

Ocean Colour Validation Report 2016‐17

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Cover Images

Snapshots of the Australian Ocean Data Network (AODN) Portal displaying two subsequent MODIS‐Aqua orbits of GSM chlorophyll‐a around Australia (left) and the corresponding spatial distribution of IMOS Bio‐ Optical Data Base measurements.

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1 Introduction

In this second IMOS Ocean Colour validation report we provide an update on the accuracy assessment of four different chlorophyll‐a (chl‐a) products, namely *OC3* (O'Reilly et al., 2000), *OCI* (Hu et al, 2012, Wang and Son, 2016), *GSM* (Maritorena et al., 2002) and *Carder* (Carder et al., 1999, 2003), that are computed and distributed by the IMOS Ocean Colour Sub‐Facility.

Product validation is achieved through match‐up analysis, comparing the satellite derived chl‐a products to *in situ* observed chl-a measurements close to the satellite overpasses within the wider Australasian marine region ([10°N,80°E]‐[60°S,180°E]).

In situ data for validation however are not available in every region and some marine areas around Australia (e.g. the Great Australian Bight, Tasman Sea or the Gulf of Carpentaria) are sparsely or not at all covered by ground observations. To enable product validation in these under‐sampled regions, the Ocean Colour Sub‐Facility has adopted a validation approach that is based on a classification of Optical Water Types (OWT, Moore et. al 2009). This approach assumes that the match‐up results obtained for a given water type can be used to estimate the accuracy of a specific Ocean Colour product in the absence of ground observations with the help of a corresponding satellite‐derived water type map. Water type maps are produced by IMOS as a separate ocean colour product to guide this accuracy interpretation.

The *in situ* chl-a data for this comparison were extracted from the IMOS Bio-optical Data Base, which collates in‐situ discrete physical, bio‐geochemical, and optical data collected by the by the Australian bio‐ optical community from 1997 to date.

Significant additions to the Bio‐optical Data Base were achieved during the 2016‐17 reporting period and specifically the large data sets provided by the Australian Institute of Marine Science (AIMS) and James Cook University (JCU) for the Great Barrier Reef region, led to a significant increase in the number of valid chl-a match-ups for both VIIRS and MODIS Aqua compared to the previous reporting period.

The MODIS match‐up data for the *OC3* algorithm for example, with a time difference (ΔT) of ±2 hours to the *in situ* data, more than doubled and increased from N=413 in 2015-16 to N=1,085 for this reporting period. More importantly the VIIRS match‐ups increased significantly now allowing a split into Optical Water Types (OWT), which was not possible in 2015‐16 when only N=32 match‐up were reported. For this report match‐ ups increased to N=836 for the *OC3* algorithm at ΔT=±24 h and N=510 at ΔT=±2 h.

The IMOS baseline processing system for MODIS and VIIRS remains unchanged since the last reporting period. It is based on SeaDAS version 7.3.1 and included regular updates to calibration files and Look‐up‐ Tables released by NASA.

A detailed summary of the IMOS satellite data processing system and a description of the validation methodology are provided in the 2015‐16 report (Schroeder et al., 2016).

This report provides a brief summary on the 2016-17 water type-based chl-a validation results for the Australasian marine region.

2 Results and discussion

The accuracy of satellite-derived chl-a products generated and distributed by the IMOS Ocean Colour Sub-Facility were evaluated using *in situ* chl-a observations collated by the IMOS Bio-optical Data Base activity. The comparison was performed for the full mission time series of the MODIS-Aqua and the VIIRS Suomi-NPP Ocean Colour sensors covering the Australasian marine region.

The range of *in situ* measured chl-a extracted from the Bio-optical Data Base and used in the match-up analysis covered three orders in magnitude $[0.01$ -10] mg m⁻³. Due to the significant increase in available measurements for validation, statistics became more robust for this reporting period, especially when splitting into water types. However, measurements for OWT 5, representing highly absorbing CDOM-rich waters, remain under‐sampled and were not available in sufficient number to validate algorithms for either MODIS or VIIRS.

As already noted for the last reporting period large observational gaps in *in situ* chl‐a still exist for the Tasman Sea, Gulf of Carpentaria and the Great Australian Bight.

