



1 1. Introduction

2 In the decade of 1990s, the Topex/Poseidon (T/P) mission combined with the ERS-1/2
3 satellites changed our sight of the sea level and ocean circulation fluctuations radically. Since
4 then, altimeter missions have furnished accurate measurements of Sea Surface Height (SSH).
5 Mean Sea Level (MSL) has been monitored from multiple datasets provided by different
6 altimetry missions presently in flight (Jason 2 and 3, HY-2A, CryoSat-2, Sentinel-3A,
7 SARAL/Altika) and those no longer supplying data (the aforementioned T/P, ERS-1 and 2, Jason
8 1, Envisat, Geosat Follow-On) [Legeais *et al.*, 2016]. Altimetry resolves the mesoscale thanks to
9 a finest spatio-temporal sampling. Nevertheless, even though SSH estimates are becoming
10 more precise, the uncertainty associated with altimeter measurements and the geophysical
11 altimeter corrections applied in the SSH computation remains relatively high [Ablain *et al.*,
12 2009; Couhert *et al.*, 2014; Legeais *et al.*, 2014; Rudenko *et al.*, 2014]. For this reason, some
13 external and independent measurements provided by in-situ observations and numerical
14 models are required to calibrate and validate the altimeter Sea Level Anomaly (SLA) data.
15 These comparisons allow us to obtain the altimetry errors relative to the external
16 measurements and provide an improved picture of SSH that can be used for global and
17 regional studies.

18 Tide gauges data are usually considered [e.g. Mitchum 1998, 2000; Nerem *et al.*, 2010;
19 Arnault *et al.*, 2011; Villadeau *et al.*, 2012] because they furnish high temporal resolution time
20 series of Sea Surface Height (SSH) in coastal areas. However, these instruments are not
21 homogeneously allocated over the coasts. A complementary approach can be done by using (i)
22 in-situ Dynamic Heights Anomalies (DHAs) derived from the Temperature and Salinity (T/S)
23 vertical profiles computed from the Argo network [see e.g. Guinehut *et al.*, 2012; Valladeau
24 *et al.*, 2012; Legeais *et al.*, 2016] and glider measurements [e.g. Ruiz *et al.*, 2009a, 2009b;
25 Bouffard *et al.*, 2010] or (ii) velocity data provided by drifters [e.g. Escudier *et al.*, 2013;
26 Troupin *et al.*, 2015].

27 In this study, we will focus on the altimeter products for the Mediterranean basin. The
28 Mediterranean Sea is a semi-enclosed basin connected with the Atlantic Ocean through the
29 Strait of Gibraltar. It also communicates with the Black Sea through the Turkish Bosphorus and
30 Dardanelles Straits. The Sicily channel separates the eastern and western basins [Criado-
31 Aldeanueva *et al.*, 2012]. The basin-scale circulation of the Mediterranean interacts with sub-
32 basin scale and mesoscale processes, then forming a highly variable general circulation. As a
33 consequence, the Mediterranean Sea is a particularly interesting area for among others



1 physical studies since most of the ocean processes that occur in the world ocean can be also
2 found in this basin. Therefore, the Mediterranean can be considered as a reduced scale ocean
3 laboratory, where processes can be characterized with smaller scales than in other ocean
4 regions [Malanote-Rizoli *et al.*, 2014]. In this context, the internal Rossby Radius of
5 deformation in the basin is $O(10-15\text{ km})$, this being four times smaller than typical values for
6 much of the world ocean according to Robinson *et al.* [2001]. This fact promotes that in the
7 Mediterranean Sea the spatial resolution of the lagrangian profiling floats of the Argo
8 programme, which consists of a global network of more than 3000 operating floats [Roemmich
9 *et al.*, 2009; Riser *et al.*, 2016] drifting with less than 3 degrees mean spacing, should be
10 reduced four times compared to the open ocean. The Argo programme is a major component
11 of the Global Ocean Observing System and aims to monitor the changing temperature and
12 salinity fields in the upper part of the ocean [Riser *et al.*, 2016]. The majority of the profiling
13 floats used in Argo are programmed to drift at a nominal depth (known as the parking depth)
14 of 1000 m [Riser *et al.*, 2016]. They collect temperature and salinity data every 10 days from
15 the upper 2000 m of the world oceans in order to observe the slow evolution of the large-scale
16 ocean structure.

17 Argo and satellite altimetry are entirely complementary. The combination of in-situ Argo
18 data with SSH anomalies derived from satellites allows us to construct time series of the
19 dynamical state of the ocean circulation [Riser *et al.*, 2016]. At present, Argo data are
20 systematically used together with altimeter data to describe and forecast the 3D ocean state,
21 for ocean and climate research and for sea level rise studies [see e.g. Guinehut *et al.*, 2012; Le
22 Traon, 2013]. Although Argo does not resolve the mesoscale, the joint use of Argo, altimetry
23 and numerical models, through effective data assimilation techniques can provide a good
24 representation of mesoscale temperature and salinity fields. This fact demonstrates the very
25 strong and unique complementarities of the two observing systems [Le Traon, 2013].

26 The Argo network in the Mediterranean Sea consists presently of around 80 operating
27 floats deployed in the frame of the MedArgo program
28 (<http://nettuno.ogs.trieste.it/sire/medargo/active/index.php>). The specific semi-enclosed
29 morphology with a large fraction of coastal areas, shallow bathymetry and circulation
30 structures of the basin make profilers programmed with the Argo standard global parking
31 depth of 1000 m not appropriate for this program [Poulain *et al.*, 2007]. This is why a parking
32 depth of 350 m was chosen for the Mediterranean basin. The objective was to track the
33 intermediate waters throughout the Mediterranean which are mostly composed by Levantine
34 Intermediate Water (LIW). This water mass is formed during winter convection in the northern



1 Levantine sub-basin being a crucial component of the Mediterranean thermohaline “conveyor
2 belt” circulation [Poulain *et al.*, 2007]. According to the small radius of deformation of the
3 Mediterranean compared with the open ocean, the current number of operating floats in the
4 basin (equivalent to an average spatial resolution of around 2 degrees) improves the global
5 coverage of the Argo network. Nonetheless, it is not enough to properly capture the significant
6 mesoscale circulation features of the basin.

