

BoM Himawari-8 AHI Cloud/SST/SSES Algorithms

Chris Griffin

Bureau of Meteorology, Melbourne, Australia

NOAA-BoM SST Workshop,
College Park, MD, USA,

3rd to 6th April 2018



Introduction

- BoM is moving towards provision of Himawari-8 (H08) products on the IMOS Australian regional grid.
 (July 2018 committed release date, near real-time, for research)
- H08 products are currently *as originally released* and largely unchanged from mid 2015.
- Target products (on IMOS domain):
 - L3U 10 mins (probably not generally available due to data volume)
 - L3C hourly, L3C before sunrise daily (to merge with L3S)
- Long standing "problems".
 - Estimating uncertainties
 - Clouds
 - Time trends

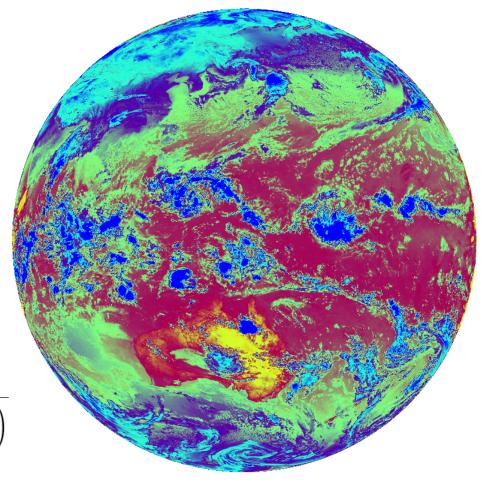


 $q_s = \lfloor 5 \exp^{\eta q_{\text{sses}}} \rfloor$

H08 GHRSST products – Current Status

- First cut VIIRS cross tuned H08 product (single epoch of calibration, 20150721)
- Clouds from GEOCAT, SEVIRI cloud properties algorithm.
- Errors estimated based on "local SST variability"
- Quality based on distance assessment to cloud and size of estimated error.

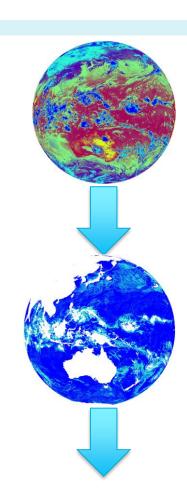
$$q_{\rm sses} = \frac{1}{\sqrt{2}} \sqrt{\max\left(\left(\frac{\sigma_{\rm sses}}{\sigma_0}\right)^2 + \left(\frac{\mu_{\rm sses} - \mu_0}{\sigma_{\rm sses}}\right)^2 - 1, 0\right)}$$





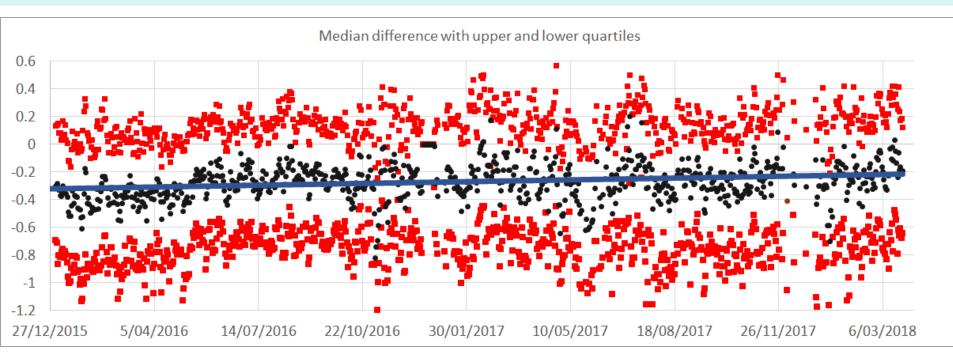
A list of things that are desirable

- Open loop characterization. Applied iteratively.
- Quality determination is potentially from a disparate source to sses error estimate – considered non-parametric.
- Highest spatial resolution possible, even if it means that we have *noise*.
- Exploit temporal "correlations" or persistence, to guide selection of measurement.
- Consistency of approach between GEO/POLAR.
- Having a measurable means to identify how closely we meet objectives.