The water type-combined match-up analysis for MODIS-Aqua showed that all algorithms overestimated chl‐a compared to the *in situ* measurements indicated by their positive bias. At ΔT=±2 h and for the OWT‐ combined match-ups the smallest bias of 0.07 mg m⁻³ was observed with the *Carder* algorithm, while OCI showed the largest bias of 0.37 mg m⁻³. When match-ups were split into OWTs only Carder showed a negative bias underestimating chl‐a for OWTs 1‐3 within the same ΔT. In terms of percentage error *Carder* performed best across all OWTs on MODIS except for OWT 6. In open ocean waters *OCI* showed a slight improvement compared to *OC3* reducing the percentage error from 97% to 86% for OWT 1. With 206% error *GSM* performed worst on MODIS for the open ocean waters represented by OWT 1. The retrieval performance was generally low for all algorithms applied to MODIS in coastal waters, e.g. across OWTs 4‐8 showing errors of up to 590%.

The application of *OC3* and *GSM* to VIIRS showed overall a decreased performance compared to their application to MODIS, with OWT‐combined *OC3* retrieval errors of 308% for VIIRS compared to 273% with MODIS, and *GSM* errors of 530% for VIIRS compared to 230% with MODIS at ΔT=±2 h. The correlation between *in situ* and VIIRS *GSM* chl-a is extremely low (R²<0.08) due to the significant overestimation at lower concentration levels <1 mg m⁻³ (Fig 5). Average *OC3* retrieval errors for OWT 1-3 were 86% while GSM errors exceeded 1,200%. While the water type-combined number of match-up for VIIRS were sufficient for this analysis (N=510 OC3, N=416 GSM, ΔT=±2 h), their split into water types remain sparse for OWT 1 with N=9 (open ocean waters), and OWT 4 with N=8 and OWT 5 with N=1 (high CDOM waters). Correlations between *in situ* and VIIRS *GSM* derived chl‐a are low across all water types. Better correlations are achieved with VIIRS *OC3* especially for water types 1, 6, 7, and 8.

The overall poor results for coastal waters highlight the need for more accurate ocean colour algorithms.

This water‐type based validation approach could be extended to other satellite products with a sufficient number of matching ground observations.

Match-up statistics of the 2015-16 report are superseded by the results provided with this report.

Recommended product use

- **For VIIRS:** Use *OC3* over *GSM* in marine regions with water types 1-4. Chl-a errors in regions associated with water types 6‐8 exceed 200% for both algorithms applied to VIIRS. Data in these regions should be used with care.
- **For MODIS Aqua:** Use *Carder* in marine regions with water types 1‐4. Chl‐a errors in regions associated with water type 4 however exceed 400%. Also use *Carder* in regions with water types 7 and 8 and *GSM* for water type 6 applied to MODIS.

Figure 1 (a) Location of all *in situ* chl‐a measurements, red dots included in 2015‐16 report, blue dots, new data added in this report. (b) Location of in situ chl-a measurements matched with MODIS observations. (c) Location of in situ chla measurements matched with VIIRS observations. Maximum time difference ΔT between in situ and satellite data for this plot is ±24 h.

Figure 2 Spatial distribution of MODIS‐Aqua chl‐a match‐up data (Fig 8b) classified by OWT (ΔT=±24 h).

Figure 3 Spatial distribution of VIIRS chl‐a match‐up data classified by OWT (ΔT=±24 h).

Figure 4 Scatter plots of MODIS‐Aqua chl‐a match‐ups at a maximum time difference of ΔT=±24 h. OWT is indicated by colour. Dashed line is 1:1, solid line is regression for all water types combined. Error bars represent the standard deviation within the match‐up area.

Figure 5 Scatter plots of VIIRS chl‐a match‐ups at a maximum time difference of ΔT=±24 h. OWT is indicated by colour. Dashed line is 1:1, solid line is regression for all water types combined. Error bars represent the standard deviation within the match-up area.

Figure 6 Histograms of chl‐a derived from *in situ* and MODIS Aqua data. Note histogram bins are equal in log‐ transformed chl‐a concentration.

Figure 7 Histograms of chl-a derived from *in situ* and VIIRS data. Note histogram bins are equal in log-transformed chla concentration.

Figure 8 Histograms of the normalised difference chl‐a concentration for MODIS Aqua.

Figure 9 Histograms of the normalised difference chl‐a concentration for VIIRS.

Table 1 Chl-a match-up statistics for the MODIS OC3 algorithm arranged by Optical Water Type at a maximum time difference of ±24 hours. Correlations marked with ** are statistically significant at the P<0.01 probability while those marked with * are statistically significant at P<0.05.