7 The aim of this paper is to investigate which Argo design sampling in the Mediterranean
8 Sea is necessary to recover the mesoscale signal as seen by altimetry. The monitoring of the
9 mesoscale structures is not an initial target of the Argo network [Riser *et al.*, 2016]. However,
10 this is an interesting question in the perspective of future network extensions in order to
11 improve ocean state estimates. Actually, the Argo Steering Team has recently provided a
12 roadmap for how the Argo mission might expand in the near future [Riser *et al.*, 2016].
13 According to these authors, one of the proposed projects is to support an increase in the
14 spatial sampling resolution in particular areas of the world ocean. The objective is the
15 improvement of our view of the complex structure of oceanic variability at spatial scales lesser
16 than the climate scale.

17 To accomplish the proposed aim, we conduct several Observing System Simulation
18 Experiments (OSSEs) in the basin. OSSEs provide a methodology to evaluate and design
19 optimum sampling strategies in ocean observing systems (OOS) [Alvarez and Mourre, 2012].
20 Usually, the method consists in considering the outputs of an ocean model simulation of the
21 area monitored by the OOS as “truth.” Virtual observations from different ocean observing
22 platforms in the OOS are then simulated from the model run and analysed in the same manner
23 than real data [e.g. Alvarez and Mourre, 2012]. OSSEs have been used in oceanography to
24 analyse the impact of different components of the global OOS for ocean analysis and
25 forecasting (see e.g. Oke and Schiller [2007]; Guinehut *et al.* [2012]; Alvarez and Mourre
26 [2012]; Ninove *et al.* [2015]; Oke *et al.*, [2015a] or Oke *et al.*, [2015b]). Here, however, a
27 different approach will be followed: we simulate the Argo observing system in the
28 Mediterranean based on specific altimetry gridded merged product for the Mediterranean Sea
29 and not from an ocean model simulation. This approach is similar to the one followed by
30 Pascual *et al.*, [2009]. These authors evaluated the quality of global real-time altimetric
31 products by comparing them with independent in-situ tide gauges and drifter data. Our study
32 will then assess the scales covered by altimetry which are larger than 100 km [Pujol and
33 Larnicol, 2005]. Notice that the scales mentioned in this paper allude to a definition based on
34 the diameter of individual structures, usually referred to as “feature scales”.



1 The paper is organized as follows: the datasets are described in Section 2. Section 3 details
2 both the processing sequence developed to compare the altimeter data with Argo in-situ
3 measurements and the quantification of the differences between Argo – SLA . These
4 differences are needed to conduct the OSSEs. Thus, a quality assessment of the performances
5 of the altimeter product in the Mediterranean Sea is performed in the first part of this study.
6 The method used here to evaluate the altimeter data is based on the comparison of SLAs from
7 altimetry and DHAs computed from the in-situ Argo network. Section 4 is devoted to the
8 experiments conducted to recover the SLA fields in the basin from the different configurations
9 of the simulated Argo arrays. Finally, discussion and suggestions to the Argo community
10 regarding future prospects of the in-situ network in the Mediterranean Sea are given in Section
11 5.

12 **2. Datasets**

13 **2.1 ARGO dataset**

14 We use delayed mode quality-controlled T/S profiles from 2003 to middle 2015 as
15 obtained from the Coriolis Global Data Assembly Centre (www.coriolis.eu.org, ARGO GDAC
16 global distribution database) in the Mediterranean Sea (Figure 1). Dynamic Height (DH) was
17 computed at 5 m depth as an integration of the pressure, temperature and salinity vertical
18 profiles through the water column using a reference level at 400 dbar and 900 dbar (close to
19 400m and 900m, respectively). The choice of these reference levels is conditioned by the
20 availability of the climatology used to compute DH anomalies. This issue will be addressed
21 later. An additional quality control criterion relative to both the profile's position and the T/S
22 measurements was applied: only profiles with a quality position flag of 1 (good data) were
23 employed. The major restriction, however, comes from the salinity data close to the sea
24 surface. Profiles exhibiting salinity flags of 3 or 4 (bad data) in the first 5 meters of the water
25 column were removed before the DH computation. As a result of this additional quality check,
26 194 Argo floats and about 17000 T/S profiles distributed over almost the whole Mediterranean
27 basin are available to compute DH. Their deployment's temporal evolution is shown in Figure
28 2. More than 90 floats and almost 9000 profiles have been deployed in the last three years of
29 the period investigated. They represent more than 50 % of the Mediterranean Argo network.
30 Actually, the number of both floats and profiles has been systematically increasing from 2008
31 until 2015 reaching its maximum value in 2014 (36 floats deployed and nearly 4000 profiles
32 carried out).



1 To calculate a consistent DHA with the altimeter SLAs, we use a mean dynamic height as a
2 reference computed through a synthetic climatology approach [Guinehut *et al.*, 2006]. The
3 method to compute the synthetic climatology described in Guinehut *et al.* [2006] consists in
4 the combination of altimeter SLA with simultaneous in-situ dynamic height in order to
5 compute a mean dynamic height, which is referred to the time period spanning from January
6 2003 to December 2011. This climatology presents a global coverage and it has been recently
7 used by Legeais *et al.* [2016] to analyse global altimetry errors by using Argo and GRACE data.
8 In this paper we will test the mean dynamic height computed in the Mediterranean Sea at 400
9 dbar and 900 dbar to estimate DHAs.

