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ABoM SSES

Applying *in situ* regressed SSES to *in situ* regressed SST Retrievals over the Australian Region

<http://imos.org.au/sstproducts.html>

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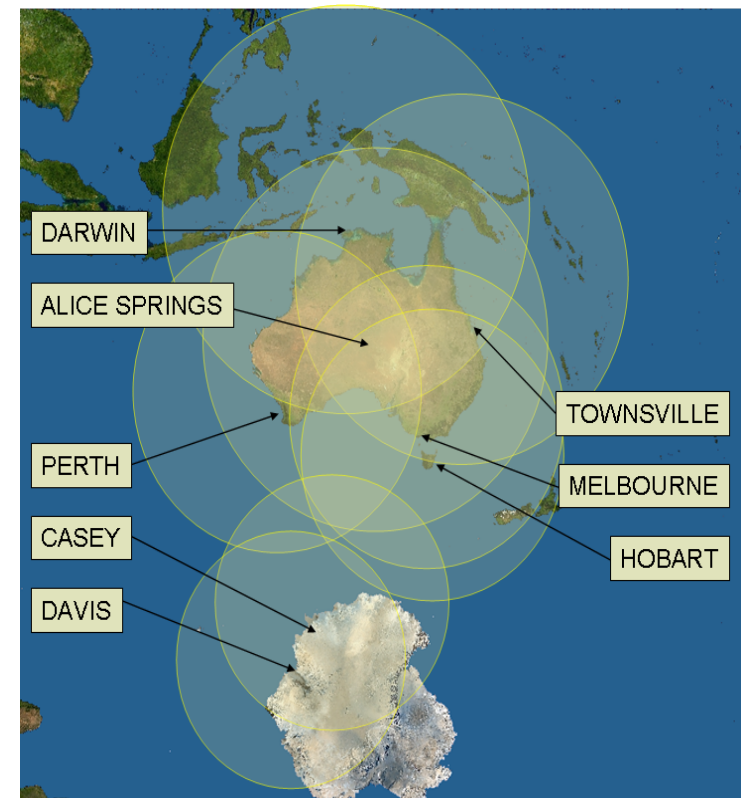


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Background

- ABOM provides High resolution SST from NOAA AVHRR missions.
- Current archives give good time coverage from April 1992.
NOAA-11,12,14,15,16,17,18,19
- Retrievals are performed based on regression to *in situ* measurements over a *long* time span.
- Error Statistics reflect the residuals of this regression based on possible sources of systematic bias over a *short* time span.
- The quality and quantity of *in situ* measurements varies over time





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Context / Features

- Field of View (**Swath**) and **Geographic** variation, decoupled and smooth.
- Sensitive to *short* term variation implies:
 - Adaptive, updated frequently
 - Time based tracking of Error statistics
 - Time based weighting of measurements
- Great variation in quality and number of measurements:
 - Favour robust measures
 - Use as many measurements as possible



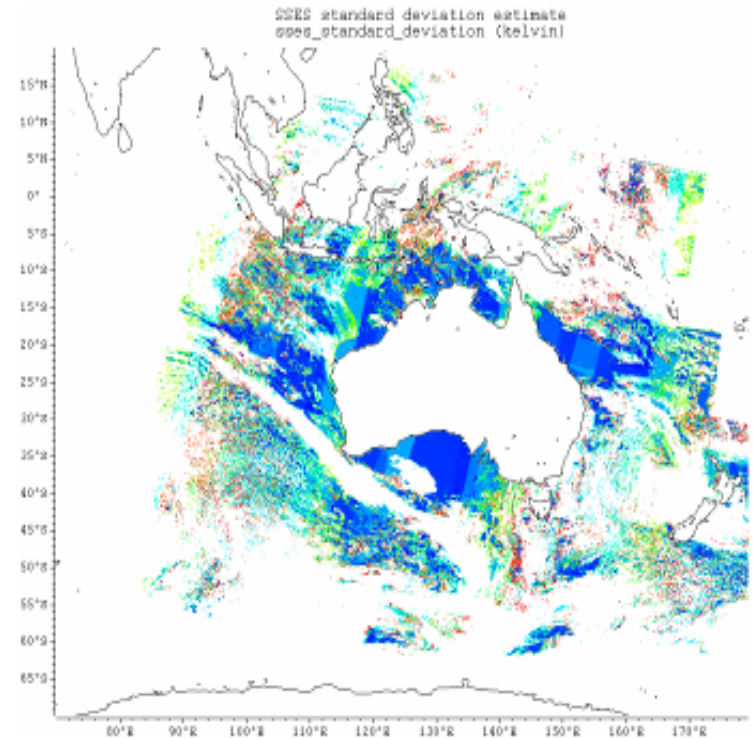


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Adaptive Error Statistics

- Per platform basis
- Rolling 1 year window adjusted frequently (every 1 to 6 days)
- Measurements are weighted by time (120 day time constant)
- Attributes considered (6-dimensions)
 - time of day,
 - satellite zenith angle,
 - quality level,
 - latitude, longitude, age





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Adaptive Error Statistics

- Includes "degrees of freedom" model
- Model for bias as a median
- Model for standard deviation

Empirical Equation

Geographic component

Swath component

$$n = n_{\text{swath}} g_n$$

$$\mu = \mu_{\text{swath}} + g_\mu$$

$$\sigma = \sigma_{\text{swath}} g_\sigma$$

- Least squares regression to
highly correlated components
- Continuous



Adaptive Error Statistics

- Resulting model has fewer degrees of freedom than "binning" method (66 vs. 144)