Table 2 Same as table 1 but for the MODIS OCI algorithm.

Table 3 Same as table 1 but for the MODIS GSM algorithm.

Table 4 Same as table 1 but for the MODIS Carder algorithm.

Table 5 Chl‐a match‐up statistics for the MODIS OC3 algorithm arranged by Optical Water Type at a maximum time difference of ±2 hours. Correlations marked with ** are statistically significant at the P<0.01 probability while those marked with * are statistically significant at P<0.05.

Table 6 Same as table 5 but for the MODIS OCI algorithm.

Table 7 Same as table 5 but for the MODIS GSM algorithm.

Table 8 Same as table 5 but for the MODIS Carder algorithm.

Table 9 Chl-a match-up statistics for the MODIS OC3 algorithm arranged by time difference. Correlations marked with ** are statistically significant at the P<0.01 probability while those marked with * are statistically significant at P<0.05.

Table 10 Same as table 9 but for the MODIS OCI algorithm.

Table 11 Same as table 9 but for the MODIS GSM algorithm.

Table 12 Same as table 9 but for the MODIS Carder algorithm.

Table 13: Chl‐a match‐up statistics for the VIIRS OC3 algorithm arranged by Optical Water Type at a maximum time difference of ±24 hours. Correlations marked with ** are statistically significant at the P<0.01 probability while those marked with * are statistically significant at P<0.05.

Table 14 Same as table 13 but for the VIIRS GSM algorithm.

Table 15 Chl‐a match‐up statistics for the VIIRS OC3 algorithm arranged by Optical Water Type at a maximum time difference of ±2 hours. Correlations marked with ** are statistically significant at the P<0.01 probability while those marked with * are statistically significant at P<0.05.

Table 16 Same as table 15 but for the VIIRS GSM algorithm.

Table 17 Chl‐a match‐up statistics for the VIIRS OC3 algorithm arranged by time difference. Correlations marked with ** are statistically significant at the P<0.01 probability while those marked with * are statistically significant at P<0.05.

Table 18 Chl‐a match‐up statistics for the VIIRS GSM algorithm arranged by time difference. Correlations marked with ** are statistically significant at the P<0.01 probability while those marked with * are statistically significant at P<0.05.

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Appendix A Statistics

The statistical measures used in this report are described by the following equations. In the case of MAPE, *x* is the *in situ* measurement and *y* is the satellite observation and *N* is the number of samples (valid match‐ ups).

$$
MAPE = \frac{100}{N} \sum \frac{|y - x|}{x}
$$

For calculation of Bias, RMSE and linear correlation coefficient the input data are log transformed, such that *x* is the log₁₀ of the *in situ* measurement and *y* is the log₁₀ of the satellite observation and *N* is the number of samples (valid match‐ups).

$$
Bias = \frac{1}{N} \sum (y - x)
$$

\n
$$
RMSE = \sqrt{\frac{1}{N} (x - y)^2}
$$

\n
$$
R^2 = \left[\frac{\sum xy - \frac{(\sum x)(\sum y)}{N}}{\sqrt{\left(\sum x^2 - \frac{(\sum x)^2}{N}\right) \left(\sum y^2 - \frac{(\sum y)^2}{N}\right)}} \right]^2
$$

Appendix B Data Repositories

The AODN data portal (https://portal.aodn.org.au) is the primary means of discovering and accessing all IMOS satellite data products. The portal allows browsing of the gridded (mapped) products, download of spatio-temporal subsets in netCDF, and access via THREDDS, which supports OPeNDAP.

A copy of all gridded data sets is also held by CSIRO where a THREDDS server supports direct file access, and also the OPeNDAP and OGC Web mapping service protocols (http://rs-data1-mel.csiro.au/imos-srs). An experimental ERDDAP server (created by NOAA in the US) is also available to access selected gridded data products (http://rs‐data2‐mel.csiro.au/erddap/index.html).

For users requiring direct access to any of the MODIS or VIIRS data sets including the unmapped data in swath format, all data are openly available on the large data storage at the NCI in Canberra, from where they are exposed in the file‐system and via WWW and THREDDS servers.

(http://dap.nci.org.au/thredds/remoteCatalogService?catalog=http://dapds00.nci.org.au/thredds/catalog/ u39/public/data/catalog.xml)

The IMOS Bio‐optical Data Base is available through the AODN portal.

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