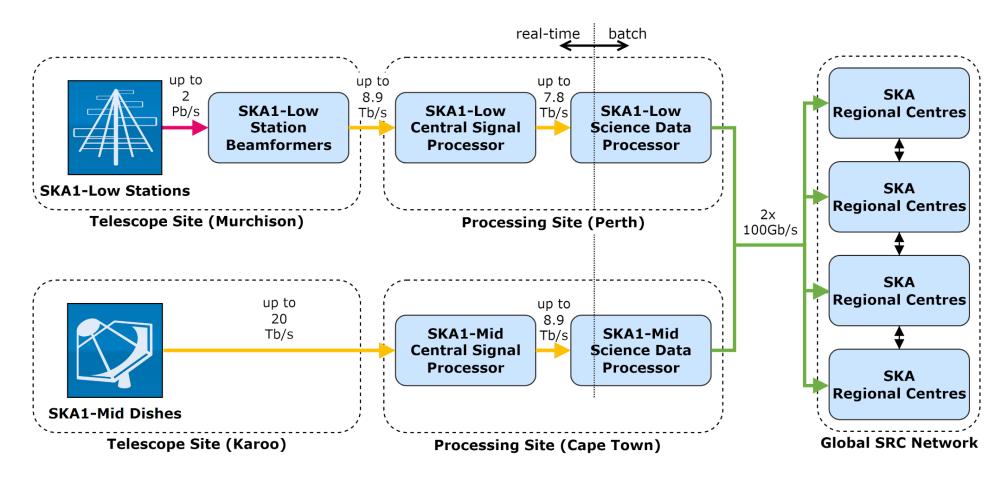


SKA system overview



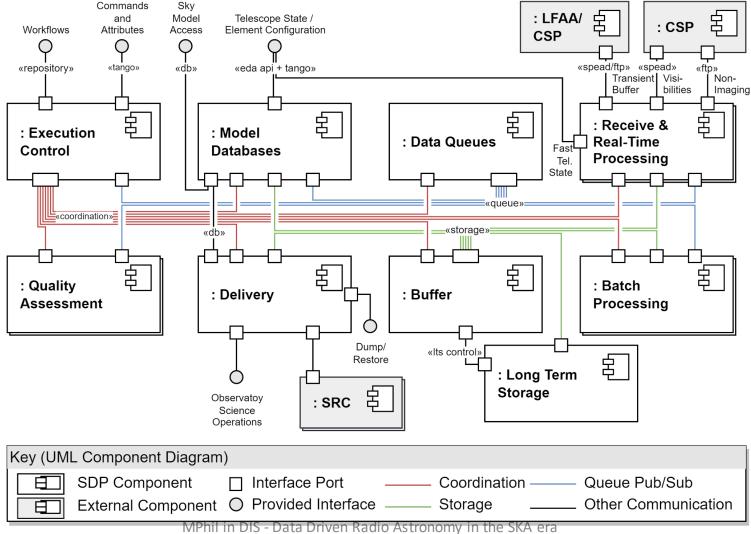


SKA system overview



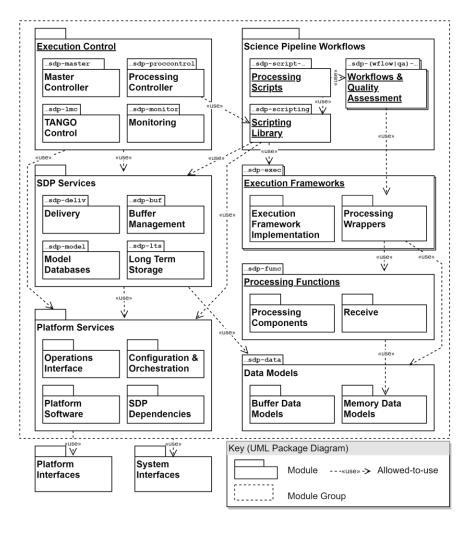


SKA Science Data Processor (SDP)





SKA Science Data Processor (SDP)





The problem of scalability

Storage

> 700 Pb of data archived peryear

Compute processing

- Typical complexity ~ 10 Pflops/s, mostly for calibration and self-calibration
- o Compute node storage limited to < 1 Tb cannot keep data in memory for more than a few s
- Images will be 10s of Tb we need to process these in memory for self-cal. Remember every visibility affects every image pixel and vice versa. We need to go from visibilities < -- FT --> images multiple times for self-calibration, while doing the calibration, for every observation.