Empirical Equation

$$n = n_{\text{swath}} g_n$$

$$\mu = \mu_{\text{swath}} + g_\mu$$

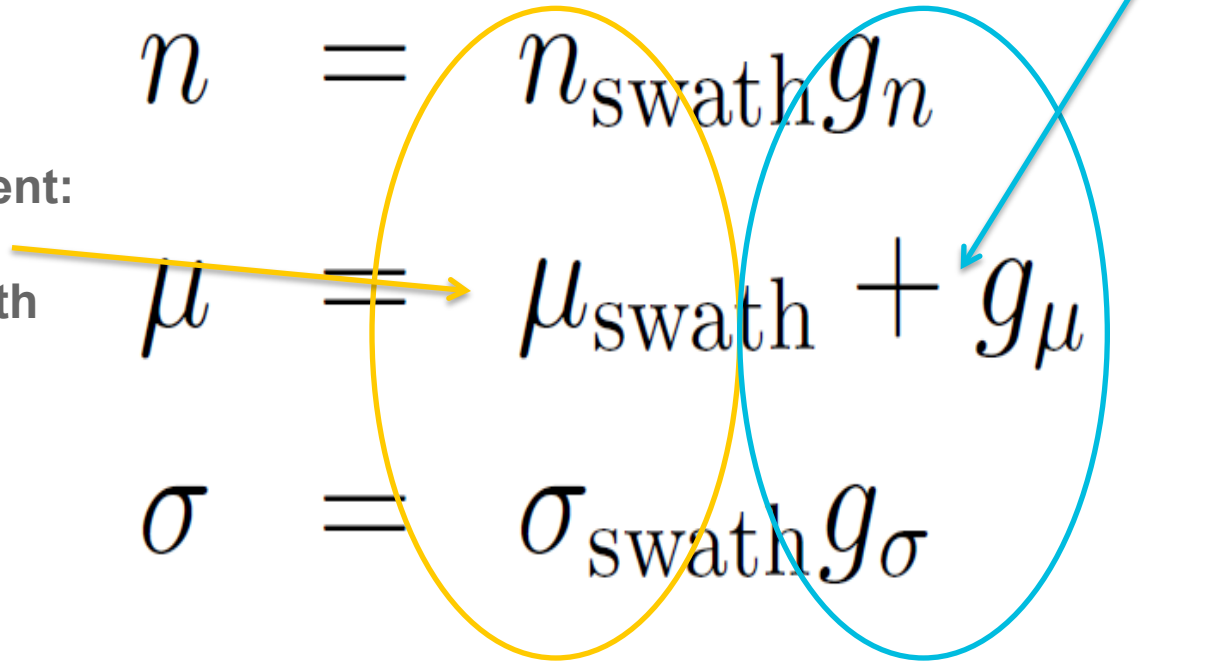
$$\sigma = \sigma_{\text{swath}} g_\sigma$$

Geographic Component:

- Latitude
- Longitude
- Time
- Quality Level

Swath Component:

- Satellite Zenith Angle
- Time of day
- Quality Level



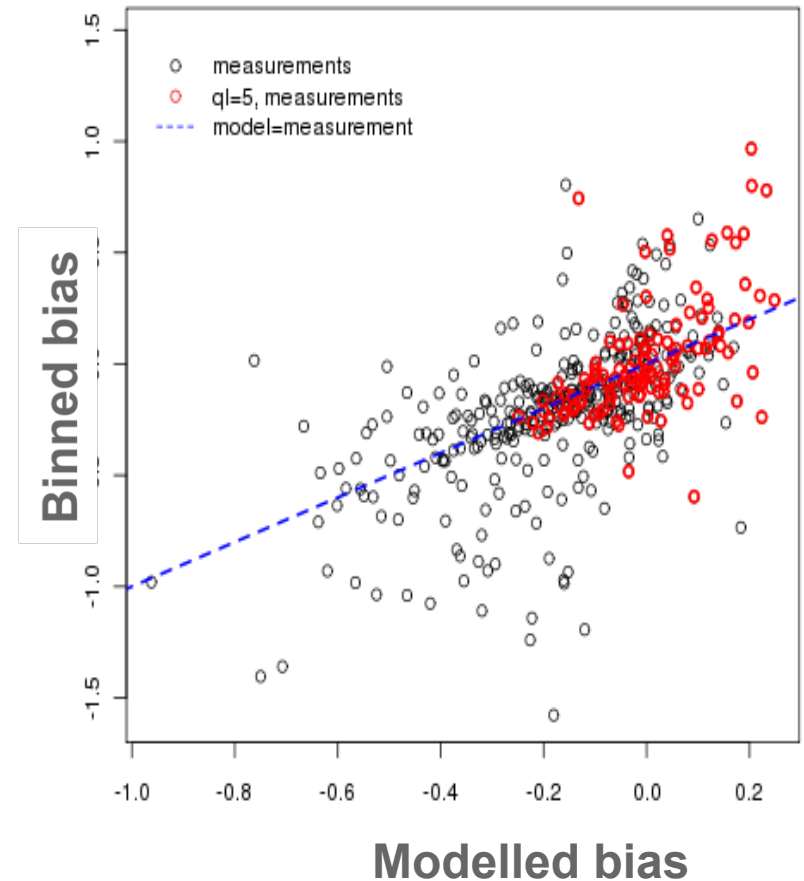
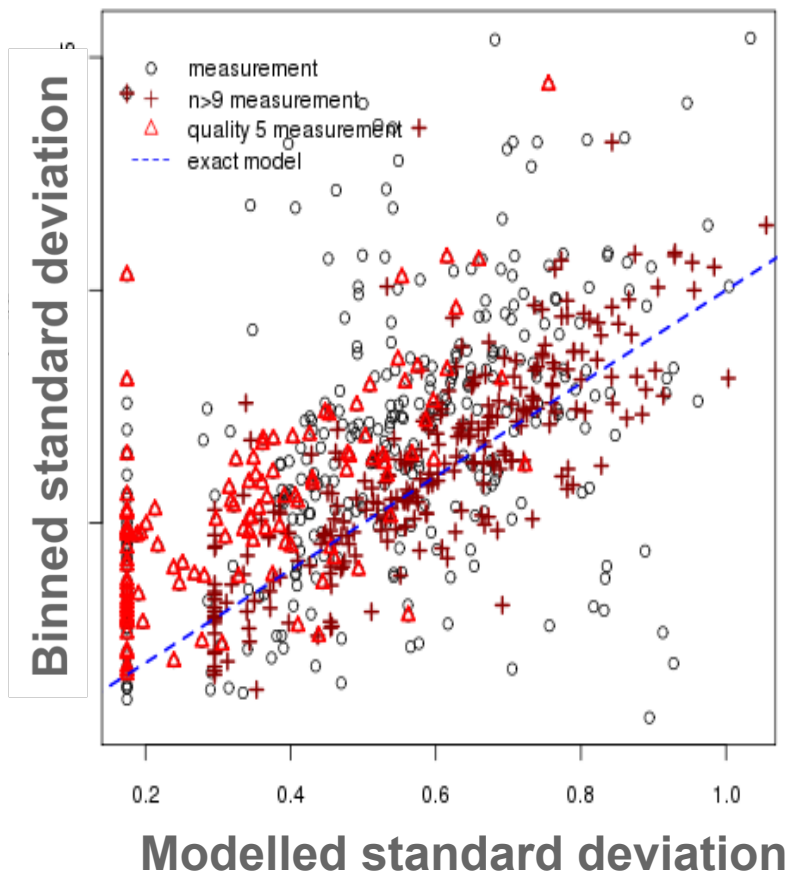