10 2.2 Altimeter measurements

11 Radar altimeters provide SSH measurements that are not directly comparable with in-situ
12 measurements. Therefore, they must be first referenced and corrected from geophysical
13 signals in order to determine SLAs. In this work, we use SLAs obtained from SSALTO/DUACS
14 multimission (Saral, Cryosat-2, Jason-1, Jason-2, T/P, Envisat, GFO, ERS-1, ERS-2, and Geosat)
15 specific reprocessed gridded merged product (level 4) for the Mediterranean Sea. This product
16 is available in the Mean Sea Level Anomaly (MSLA) section of the Archiving, Validation and
17 Interpretation of Satellite Oceanographic website (AVISO, <http://www.aviso.altimetry.fr>). It
18 has been computed with respect to a twenty-year mean referred to the period 1993 – 2012. A
19 comprehensive description of SSALTO/DUACS is given in Pujol *et al.* [2013] and Pujol *et al.*
20 [2016]. The spatial resolution of the dataset is $\frac{1}{8}^{\circ} \times \frac{1}{8}^{\circ}$ and the time period used in this work
21 spans from January 1993 to April 2014. The quality of this product can be estimated among
22 others by comparison with in-situ Argo data. To perform this comparison, it is critical that both
23 types of data have the same interannual temporal reference [Legeais *et al.*, 2016]. Thus, the
24 temporal reference of the altimeter SLA is adapted to the time period spanning from 2003 to
25 2011 (reference period of the synthetic mean Argo dynamic height). To do that, we subtract
26 the mean of altimetric SSALTO/DUACS maps (mean value of 3.54 cm for the whole basin) over
27 2003 – 2011 from the original SLA time series [Valladeau *et al.*, 2012]. On the other hand, the
28 physical content captured by altimetry and Argo profiles is not precisely the same [Dhomps
29 *et al.*, 2011] because the barotropic and the deep steric (deeper than the reference level of the
30 Argo DHA) contributions are missing from the Argo measurements. Therefore, the comparison
31 of altimeter SLA and in situ Argo DHA is used to detect relative anomalies in altimeter data and
32 not absolute bias [Valladeau *et al.*, 2012]. This comparison allows us to obtain both the
33 instrument and the representation errors which are needed to perform the OSSEs.
34 Representation error can be defined as the component of observation error due to unresolved



1 scales and processes [Oke and Sakov, 2008]. In other words, it is the part of the true signal that
2 cannot be represented on the chosen grid due to limited spatial and temporal resolution.

3 **3. Error estimates from comparison of Argo dynamic heights and** 4 **altimetry sea level anomalies**

5 This section focuses on the comparison of altimetry data with Argo dynamic height in
6 order to estimate the errors in Argo – SLA differences needed to conduct the OSSEs. In
7 addition, this analysis can contribute to validate satellite SLAs with an increased confidence. A
8 sensitivity analysis of the method of comparison of both datasets is provided. This analysis
9 mainly focuses on the impact of the reference depth selected in the computation of the Argo
10 DH on the comparison with specific altimetric SLA gridded merged product for the
11 Mediterranean Sea.

12 **3.1 Method for comparing Altimetry and in-situ Argo data**

13 The comparison method of altimetry with Argo data consists in co-locating both types of
14 datasets since spatial and temporal sampling of altimetry and Argo data are different
15 [Valladeau *et al.*, 2012]. Altimeter grids and synthetic climatologies were spatially and
16 temporally interpolated at the position and time of each in situ Argo profile, which is
17 considered as reference, by using a mapping method based on an optimal interpolation
18 scheme. This considerable reduces errors due to different sampling characteristics of altimeter
19 and in-situ data. The period investigated spans from January 2003 (beginning of the Argo
20 dataset) to April 2014 (ending of the altimetric data used in this study). Then, statistics
21 analyses are performed between both datasets. Co-located altimeter and Argo DH differences
22 are analysed in terms of the standard deviation (STD) for the two reference levels used to
23 compute DHAs from the Argo profiles (namely 400 and 900 dbar). In addition, the robustness
24 of the results was investigated by computing means of a bootstrap method with 10^3 random
25 samples taken from the original SLA-DHA series (see details of the method in Efron and
26 Tibshirani [1993]). The studies conducted include: (i) the assessment of the method of
27 comparison between Altimetry and Argo data in the Mediterranean Sea; and (ii) the evaluation
28 of the impact of the reference depth selected in the computation of the Argo dynamic height.

29 **3.2 Sensitivity to the reference depth for the integration of the Argo dynamic height**

30 The integration of the Argo T/S profiles for the computation of the in-situ dynamic heights
31 requires a reference level (pressure) where null horizontal velocities are assumed [Legeais *et*



1 *al.*, 2016]. As a rule, the deeper the reference level, the more information from the T/S profiles
2 is considered. This involves a well sampled steric signal through the water column. However, a
3 lower number of vertical profiles (those that reach the reference level) are used in the
4 computation. On the contrary, shallower reference levels allow us to use more floats, although
5 the vertical steric signal will be less sampled. Thus, we aim at determining the impacts of a
6 given reference depth of integration on the Argo spatial sampling and on the comparison with
7 altimeter data in the Mediterranean basin.

8 As it was aforementioned, the choice of a deep reference level for Argo DHAs provides a
9 better estimation of the baroclinic signal. This is more in agreement with the observed signal
10 by altimetry [*Legeais et al.*, 2016]. Therefore, we conduct the analysis on DH comparison
11 computed from Argo data referred to the deeper available reference depth of 900 dbar (nearly
12 900 m) and the specific altimetry product for the Mediterranean Sea. Results are reported in
13 Table 1. The number of T/S Argo profiles used to compute DH (those that reach at least 900 m
14 depth) was 416, corresponding to 23 floats. The standard deviation of the differences between
15 DH from altimetry and Argo (SLA minus DHA) for the common period investigated (from
16 January 2003 to April 2014) was 5.31 cm. It is equivalent to more than 95 % of SLA signal
17 variance. The correlation between both datasets was 0.80.

