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► To cite this version:

Stéphanie Guinehut, B Buongiorno Nardelli, T Chau, F Chevallier, D Ciani, et al.. THE MULTI OBSERVATIONS THEMATIC ASSEMBLY CENTRE OF THE COPERNICUS MARINE ENVIRONMENT MONITORING SERVICE. 9th EuroGOOS International conference, Shom; Ifremer; EuroGOOS AISBL, May 2021, Brest, France. pp.284-291. hal-03335202v2

HAL Id: hal-03335202

<https://hal.science/hal-03335202v2>

Submitted on 24 Sep 2021

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THE MULTI OBSERVATIONS THEMATIC ASSEMBLY CENTRE OF THE COPERNICUS MARINE ENVIRONMENT MONITORING SERVICE

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Abstract

Complementary to ocean state estimates provided by modelling/assimilation systems, a multi observations-based approach is available through the MULTI OBSERVATIONS (MULTIOBS) Thematic Assembly Center (TAC) of the European Copernicus Marine Environment Monitoring Service (CMEMS). CMEMS MULTIOBS TAC provides multi-observation-based ocean products at global scale derived from the combination of two or more different sensors from satellite and in situ, and using state-of-the-art data fusion techniques. These products cover the blue ocean for physics and the green ocean for the carbonate system and biogeochemical variables. MULTIOBS products are available in Near-Real-Time (NRT) or as Multi-Year Products (MYP) for the past 25 to 35 years with regular temporal extensions. MULTIOBS TAC provides also associated Ocean Monitoring Indicators (OMIs). It uses mostly inputs from other CMEMS TACs.

Keywords: *in situ*, satellite, multi observations, data fusion techniques, CMEMS

1. Introduction

MULTIOBS TAC products are available through CMEMS catalogue (<https://marine.copernicus.eu/>). They consist of global ocean state-estimates of variables still currently critically under-sampled at most of the major scales, take advantage of the strength of the Global Ocean Observing System (*in situ* and satellite), stay close to the observations, resolve mesoscale structures at the right place (when eddy permitting) and provide long, stable time series enhancing ocean climate and ocean health monitoring capabilities. MULTIOBS TAC delivers 4 physical, 1 carbon and 2 biogeochemical products and 3 associated OMIIs which are first described. Some perspectives are then listed below.

2. MULTIOBS TAC products

2.1 SSS/SSD

CNR produces a global 2D sea surface salinity (SSS) and sea surface density (SSD) L4 product by interpolating *in situ* SSS and SSD data with the multidimensional OI technique originally introduced by Buongiorno Nardelli (2012). This method is able to extract information on the surface patterns from satellite Sea Surface Temperature (SST) L4 data, increasing the effective resolution of the interpolated fields. The technique, originally developed to interpolate SSS, has been modified by Droghei et al. (2016) to provide dynamically consistent SSD field. It has been further modified to ingest satellite fields from SMOS. The product is provided on a 1/4° regular grid at weekly sampling (monthly averaged fields are also available) from 1993 onward (MULTIOBS_GLO_PHY_S_SURFACE_MYNRT_015_013).

2.2 ARMOR3D

ARMOR3D which is developed by CLS provides on a 1/4° horizontal grid at weekly sampling (monthly averaged fields are also available) 3D global fields of temperature, salinity, geopotential heights and geostrophic currents down to the bottom on 50 vertical levels and 2D global fields of mixed layer depth. This L4 product is available as a multi-year time series since 1993 and in near-real time (MULTIOBS_GLO_PHY_TSUV_3D_MYNRT_015_012). It is obtained by combining satellite (sea level anomaly (SLA), geostrophic currents, SST and SSS) and *in situ* (T/S profiles) observations through statistical methods (multiple linear regression, optimal interpolation, thermal wind equation: Guinehut et al., 2012; Mulet et al., 2012).

ARMOR3D is used together with 4 global reanalyses (GREP) and the *in situ* gridded field CORA (CORiolis Re-Analysis) to monitor global ocean warming through a specific Ocean Monitoring Indicator for the Global Cumulative Trend of zonal mean Subsurface Temperature. Estimation of the robustness of the indicator is provided from a multi-product approach. From ARMOR3D solution, confirmed by all solutions, the warming is obvious in almost all part of the ocean and at depths of up to 800 m depth (Figure 1).