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Comparison with binned data

- Modelled parameters considered against binned *in situ* measurements



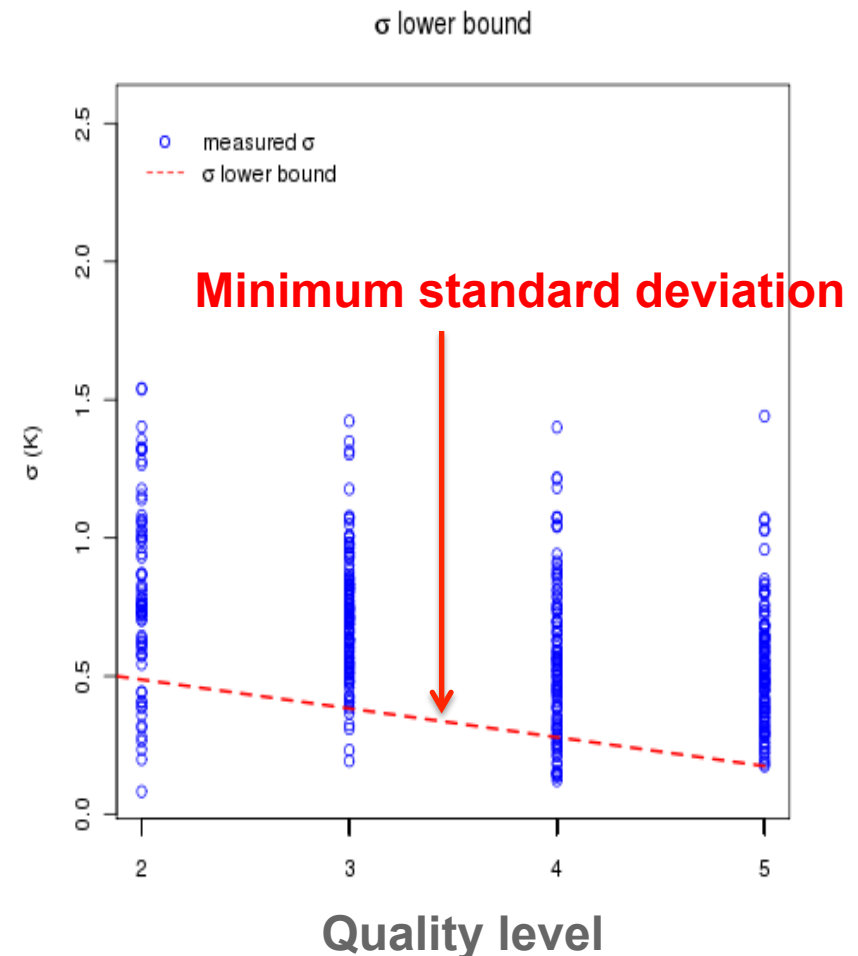


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Keeping small variations in check

- Data is binned so that medians and standard deviations can be determined.
- Ensure that the range of a computed standard deviation is *reasonable* by computing the minimum acceptable value
- Limit is based on view model, but is applied separately for view model and overall standard deviation.