18 In order to study the impact of the reference level, we repeated the analysis using the
19 shallower reference level of 400 dbar (almost 400 m) for the Argo anomalies but using the
20 same array of Argo profiles reaching 900 m. Now, 24 floats and 479 profiles are available to
21 compare with altimetry due to the synthetic climatology used to compute DHA referred to 900
22 dbar (see Table 1). Nonetheless, we kept the same number of floats and profiles than in the
23 previous computation in order to make both results comparable. The standard deviation of the
24 differences between SLA and DHA referred to 400 dbar computed from profiles spanning until
25 900 m depth was 5.04 cm (see Table 1). It represents an improvement of nearly 10 % in terms
26 of signal variance with respect to the STD diff. computed from Argo DHA referred to 900 dbar
27 (5.31 cm). Moreover, the correlation coefficient increased from 0.80 to 0.82. These slightly
28 better results (also confirmed from the bootstrap analyses) show that in the Mediterranean
29 basin, it will be advisable to compare SLA from altimetry with DHA from in-situ Argo data
30 referred to 400 dbar.

31 Consequently, DHA referred to 400 dbar was recomputed but using all the available
32 profiles reaching 400 m depth. Now, the number of T/S Argo profiles used to compute DH
33 increased to 2258, thus corresponding to 41 Argo floats. Notice that this more comprehensive



1 number of Argo profiles is almost 6 times larger than the profiles used to compute DHAs
2 referred to 900 dbar. The standard deviation of the differences of SLA– DHA was 4.92 cm
3 while the correlation between both datasets decreased to 0.76. We will consider this STD
4 value as the mean error of the Argo – SLA differences in the Mediterranean and therefore it
5 will be used to perform the OSSEs. Notice that this result represents an improvement of 14 %
6 in terms of signal variance with respect to the one obtained from the differences between SLA
7 and DHA referred to 900 dbar. This is an unexpected result since the larger thickness of the
8 water column integrated in the latter should promote a lower value of STD. A plausible
9 explanation of this outcome will be done in Section 5.

10 **4- Impact of the number of Argo floats on the reconstructed SLA fields**

11 In this section we aim to investigate which configuration in terms of spatial sampling of the
12 Argo array in the Mediterranean Sea will properly reproduce the mesoscale dynamics in this
13 basin, which is comprehensively captured by new standards of specific altimeter products for
14 this region. To do that, several OSSEs have been conducted to simulate the Argo observing
15 system in the Mediterranean assuming altimetry data computed from specific reprocessed
16 gridded merged product for the basin as the “true” field.

17 **4.1 Experiments design**

18 In a first step, OSSEs have been performed for daily SLA maps along 2014 by applying the
19 Optimal Interpolation (OI) technique. The region considered covers the entire Mediterranean
20 basin. The original altimetry dataset has a spatial resolution of $\frac{1}{8}^{\circ} \times \frac{1}{8}^{\circ}$ and presents 17283 grid
21 points (see Table 2). Daily SLA maps were sub-sampled with the different spatial resolutions
22 displayed in Table 2 in order to reproduce some possible configurations of the Argo array
23 network in the Mediterranean. The stations (grid points) associated with each sub-sampled
24 field (figures not shown) will simulate the positions of the Argo floats over a regular grid.

25 Before the computation, the sub-sampled daily SLA maps were perturbed with a random
26 noise by using a normal distribution function only depending on the standard deviation of the
27 differences of SLA – DHA (4.92 cm) computed in Section 3. This STD diff. corresponds to the
28 sum of the instrument and the representation errors. Then, seven experiments were
29 conducted to reconstruct the different sub-sampled daily SLA fields in the Mediterranean
30 along 2014 with a spatial resolution of $\frac{1}{8}^{\circ} \times \frac{1}{8}^{\circ}$ by applying the OI technique. The parameters
31 used for the computation of the reconstructed fields were the following: (i) the first guess used
32 to obtain the statistically null-mean residuals was computed by fitting a polynomial of degree



1 1. This first guess will be subsequently added after the computation to recover the total daily
 2 field; (ii) the filtering scale was set to be twice over the spatial distance between stations
 3 (according to the box size used in each experiment). Table 2 summarizes the filtering scale
 4 used to compute the recovered SLA fields in the different reconstructions; (iii) the spatial scale
 5 of correlation between stations was determined from a Gaussian correlation curve computed
 6 as follows:

$$7 \quad W = e^{-d^2/2 \cdot S^2} \quad (1)$$

8 where d is the mean distance between stations and S the spatial scale of correlation. In order
 9 to determine the more suitable spatial scale of correlation for the Mediterranean basin we
 10 computed the correlation curve W for spatial scales varying from 15 km to 50 km. The mean
 11 distance between stations ranged between 0 km and 100 km. Then we compared these
 12 correlation curves with the one obtained for altimetric data computed for the same distances
 13 between stations as follows:

$$14 \quad COR(x) = \left[1 + ar + \frac{1}{6}(ar)^2 - \frac{1}{6}(ar)^3 \right] e^{-ar} \quad (2)$$

15 with $r = x/L$ and $a = 3.337$; where x is the spatial coordinate of the studied point, and L is
 16 the zonal correlation scale (km) of the Mediterranean basin (100 km). The reader is referred to
 17 *Pujol and Larnicol* [2005] for a more detailed description of this computation. Figure 3 shows
 18 the correlation curve computed for the altimetric data from Eq. (2) and the best fitting curve
 19 obtained from Eq. (1), which corresponds to a spatial correlation scale of 40 km. Therefore, the
 20 S parameter was set to 40 km in all the experiments. (iv) the last parameter to include in the
 21 experiments is the noise to signal variance ratio (γ), defined as the ratio between the Argo
 22 error and the altimetry variance. The former can be established as the variance of the
 23 differences between SLA and DHA in the Mediterranean. This parameter is estimated from the
 24 standard deviation of SLA-DHA differences (4.92 cm) computed in Section 3. As a result, we
 25 obtain $\gamma=0.9$ as the true value for the datasets used here (see further details about this
 26 parameter in *Gomis et al.* [2001]).