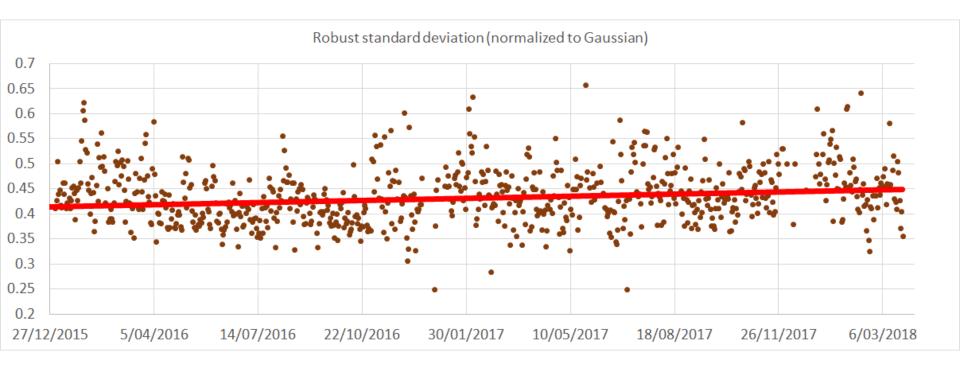
How have things been tracking?



- H08 vs DRIFT+TROPICAL MOORINGS, DAILY VERIFICATION
- Median Bias has shifted (expect cool skin).
- Cold tail is obvious.
- Retuning and/or dynamic bias correction required.



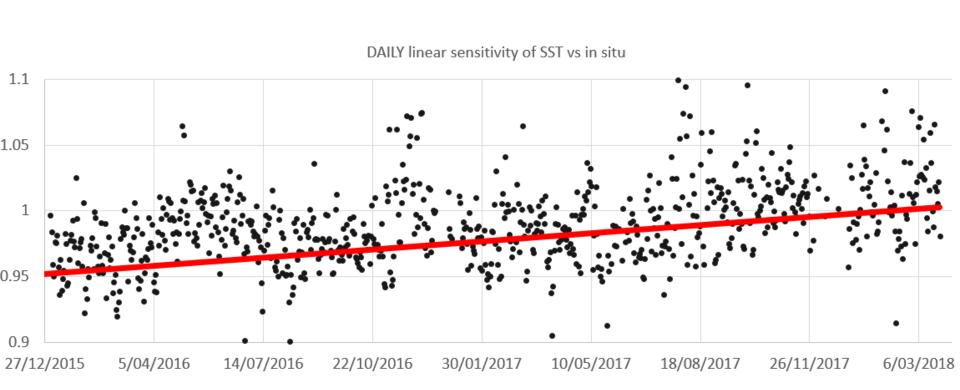
How have things been tracking?



- Points to need to better bias correction on sub-groups of SST retrievals
- Points to need to include retuning.



How have things been tracking?

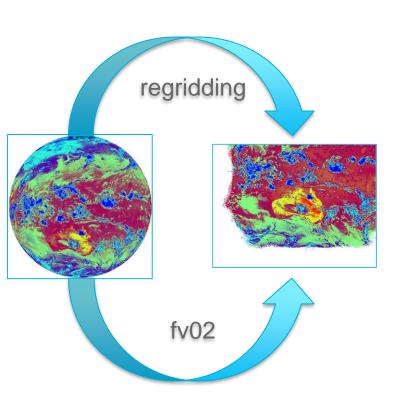


- Same comments as before.
- On average, the linearity / sensitivity looks really good.



What we need to do!