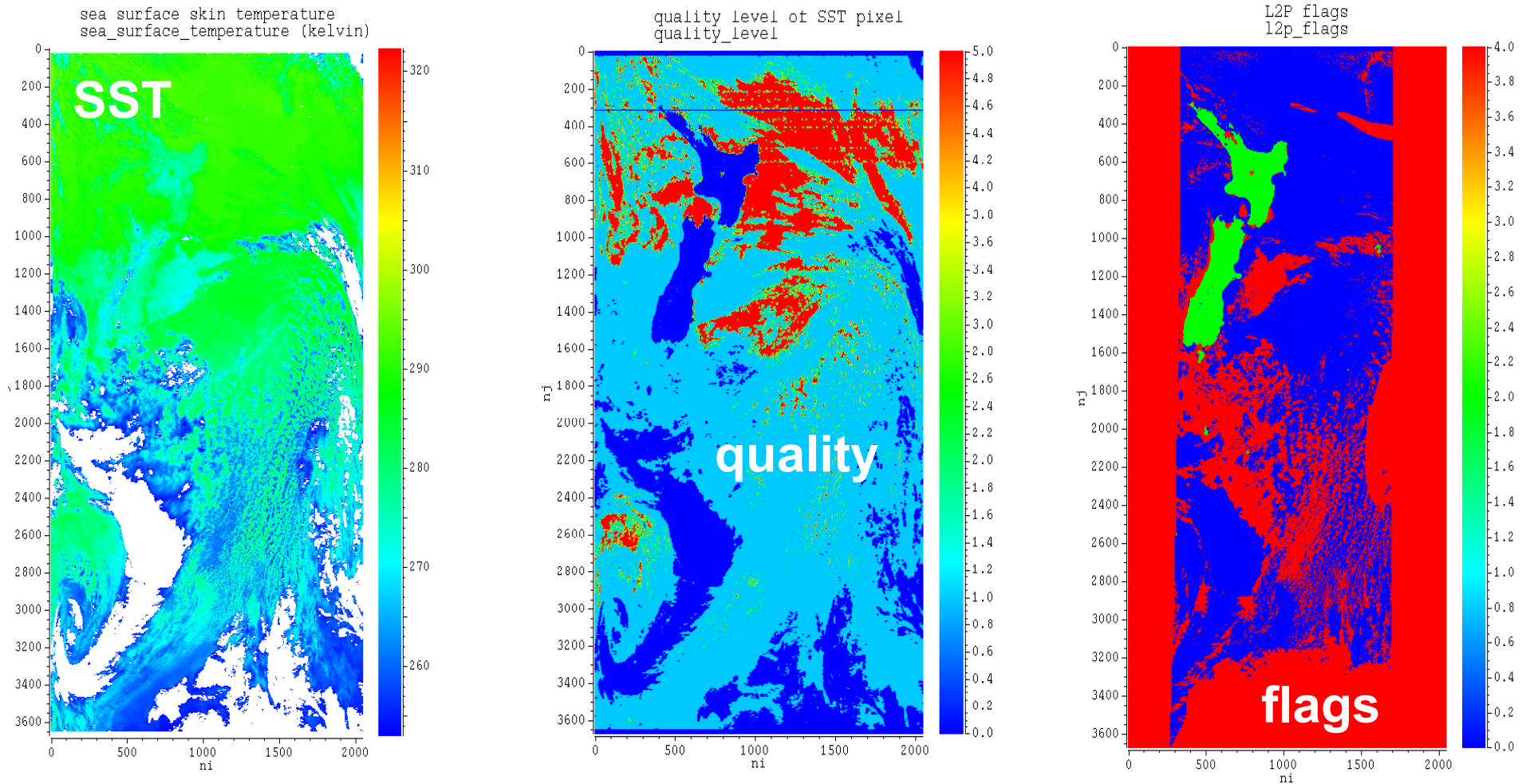




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Applied to a swath



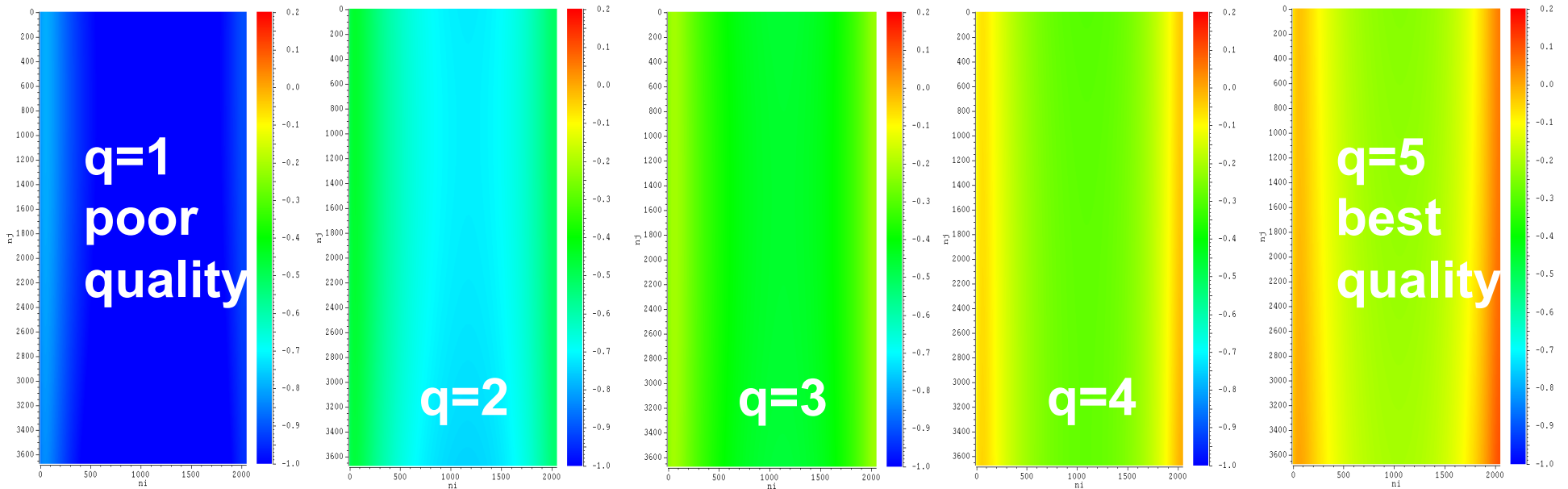


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Smoothly varying bias

- Smoothly takes care of edge of swath
- Asymmetry from geographical factors



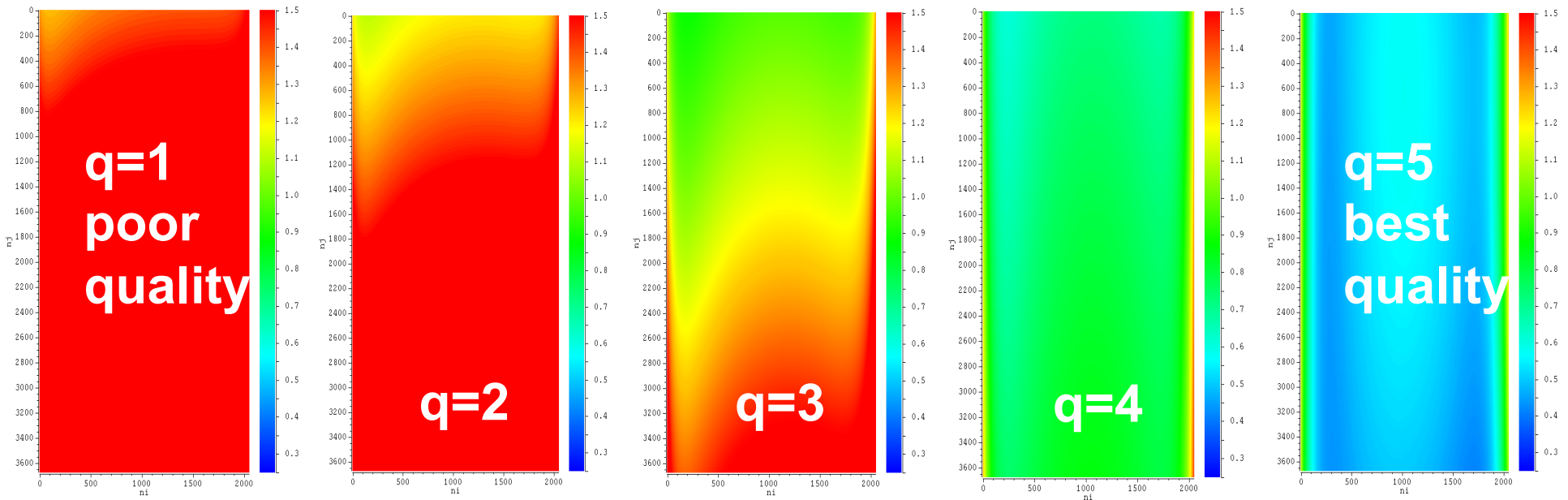


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Smoothly varying standard deviation