Fig. 1. ARMOR3D Depth/Latitude global mean temperature cumulative trend over 1993-2019 (in °C).

2.3 OMEGA3D

A new multiyear product called OMEGA3D (MULTIOBS_GLO_PHY_W_3D_REP_015_007) and developed by CNR provides observation-based 3D quasi-geostrophic vertical and horizontal ocean currents over 75 levels from the surface to 1500 m depth, at $1/4^\circ$ horizontal resolution, for the period January 1993 to December 2018. The current velocities are obtained by solving a Q-vector formulation of the Omega equation, including diabatic forcing terms (Buongiorno Nardelli, 2020). This product is based on the combination of the fields of temperature, salinity and geostrophic currents provided by ARMOR3D and ERA-Interim surface fluxes.

2.4 Copernicus-Globcurrent Surface and near-surface current

Although it is well recognized that surface currents in the ocean are not the simple addition of different current components, a simple approximation of ocean currents can be made by adding the geostrophic currents derived from the altimeter Absolute Dynamic Topography fields and an estimate of the Ekman response to wind forcing. Based on the work by Rio *et al.*, (2014), CLS produces total velocity fields (zonal and meridional) at 0 m and 15 m depth and at 6 h frequency in near-real time (MULTIOBS_GLO_PHY_NRT_015_003) and at 3 h frequency for the multi-year time series (MULTIOBS_GLO_PHY_REP_015_004). Daily and monthly means are also available. The multi-year time series currently covers the 1993-May 2020 period and is extended in the near-real time.

The geostrophic currents are calculated through the geostrophic approximation applied to the sum of altimeter SLA and a mean dynamic topography (MDT), both coming from CMEMS Sea Level TAC. The Ekman currents are computed at two depths (0 m and 15 m) applying an empirical Ekman model updated from Rio *et al.*, (2014) to ECMWF wind stress fields. The parameters of the empirical Ekman model are

computed using in situ observations from Argo drifts at the surface, SVP-type drifters at 15 m and an estimation of the ocean stratification coming from the mixed-layer depth of ARMOR3D.

2.5 CMEMS-FFNN Surface Carbon

The surface carbon multiyear product includes air-sea flux of CO_2 (fgco_2), partial pressure of CO_2 (spco_2) and pH (MULTIOBS_GLO_BIO_CARBON_SURFACE_REP_015_008). The model called CMEMS-FFNN is inherited from the EU H2020 AtlantOS project and is developed by LSCE. It relies on the implementation of feed-forward neural network models (FFNN) for the interpolation of sparse carbon system measurements to basin-wide maps on a 1° horizontal resolution grid at a monthly period (Denvil-Sommer *et al.*, 2019). This method establishes non-linear relationships between chosen drivers or 'predictors' such as SST, SSS, chlorophyll, mixed layer depth, surface height, pCO_2 climatology, and the target, here surface ocean pCO_2 measurements from SOCAT (<https://www.socat.info/>). Reconstructed variables (fgco_2 , spco_2 , pH) are distributed with associated uncertainties derived from the 100-member CMEMS-FFNN ensemble. Related OMs for ocean carbon sink (i.e. global yearly integrated air-sea flux of CO_2) and ocean acidification (i.e. global mean sea water pH) are also available. CMEMS-FFNN air-sea CO_2 flux product contributes since 2019 to the yearly assessment of the Global Carbon Budget (Friedlingstein *et al.*, 2020).

2.6 CANYON Nutrient profiles

The so-called method CANYON-B (Carbonate system and Nutrients concentration from hYdrological properties and Oxygen using a Neural-network) is inherited from the EU H2020 AtlantOS project and R.Sauzède PHD thesis and is developed by LOV/IMEV. It relies on a neural-network method to derive, from 'simple' and cost-effective measured parameters from Biogeochemical-Argo (BGC-Argo) profiling floats, some more complex biogeochemical measurements, not yet easily or cost-effectively amenable to robotic detection (Sauzède *et al.*, 2017; Bittig *et al.*, 2018). CANYON-B uses measurements of temperature, salinity, pressure, and oxygen together with sampling latitude, longitude, and date to retrieve the concentrations of three nutrients including nitrates (NO_3^-), phosphates, (PO_4^{3-}) and silicates (Si(OH)_4). It has been trained on high quality nutrient data collected over the last 30 years and made available through the GLODAPv2 database. It is then applied to all available delayed-mode qualified BGC-Argo profiling floats equipped with an oxygen sensor. Nearly 100 000 profiles are currently available in CANYON product (MULTIOBS_GLO_BIO_NUTRIENTS_PROFILES_REP_015_009).