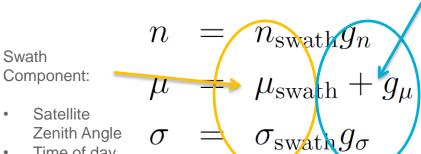
- Consistent Error model on existing L2P
 - Implement fv02 error model.
- Consistent regridding
 - What L2P information is available at a given L3U pixel, and how should it be weighted?
- Identifying misclassified SST cloud.
- Working with temporal information.
- Validation data produced in parallel. (probably on delayed near real time)





Consistent error model

- Bias and standard deviation evaluated on 5 dimensional fv02 model over a rolling time period.
- Ensure time of day information degrees of freedom are enlarged for H08.
- *Tuned* against buoys.
- Independent of regressors and L4
 - don't try to milk too much from the data
- Standard deviation should be subject to two lower limit(s)
 - NFAT
 - $\sigma(ISAR buoys)$



Quality Level

Time of day

Geographic Component:

> Latitude Longitude

- Time
- **Quality Level**



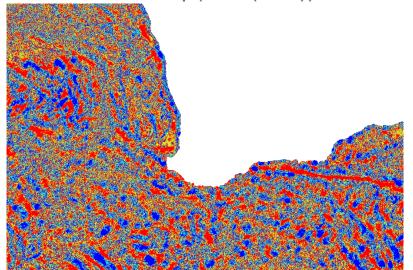
Consistent re-gridding

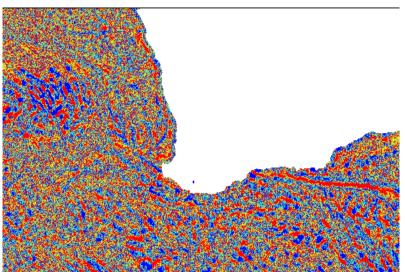
Verification of re-gridding method:

- With all pixels at the same quality level, measure
 - relative spatial correlation length (SCL) ?
 - image gradient ?, laplacian ?

remap(SCL(L2P))

SCL(remap(L2P))





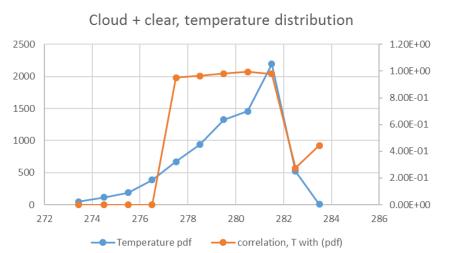


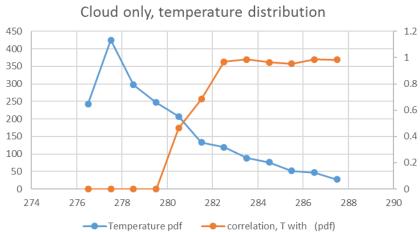
- By *cloud* we mean atmospheric interference that prevents a surface temperature observation.
- A few ideas about what might help.
 - Spatial scales
 - 500m scale for formation of features
 - 20km scale for decorrelation of effects such as precipitation
 - 1000km scale of maximum extent
 - Multifractal shape properties.
 - Distribution based clustering.
 - Planar clustering.
 - Blob classification.
 - Temporal classification.



Distribution based classification

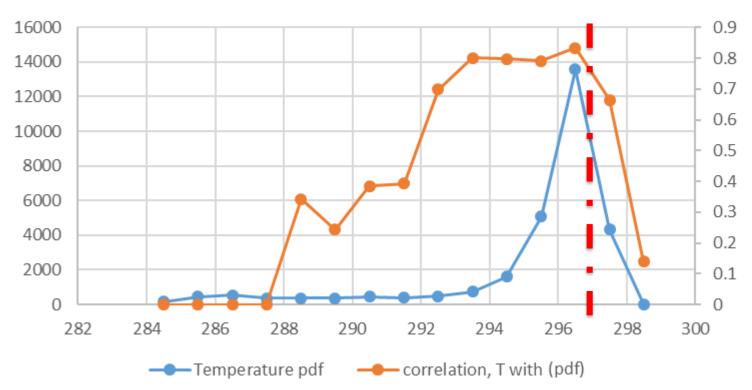
- Cloud temperature distributions have *fat* and even *linear* tails
- May aid in determination of edge temperatures for cloud.
- Need to be aware of this if using classification algorithms which assume Gaussian distributions.