- Smoothly takes care of edge of swath
- Asymmetry from geographical factors





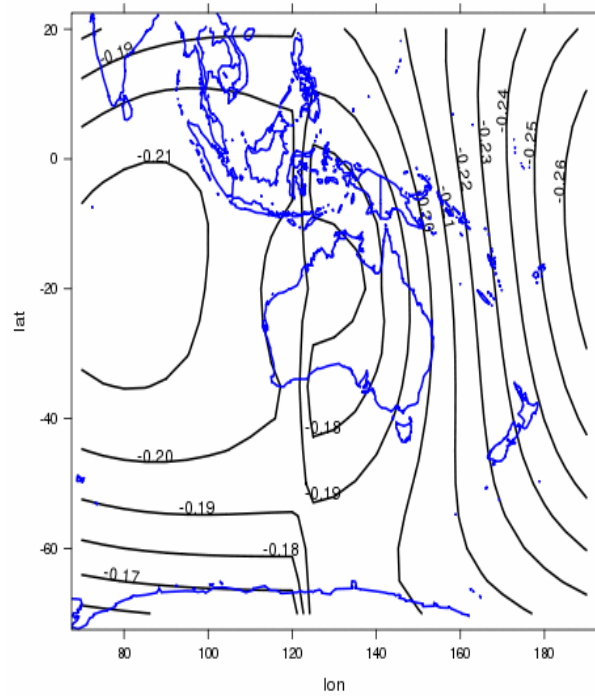
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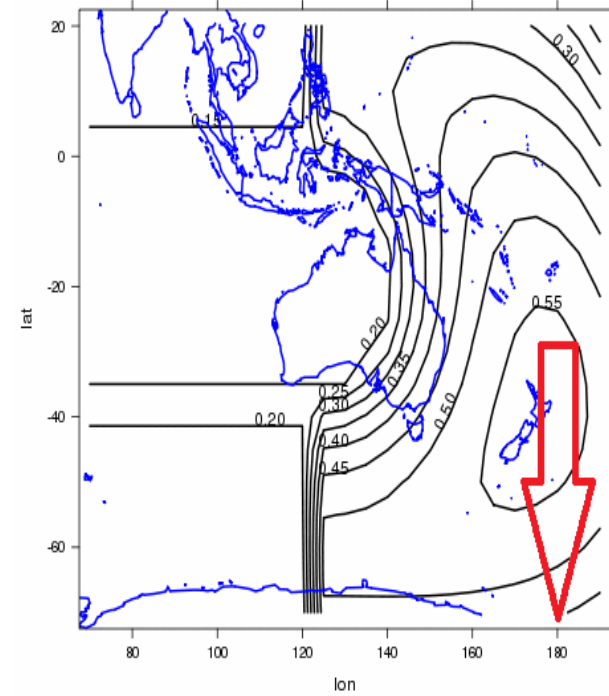
Applied Geographically

q=5, nadir

Bias



Standard deviation



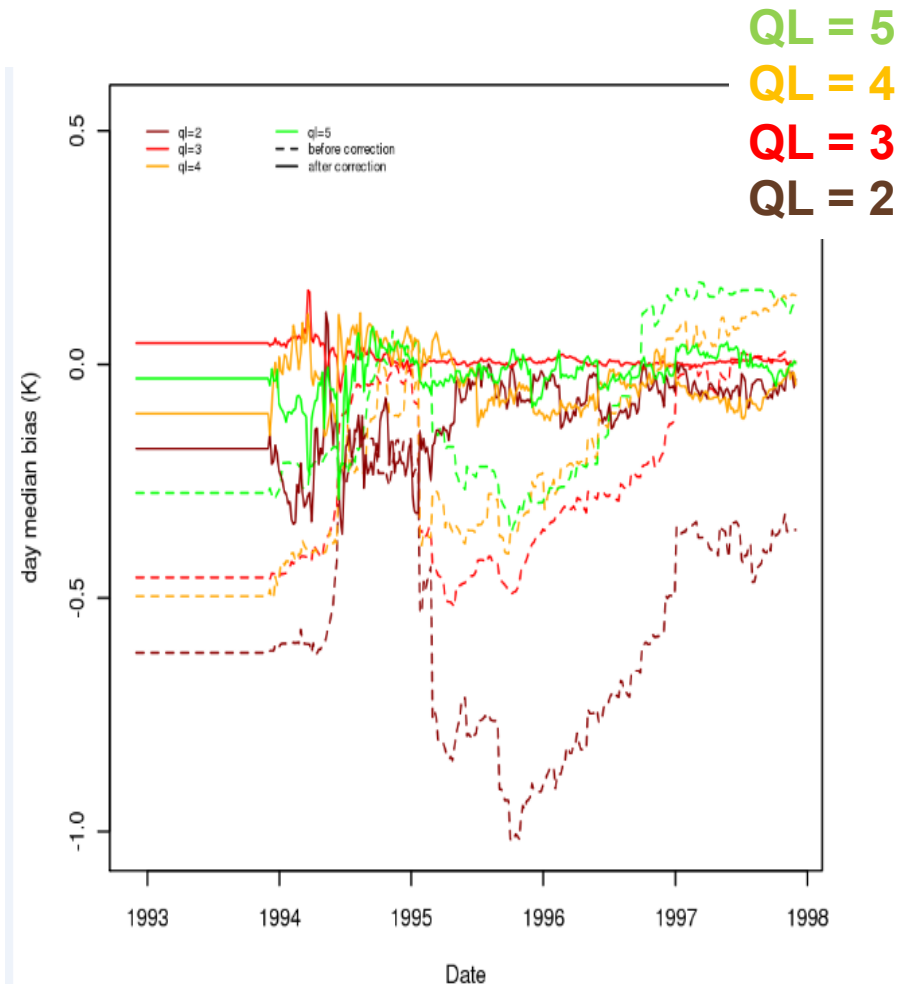


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Bias estimate performance

- Applying the bias correction improves the bias compared with *in situ* SST at all quality levels
- Dashed lines show before bias correction
- Annual performance is shown to the right. 60 day performance also shows an improvement (although the fluctuations are more erratic)