27 Finally, the retrieved daily SLA maps for 2014 were compared with the original ones (also
 28 interpolated to a spatial resolution $0.25^\circ \times 0.25^\circ$) in order to compute the root mean square
 29 errors (RMSE) associated with the recovered maps from the sub-sampled fields. This
 30 procedure will let us establish the spatial resolution that better captures the mesoscale
 31 dynamics in the Mediterranean with a feasible number of stations simulating the locations of
 32 Argo floats.



1 **4.2 Impact of the grid box size on analysed SLA fields**

2 In this section we will discuss the impact of the spatial resolution of the sub-sampled SLA
3 fields on the retrieval of mesoscale signals in the Mediterranean basin. As a previous step, the
4 RMSE obtained for the seven experiments will be analysed. The 2014 yearly mean values of
5 the RMSE associated with the altimetry maps recovered from the different sub-sampled fields
6 and their annual variability are displayed in Figure 4. Maximum mean RMSE larger than 4 cm
7 (equivalent to 79 % of SLA signal variance) are obtained for the maps recovered from the sub-
8 sampled field reproducing the current spatial resolution of the Argo array in the
9 Mediterranean ($2^\circ \times 2^\circ$). Therefore, this spatial configuration only retrieves 21 % of SLA signal
10 variance due to a poorer capture of the mesoscale features. These maps also exhibit the larger
11 annual variability. This is an expected result that can be explained by both the challenge of
12 reconstructing the same scale signals with only 69 stations (grid points) and the larger filtering
13 scale (around 450 km) used in the experiment (see Table 2). The mean RMSE of the recovered
14 maps exponentially decays as the box-size of the sub-sampled altimetry fields diminishes and
15 therefore, the number of stations enhances. As a result, the mean RMSE reaches an
16 asymptotic value of 2.4 cm (equivalent to 28.7 % of SLA signal variance) for the SLA maps
17 retrieved from the sub-sampled fields with a box-size of $0.4^\circ \times 0.4^\circ$. This configuration is
18 equivalent to 1458 stations and captures 71.3 % of SLA signal variance. The standard deviation
19 of the RMSE follows the same pattern exhibiting a minimum annual variability for this spatial
20 resolution.

21 Figure 5 shows an example of the altimetry maps recovered from the sub-sampled SLA
22 fields on 22nd December 2014. The original SLA field for that day interpolated to a spatial
23 resolution of $\frac{1}{3}^\circ \times \frac{1}{3}^\circ$ is displayed in the uppermost panel for comparisons purposes. Notice
24 that the coarse spatial resolution of the $2^\circ \times 2^\circ$ sub-sampled grid (upper-left panel in Figure 5)
25 prevents us from retrieving the mesoscale features observed in the original map and only the
26 large-scale signals are properly captured. As a consequence, the RMSE associated with this
27 reconstruction which simulates the present Argo array in the Mediterranean are around 4.6
28 cm. On the contrary, the sub-sampled grids with box-sizes of $0.4^\circ \times 0.4^\circ$ and lower (map not
29 shown) are able to retrieve most of the mesoscale structures of the basin with an RMSE of
30 around 2.6 cm. Nonetheless, the high number of stations required to reconstruct the SLA maps
31 (respectively 1458 and 1915, see Table 2) makes this option unviable. Therefore, it is
32 imperative to reach a compromise between the stations used and the extent of the
33 reconstruction performed. In this case, a reasonable solution would be to reconstruct the SLA
34 field from a sub-sampled grid with a box-size of $0.75^\circ \times 0.75^\circ$. This spatial resolution agrees



1 with the theoretical one for the Argo array in the Mediterranean extracted from the internal
2 Rossby radius of deformation computed for the Mediterranean basin. Also, it allows us to
3 retrieve the most representative mesoscale patterns of the basin, for spatial scales larger than
4 150 km, with a feasible number of Argo floats (450 stations). Moreover, the spatial scales
5 resolved by this configuration simulate the spatial scales captured by the altimetry.

6 **4.3 Sensitivity to the irregular sampling**

7 The experiments conducted above let us recover SLA maps computed from theoretical
8 regular-gridded configurations of the Argo array in the Mediterranean. In this section we aim
9 at retrieving altimetry maps from a realistic configuration of the Argo network by using the
10 actual uneven positions of the Argo floats in the basin. Figure 6.a displays the real positions of
11 the 58 Argo floats operating in the Mediterranean Sea on 22nd December 2014. SLA at each
12 single Argo float position was extracted from the original altimetry map of that day (Figure not
13 shown). Then, the SLA field for the whole basin was retrieved by following the procedure
14 applied to the regular-gridded sub-sampled fields.

15 On the other hand, and since the mean number of Argo floats in the Mediterranean is set
16 to around 80, random virtual floats were added to the actual Argo array of that day. The aim
17 was to reach the mean number of platforms normally operating on the basin. The virtual floats
18 were added by using a normal distribution function computed from the mean and standard
19 deviation of the positions of the Argo Array in the Mediterranean. Then, the SLA data was
20 obtained at the locations of both the actual and virtual floats (see Figure 6.b). We kept on
21 adding random virtual floats until reaching an Argo array of 150, 250 and 450 stations. Their
22 locations and the corresponding SLA data extracted at each position are respectively displayed
23 in figures 6.c, d and e. SLA field for the whole basin was then recovered for each configuration
24 of the Argo array according to the procedure described in the section 4.1. Reconstructed SLA
25 fields were compared with the original altimetry map of that day. Figure 7 summarizes the
26 results obtained from both the uneven and regular-gridded experiments conducted on 22nd
27 December 2014. The errors associated with the SLA maps recovered from the different
28 configurations of the Argo array (gray triangles) present a maximum RMSE of nearly 5 cm
29 when only the 58 Argo floats operating that day are used to reconstruct the SLA field. As
30 expected, RMSEs decay as the number of Argo floats increases (notice that here an Argo array
31 configuration with 750 floats has been also included in order to have a better overview of their
32 general pattern). This decrease follows the same pattern that the RMSEs obtained from the



1 regular-gridded experiments (black line) although larger values are observed here. This fact is
2 related to the uneven spatial distribution of the Argo platforms in the basin.