I/O

- Continuous stream of 0.4 Tb/s for calibration and imaging
- 35 Pb/day expected throughput of the SKA -- \$10m a day
- Solution?



The problem of scalability

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Solution?

Distributed computing



The problem of scalability

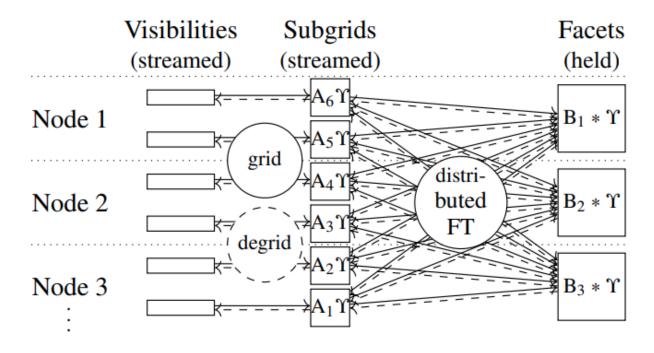
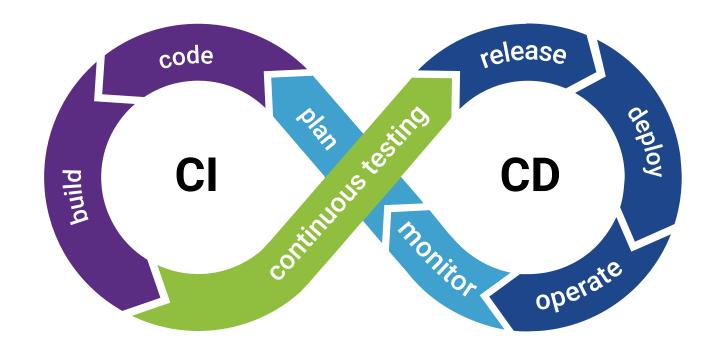


Fig. 1: Distribution Concept sketch – grid visibilities to sub-grids, accumulate contributions to facets using our algorithm (dashed: backwards direction – extract contributions from facets, then degrid visibilities from sub-grids)

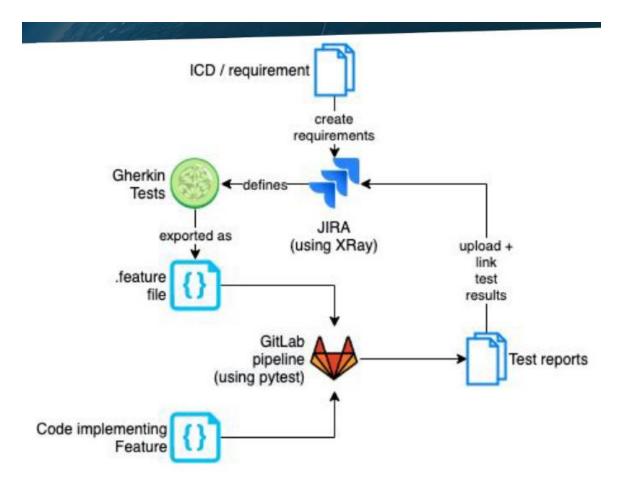
Wortmann+21

SKA Science Data Processor practices: CAMBRID



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SKA Science Data Processor practices: BDD



SKA Science Data Processor practices: Containerization

- Containers are a manifestation of a collection of features of a Linux kernel and the OS, typically launched by an Engine such as Singularity or Docker
- Allows you to have software, environments, applications stored in a single container, so that dependencies and versions are controlled faster for running calibration pipelines that use a lot of software.

SKA Science Data Processor practices: Unit/integration tests

- Unit tests test the functionality of a small unit of a large piece of code (e.g. a Python function that reads visibilities)
- Fast, agile code development requires such testing to ensure the overall codebase is clean and compiles whenever changes are made (e.g. changing calibration methods)
- Integration tests test the whole software module created, ensuring the system is compliant against functional requirement
- SKA SDP functions are unit and integration tested before they are released as working code.