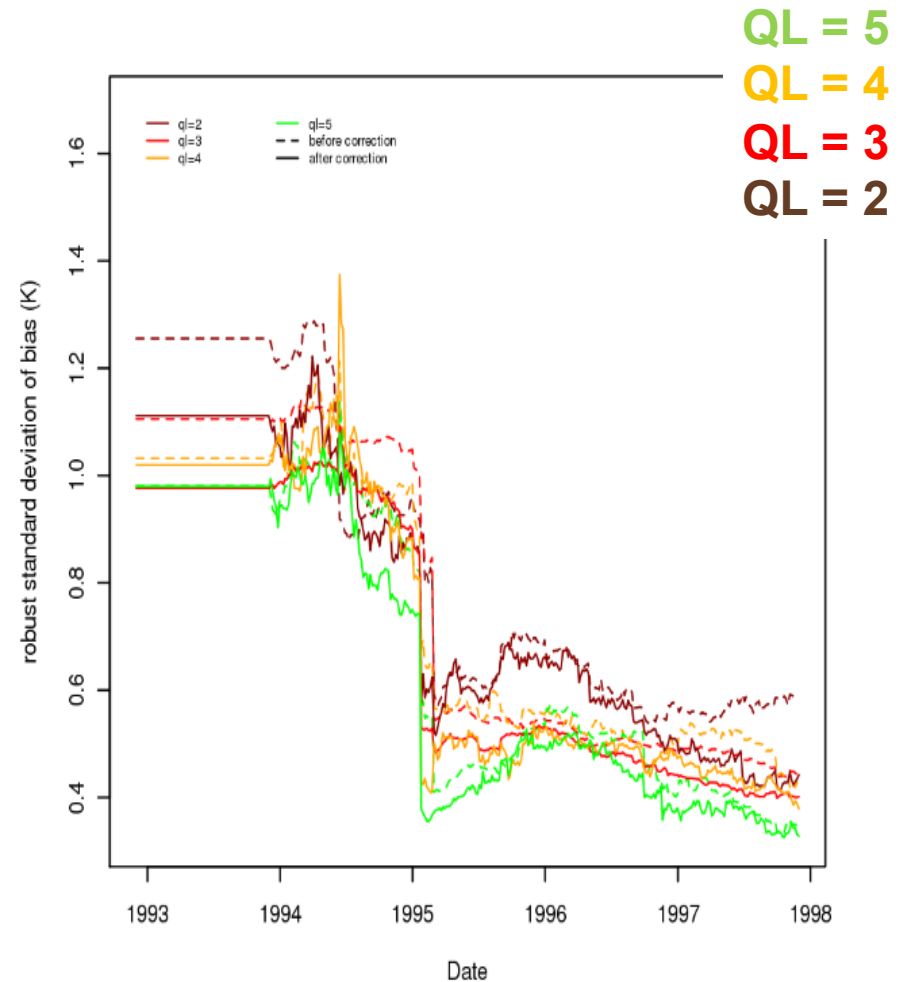


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Bias estimate performance

- Applying the bias correction improves the rsd compared with *in situ* SST at all quality levels
- Dashed lines show before bias correction
- Annual performance is shown to the right. 60 day performance also shows an improvement (although the fluctuations are more erratic)



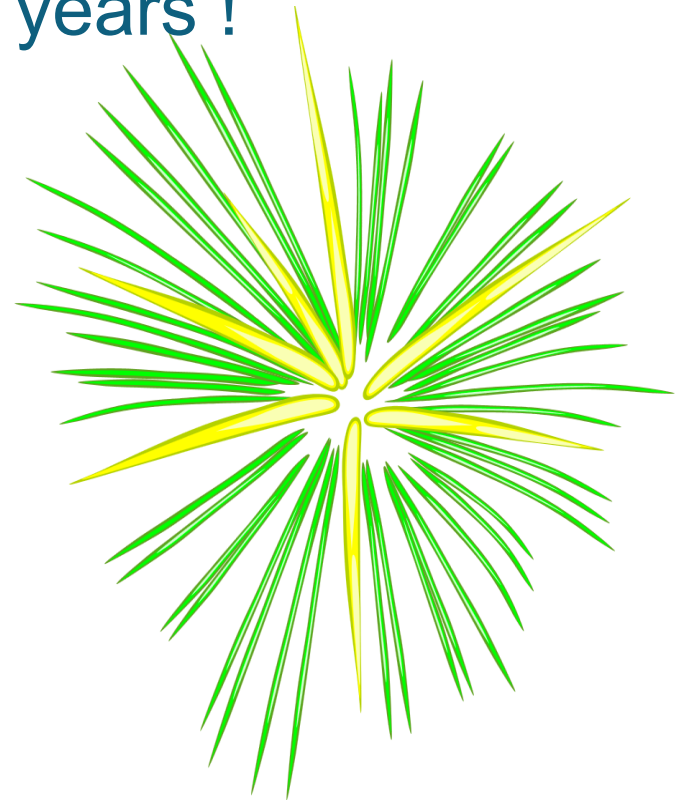


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v2.0 long term data set

Available soon in v2.0 L2P/L3U/L3C/L3S
over all platforms, and 21 years !



Thank you !

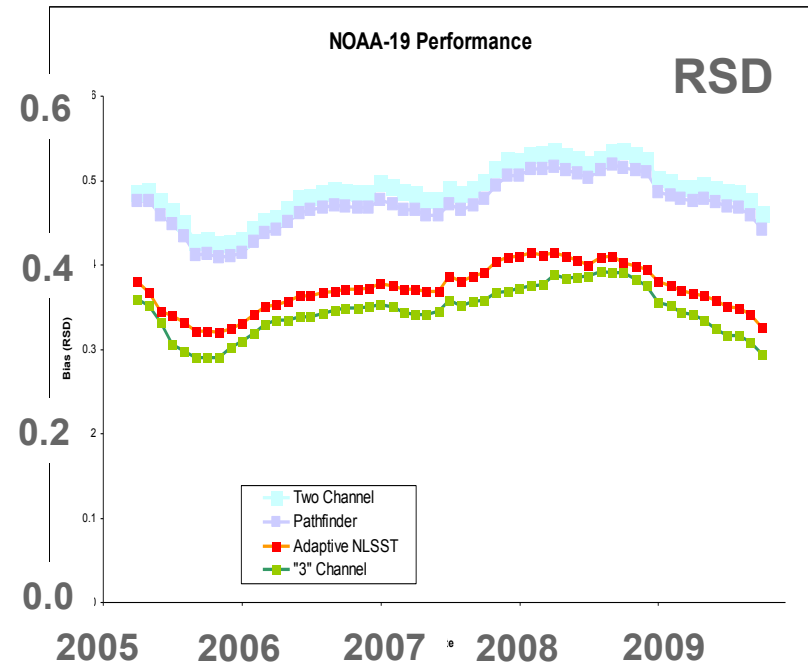
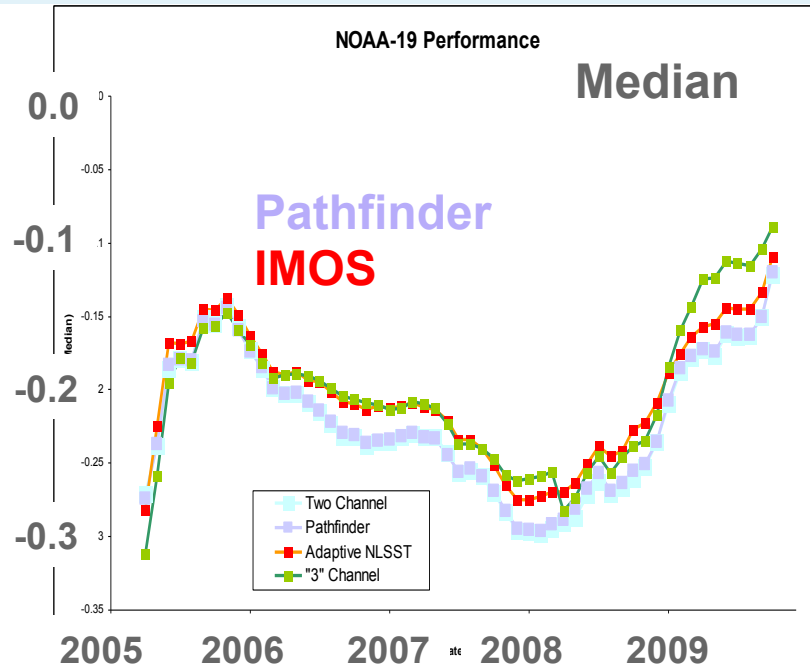
Additional slides for discussion