3 **5- Discussion**

4 The Argo network in the Mediterranean Sea consists presently of around 80 operating
5 floats drifting with less than 2 degrees mean spacing. Even though this array improves the
6 global coverage of the Argo network, it only captures the large-scale circulation features of the
7 basin. In this work, we have investigated which configuration in terms of the spatial sampling
8 of the Argo array in the Mediterranean would be necessary to recover the mesoscale dynamics
9 in the basin as seen by altimetry. The monitoring of the mesoscale features is not an Argo
10 program target. However, this issue is of concern since it can help the current ocean state
11 estimates.

12 To do that, we have conducted several Observing System Simulated Experiments (OSSEs) in
13 the basin. The errors of Argo – SLA differences required to perform the OSSEs were obtained
14 through the comparison of SLAs from altimetry and DHAs computed from the in-situ Argo
15 network. The comparisons have been focused on the sensitivity to the reference level (400
16 dbar or 900 dbar) used in the computation of the Argo dynamic height. We found that the
17 number of Argo profiles reaching 900 m used to compute DHA is almost 6 times smaller than
18 those reaching 400 m. Therefore, the choice of the reference depth has repercussion in the
19 number of valid Argo profiles and thus in their temporal sampling and the coverage of the Argo
20 network used to compare with altimeter data. In addition, the computation of the differences
21 between altimetry and Argo data referred to both 400 and 900 dbar revealed a standard
22 deviation of SLA – DHA differences 1.67 cm lower (in terms of variance) when computing DHA
23 referred to 400 dbar. This fact, together with both a higher correlation coefficient between
24 both datasets and the larger number of available profiles, make that 400 dbar should be
25 considered as reference level to compute DHA from Argo data in the Mediterranean basin.
26 This leads to a standard deviation of the differences between both datasets of 4.92 cm
27 (equivalent to 90 % of SLA signal variance). Conversely, one would expect better results when
28 using 900 dbar as reference level because the physical content (variance) of a larger fraction of
29 the water column is considered when computing Argo DH. However, the climatology used here
30 to compute DHA could be not as accurate at that level due the lower number of historical data
31 available at that depth, then resulting in larger standard deviations of the differences between
32 both datasets.



1 Another interpretation of the results obtained here could be done in terms of the
2 dynamics of the water masses residing in the Mediterranean Sea. Due to the excess of
3 evaporation over precipitation and river run-off, an Atlantic inflow through the Strait of
4 Gibraltar is required to balance the salt and freshwater budgets of the basin. As the Atlantic
5 water spreads into the Mediterranean, it becomes saltier and denser under the influence of
6 intense air-sea interactions [Criado-Aldeanueva *et al.*, 2012]. Most of this flow will return to
7 the Atlantic Ocean as Levantine Intermediate Water (LIW), formed during winter convection in
8 the Levantine sub-basin while another part will be transformed into deep waters along the
9 basin [Criado-Aldeanueva *et al.*, 2012]. The LIW spreads over different fractions of the water
10 column along its path towards the Atlantic Ocean. In some regions of the Mediterranean the
11 reference level of 400 dbar (near 400 m depth) would be close to the interface between this
12 water mass and those residing at deeper levels, which usually have different pathways. As a
13 consequence, velocities around 400 m depth would be significantly reduced as a result of
14 friction while they could be enhanced as we move towards deeper levels fed by the
15 Mediterranean deep water masses. As a result, velocities at 900 m depth could not be close to
16 zero, as we assume in the DHA computation, then promoting coarser results when comparing
17 altimetry with Argo data referred to 900 dbar. The depth of the LIW core in most of the
18 Mediterranean basin is also the reason of choosing 350 m as the parking depth for the Argo
19 floats in the Mediterranean [Poulain *et al.*, 2007].

20 Results reported from the regular-gridded experiments have shown that the reconstructed
21 SLA maps from a configuration similar to the current Argo array in the Mediterranean (spatial
22 resolution of $2^\circ \times 2^\circ$) are not able to capture the mesoscale features of the basin. As a
23 consequence, these maps only retrieve 21 % of SLA signal variance. This is an expected result
24 because the initial target of the Argo program is to monitor the large-scale ocean variability.
25 Quite the opposite, reconstructed SLA fields from a $0.75^\circ \times 0.75^\circ$ grid box of SLA observations
26 retrieve 66 % of SLA signal variance. This reconstruction captures the large-scale signal and
27 most of the mesoscale features of SLA fields in the basin exhibiting a mean RMSE lower than 3
28 cm (equivalent to 34 % of SLA signal variance). In addition, this spatial resolution agrees with
29 the theoretical one extracted from the internal Rossby radius of deformation computed for the
30 Mediterranean basin. The same outcomes were also obtained from the experiments
31 conducted by using the actual positions of the Argo array in the basin. Here, larger values for
32 the RMSEs of the recovered SLA maps were systematically obtained due to the uneven spatial
33 distribution of the Argo platforms in the basin. However, we must be cautious about these
34 results because the test has been conducted only along one Argo cycle (10 days). Anyway,



1 similar results to the ones obtained here are expected to emerge from longer experiments
2 according to the outcomes obtained from the analysis of 2014 yearly RMSEs associated with
3 the altimetry maps recovered from the different regular-gridded sub-sampled fields.

4 To summarize, and in light of a hypothetical future expansion of the Argo mission towards
5 an increase in the spatial sampling resolution, the actual Argo array in the Mediterranean Sea
6 might be enlarged until reach a spatial resolution of nearly 75×75 km according to the results
7 of the simulation experiments. Such Argo array, equivalent to around 450 floats, cycling every
8 10 days would be enough to retrieve the SLA field with an RMSE of 3 cm for spatial scales
9 higher than 150 km, similar to those captured by the altimetry. This array would also have a
10 net impact on numerical models that assimilate Argo profiles.