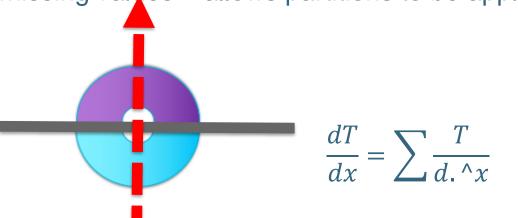
Cloud + clear, temperature distribution





Blob classification

- Classification of similar pixels based on stable SST values (edge partitioned gradients)
- Use edge partition algorithm that includes spatial scale and is robust to missing values – allows partitions to be applied on different scales.

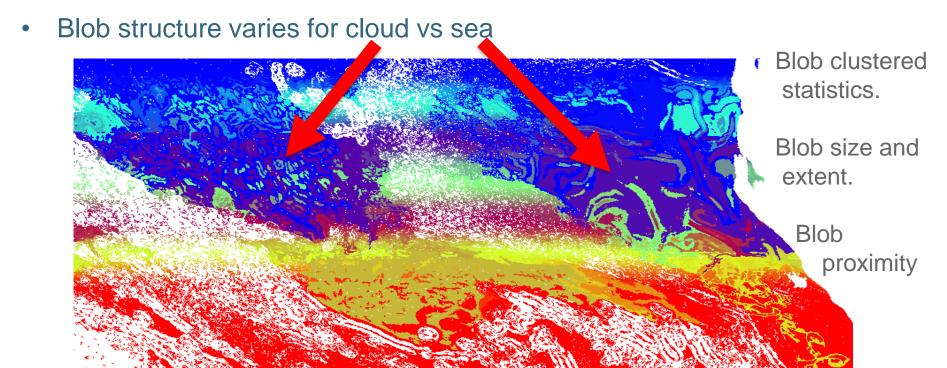


Doesn't identify cloud, but does suggest that all pixels in a blob be treated the same way!



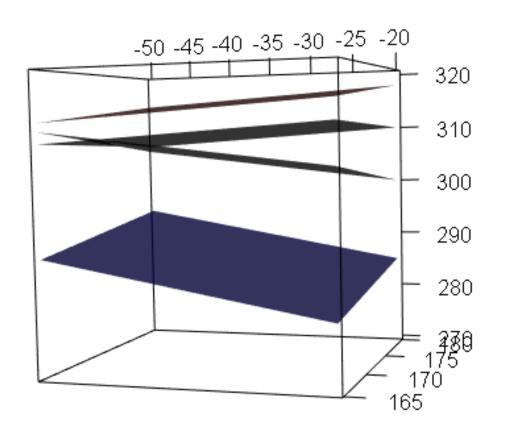
Blob classification

- Small blobs represent unstable regions can be easily identified.
- Larger blobs are stable regions (not necessarily cloud or surface)



Planar classification

- Similar to k-means clustering.
- Allows large scale structures to be identified.
- Fixed computational complexity independent of scale.
- Scales of 1,000,000km^2 are probably practical.
- Needs some *tricks* to ensure convergence.
 - Exploit cloud distribution
 - Exploit blob properties





Time series classification

- 10min is a very short time scale compared to most SST phenomena
- Simple time based difference
 - Pixels that become suddenly cold are cloudy (not assumed to be true of the reverse)
- Robust trend can be extracted from data and forward interpolated.
 (allows trended temporal cloud edge detection)
 - Refitting planar classification with time as an additional parameter.
- Blob flow
 - Consistency of trending in blob located regions.
- Optical flow
 - Identify fast flowing regions.



Methodology

- Find a bunch of things that might work
 (As discussed, a largely unfinished exercise at the moment)
- Fit them to verified cloud observations, physical retrievals of cloud properties, etc.

(Fit with traditional or trendy methods)

 Predictor is probability of cloud (Independent verification data)



Thank You!

Contact: Christopher.griffin@bom.gov.au