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Adaptive Calibration



- Running 1 year calibration window, adjusted monthly
- Tuned on best matchups with in situ SST
- Performance measured on an expanded matchup data set

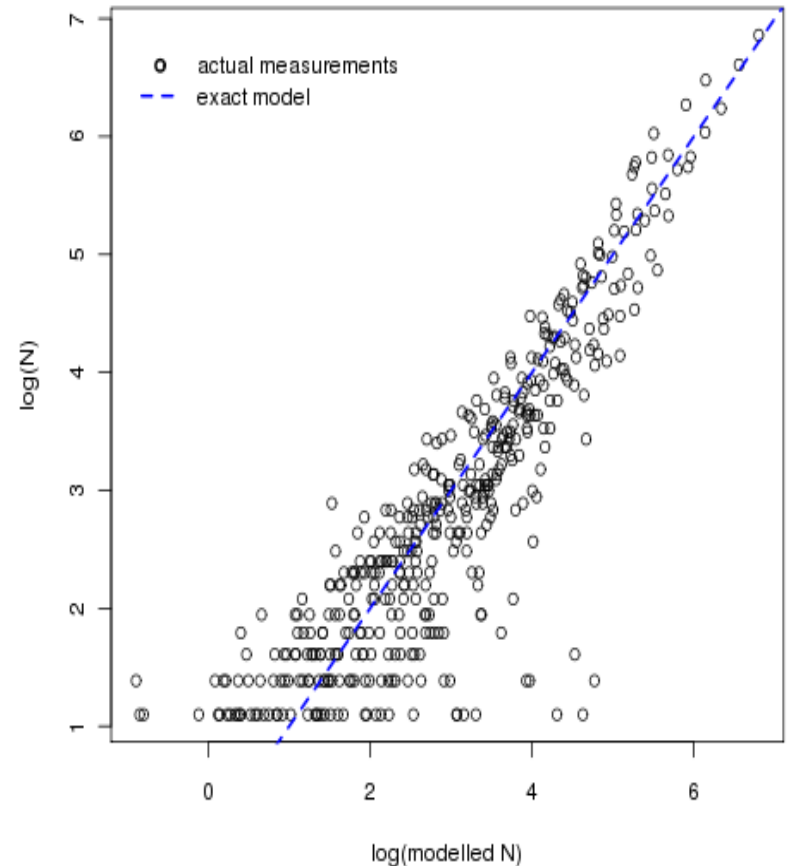


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Adaptive Calibration

- Degrees of freedom modelling provides an estimate of the number of in situ measurements that went into the analysis as a function of the model degrees of freedom.

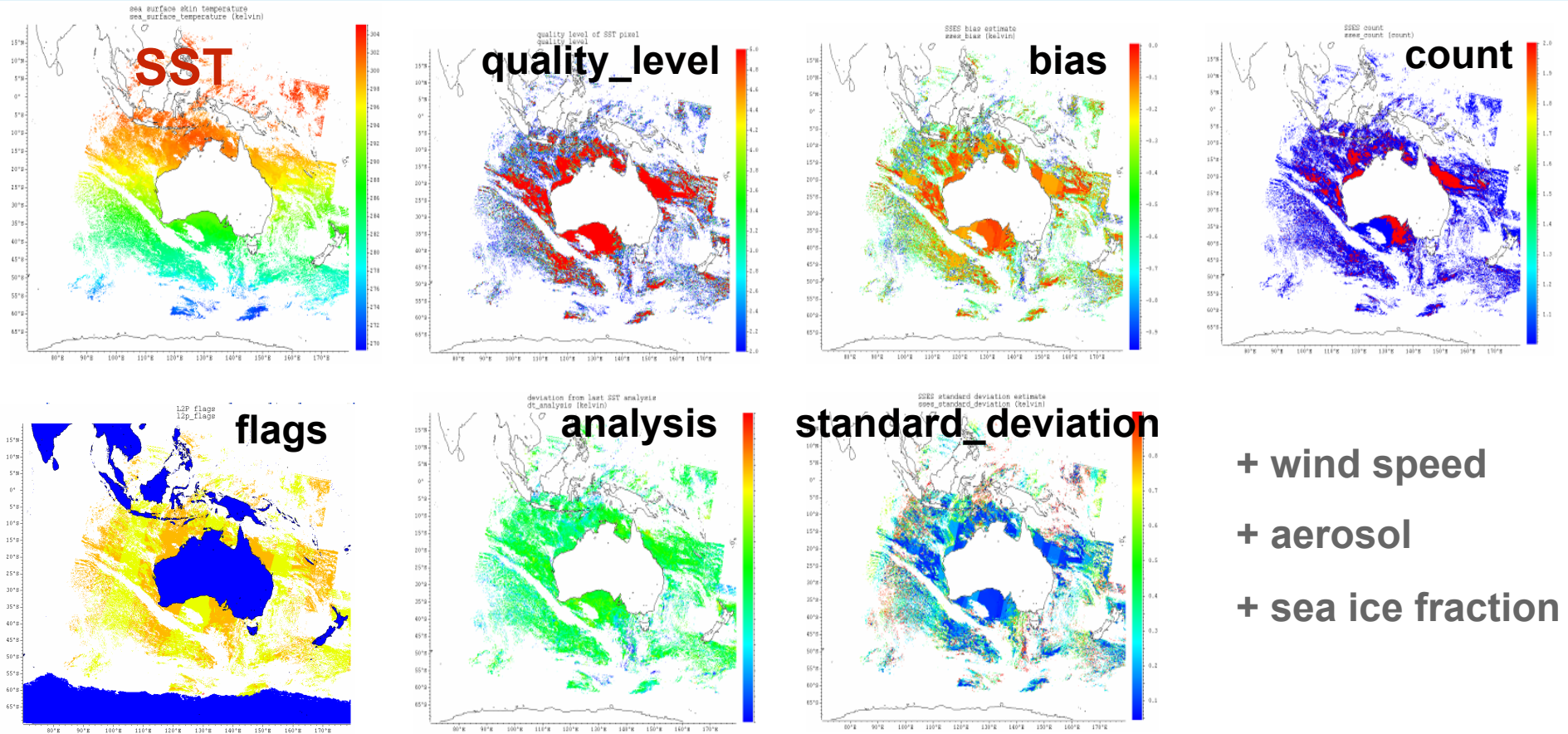




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GHRSSST 2.0 compliant format