The example along Argo float WMO 6901467 trajectory located in the North Western Mediterranean Sea shows winter mixing in the input fields (temperature, salinity, oxygen) in February/March 2013 and associated uplift of reconstructed nutrients (Figure 2).

2.7 SOCA3D POC/ b_{bp} /Chla

The SOCA3D biogeochemical multiyear product provides global gridded fields of Particulate Organic Carbon (POC), backscattering coefficient (b_{bp}) and Chlorophyll-*a* concentration (Chla) at depth (MULTIOBS_GLO_BIO_BGC_3D_REP_015_010), three biogeochemical variables still currently critically under sampled at all scales. The method relies first on a neural network method called SOCA (Satellite Ocean-Color merged with Argo) which is inherited from the EU H2020 AtlantOS project and R.Sauzède PHD thesis. SOCA estimates vertical profiles of backscattering coefficient (b_{bp}), a bio-optical proxy for POC, from surface ocean color satellite measurement of bbp and additional physical drivers (Sauzède *et al.*, 2016). It has been trained on high quality b_{bp} data collected from BGC-Argo floats. Then, an empirical parameterization is used to infer the vertical distribution of Chla from surface ocean color satellite observations of Chla and the relative position of the mixed layer and euphotic depths (Uitz *et al.*, 2006). This parameterization was established from a database of Chl profiles acquired by High Performance Liquid Chromatography (HPLC), the reference method for such measurements. Both methods are developed at LOV/IMEV. The multiyear product is available on a $1/4^\circ$ horizontal grid over 36 vertical levels from the surface to 1000 m depth, at a weekly time period and for the 1998 – 2019 period. An associated monthly mean climatology is also available.

Fig. 2. Depth/Time Temperature (in $^\circ\text{C}$), Salinity (in psu) and Oxygen (in $\mu\text{mol kg}^{-1}$) measurements from Argo float WMO 6901467 located in the North Western Mediterranean Sea and associated Nitrate, Phosphate and Silicate (in $\mu\text{mol kg}^{-1}$) profiles as reconstructed by CANYON-MED (Fourrier *et al.*, 2020).

2.8 Quality assessment

All products are fully validated applying a methodology defined and agreed with CMEMS, inheriting the long experience of MyOcen and MERSEA series of projects (Hernandez *et al.*, 2018). A dedicated QUID (QQuality Information Document) is associated with each product/OMI and is available through the generic address: <https://resources.marine.copernicus.eu/documents/QUID/CMEMS-MOB-QUID-015-XXX.pdf> (XXX to be replaced by the number of the product).

3. Perspectives

Perspectives include to remain at the state-of-the art in terms of methods and upstream data for which continuous monitoring is performed. New/improved products from CMEMS TACs (Sea Level, SST, Ocean Color, Wind, in situ) will be incorporated and tested as well as products currently outside of CMEMS, such as SSS from Aquarius/SMAP and from the future ESA-CIMR (Copernicus Imaging Microwave Radiometer) mission. Perspectives include also the improvement of the information content of the products by increasing resolution in space (horizontal & vertical) and in time for all MULTIOBS products, starting with the physical ones. Higher resolution (L3/L4) products from CMEMS TACs are thus required. OMEGA3D, nutrient profiles and 3D POC/Chla products currently available only as multi-year time series could be extended to the near-real-time, providing fully consistent multi-year and near-real time series for almost all MULTIOBS products. The representation of physical processes such as ageostrophic component (wind-driven, stokes), higher frequency processes (e.g. tides) as well as mesoscale processes (e.g. vertical extension of eddies) will be improved. New variables such as phytoplankton functional types, total primary production, or vertical profiles of carbonate chemistry (alkalinity, pH, DIC) will be added. Derived quantities such as Ocean Heat Content (ECV) could be easily provided. Efforts will be made to provide uncertainties estimate for all variables for both direct use, validation purposes and data assimilation.

Acknowledgements

This work has been funded by the European 'Copernicus Marine Environment Monitoring Service' under contract 83-CMEMS-TAC-MOB.

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