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16 International Argo Program and the national programs that contribute to it
17 (<http://aego.ucds.edu> and www.jcommops.org/argo). Altimetry data are generated, processed
18 and freely distributed by CMEMS (<http://marine.copernicus.eu/>).

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	All valid profiles (DHA ref. 900 dbar)		Profiles reaching 900m (DHA ref. 400 dbar)		All valid profiles (DHA ref. 400 dbar)	
Argo Floats	23		24		41	
Argo Profiles	416		479		2258	
std (SLA-DHA,cm)	5.31	0.20	5.04	0.17	4.92	0.07
R (SLA-DHA)	0.80	0.02	0.82	0.02	0.76	0.01

1

2 Table 1: Comparison of correlation and standard deviation (cm) of the differences between new
 3 AVISO product for the Mediterranean Sea and Argo data referred to both 400 dbar and 900 dbar
 4 (sub-columns on the left). Sub-columns on the right display the results of the robustness
 5 experiments in terms of standard deviations (see text for details). DHA referred to 400 dbar has
 6 been computed for the whole valid Argo profiles and those reaching 900 m depth for
 7 comparison purposes. The number of Argo platforms and vertical profiles used are also showed.

8

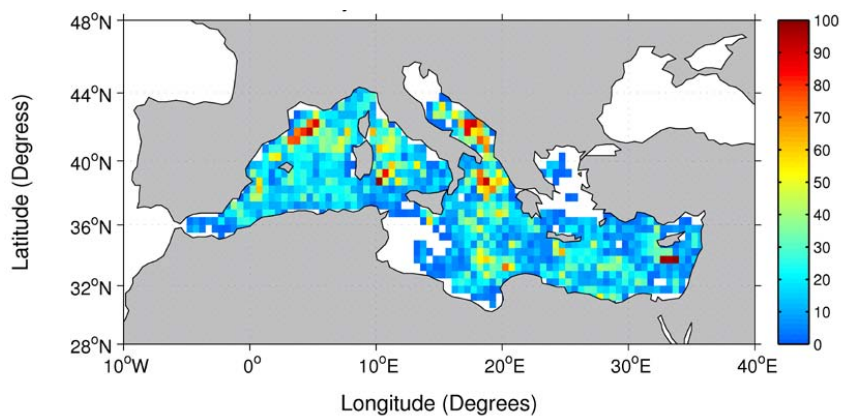


Spatial resolution (degrees)	Number of stations	Filtering scale (km)
2°×2°	69	445
1.5°×1.5°	121	333
1°×1°	273	225
0.75°×0.75°	482	167
0.5°×0.5°	1082	111
0.4°×0.4°	1458	95
0.3°×0.3°	1915	82
0.125°×0.125°	17283	—

1

2 Table 2: Spatial resolution (degrees) and associated number of stations of the different sub-
 3 sampled fields used to reconstruct the SLA in the Mediterranean. The lower line displays the
 4 spatial resolution and stations of the original altimetry maps. The filtering scale (km) used to
 5 compute the recovered SLA fields in the different reconstructions have been also included.

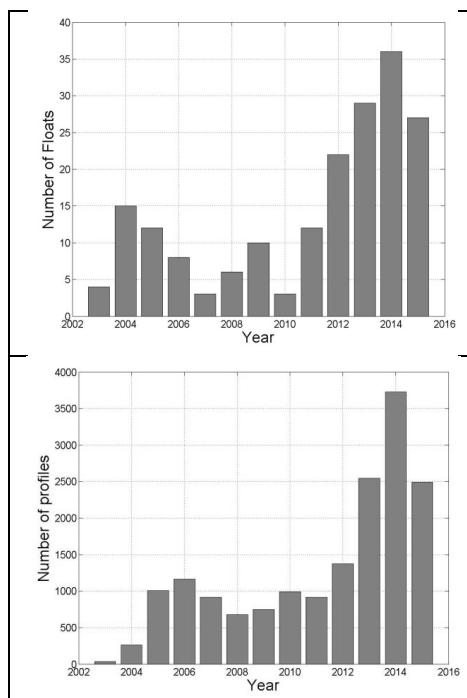
6



1

2 Figure 1. Number of Argo profiles on boxes of $0.5^\circ \times 0.5^\circ$ of lat-lon performed between 2003
3 and 2015 in the Mediterranean Sea and used to compute Argo DHs. Only profiles with a
4 position quality flag of 1 (good data) have been considered.

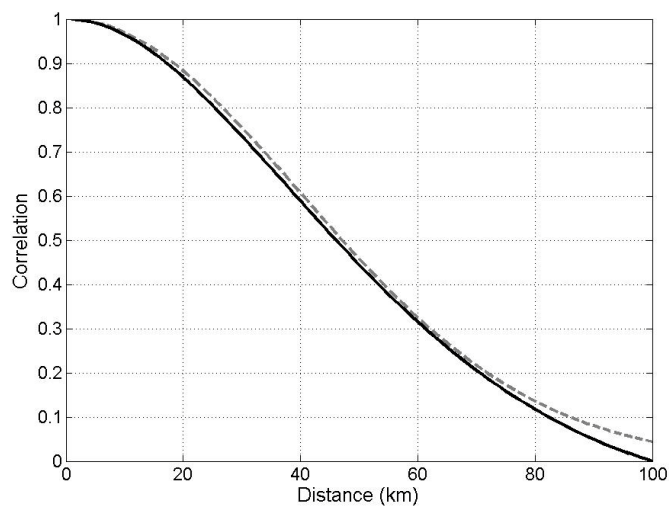
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2 Figure 2: Temporal evolution of Argo floats (upper panel) and Argo profiles (lower panel) with
3 a position quality flag of 1 deployed in the Mediterranean Sea since 2003 until the middle of
4 2015.