- + wind speed
- + aerosol
- + sea ice fraction

20131007 night composite from multiple satellites "L3S"



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Access to Australian domain data:

<ftp://aodaac2-cbr.act.csiro.au/imos/GHRSST>

Access to pre-release Antarctic domain data:

Contact h.beggs@bom.gov.au or ghrsst@bom.gov.au

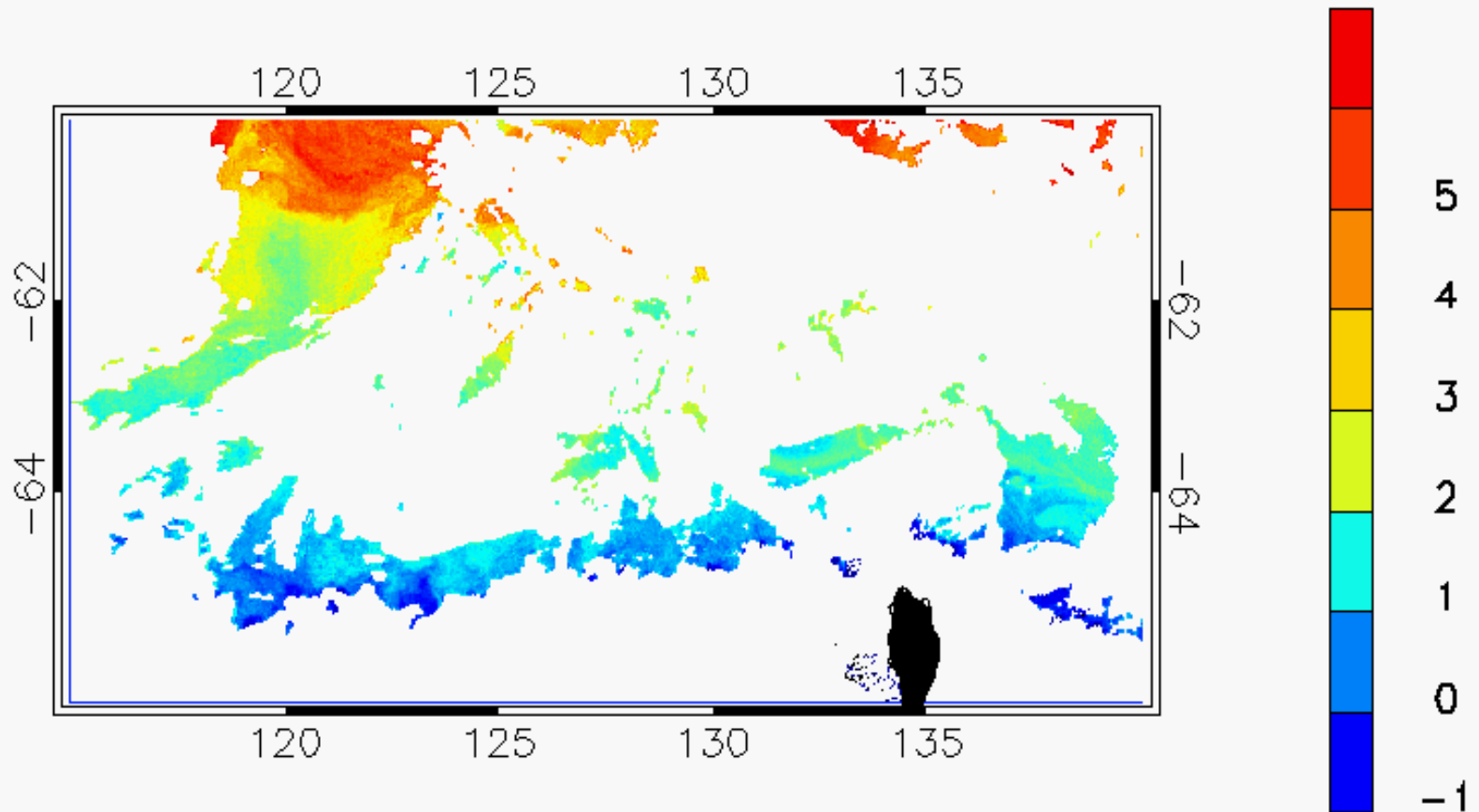
Happy to get Beta users and feedback!

Further information:

<http://imos.org.au/sstproducts.html>

Day+Night 1-day L3S

18 Feb 2014





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“open loop” retrieval

- No Analysis fields in calibration
No Radiative Transfer Model
No Background SST
In situ measurements pre-selected
Non-linear Retrieval (MCSST, NLSST like)

$$\text{Night: } SST_{skin} = a_0 T_4 + a_1 T_3 (T_3 - T_5) + a_2 (\sec \theta - 1) + a_3$$

$$\text{Day: } SST_{skin} = b_0 T_4 + b_1 T_4 (T_4 - T_5) + b_2 (T_4 - T_5) (\sec \theta - 1) + b_3$$

- Monitor:
Performance compared with Analysis
Residual error from fit
Sensitivity (use *in situ* measurements)
Propagation of sensor errors
per **GHRSSST** specifications



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Composition which respects bias

Composition of different sensors

- Consider all sources of measurement weighted by the count
- Biases are adjusted before measurements are combined.
- The combined count is recorded.

Composition from the same sensor

- When combining, consider all sources of measurement weighted by $\left(\frac{n}{\sigma^2}\right)$
- Biases are estimated by weighting.
- The combined count is recorded.



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“3” channel Day/Night performance

		2 channel linear	2/3 channel linear	2/3 channel NLSST	“3” channel
Model complexity		3 terms Includes θ_z	Day / Night 2 equations	Day / Night 2 equations	“3” terms
μ	Day	0.09 K	0 K	0 K	0.10 K
	Night	-0.07 K	0 K	0 K	-0.08 K
	Both	0 K	0 K	0 K	0 K
σ	Day	0.56 K	0.56 K	0.56 K	0.54 K
	Night	0.58 K	0.44 K	0.42 K	0.43 K
	Both	0.58 K	0.50 K	0.48 K	0.49 K
median	Day	0.05 K	-0.05 K	-0.05 K	0.06 K
	Night	-0.11 K	-0.05 K	-0.05 K	-0.14 K
	Both	-0.04 K	-0.05 K	-0.05 K	-0.07 K
rsd	Day	0.30 K	0.30 K	0.29 K	0.28 K
	Night	0.30 K	0.21 K	0.20 K	0.20 K
	Both	0.31 K	0.24 K	0.23 K	0.25 K