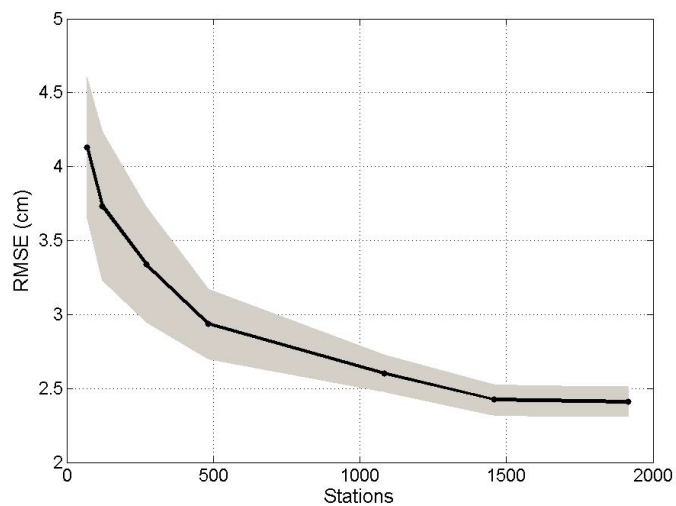
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1

2 Figure 3: Correlation curve computed for altimetric data (black solid line) for a typical zonal
3 scale of correlation for the Mediterranean region of 100 km. The gray dashed line shows the
4 best fitting correlation curve obtained for the reconstruction experiments. It corresponds to a
5 spatial scale of correlation of 40 km.

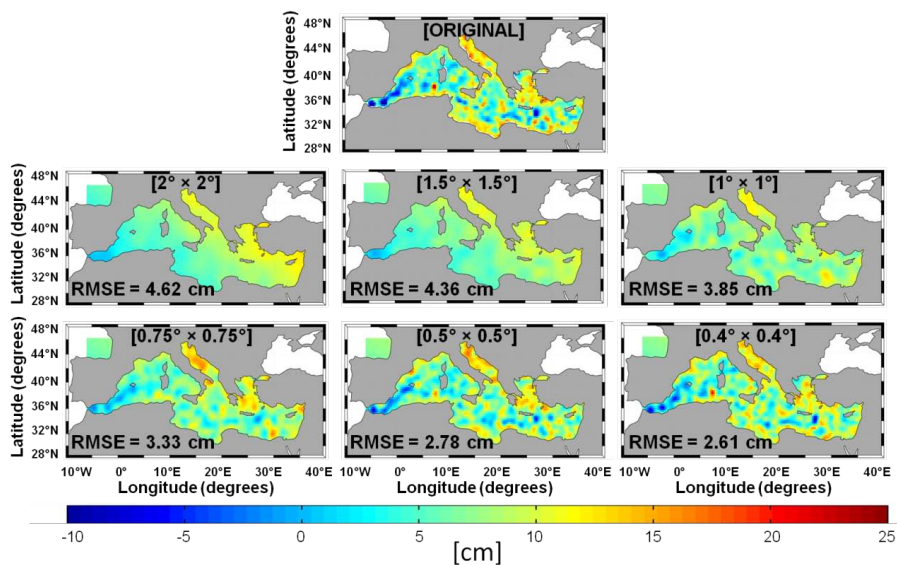
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1

2 Figure 4: Root mean square errors (cm) associated with the altimetry maps recovered along
3 2014 from the different regular sub-sampled fields mentioned in the text. The black line
4 represents the yearly mean value and the gray patch stands for the annual variability.

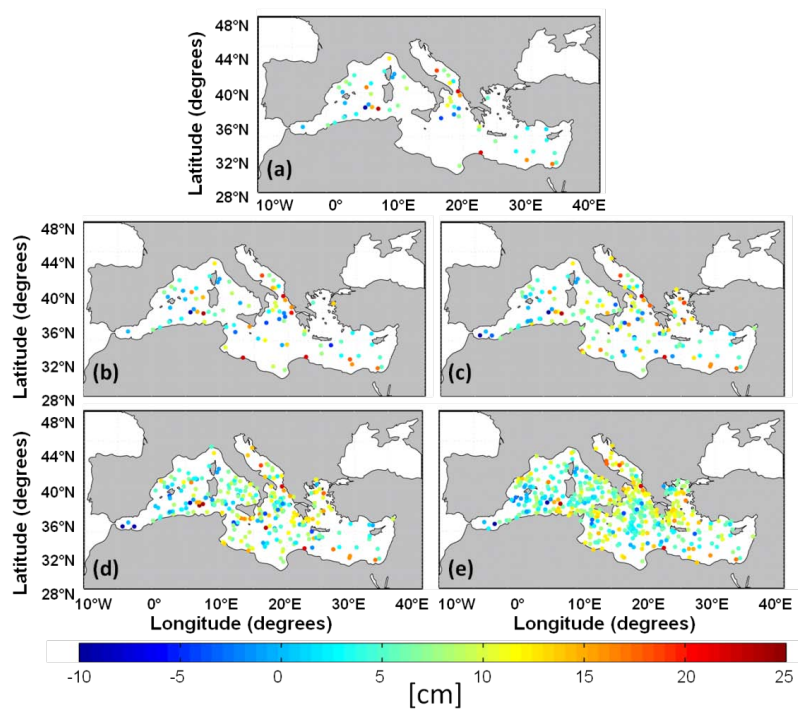
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1

2 Figure 5: Altimetry maps recovered from the different sub-sampled SLA fields (cm) on
3 December 22, 2014. The spatial resolution of the different regular grids and the RMSEs
4 associated with each reconstruction for that day are also indicated. Moreover, the original SLA
5 field of that day interpolated to a spatial resolution of $\frac{1}{3}^\circ \times \frac{1}{3}^\circ$ is displayed in the uppermost
6 panel for comparison purposes.

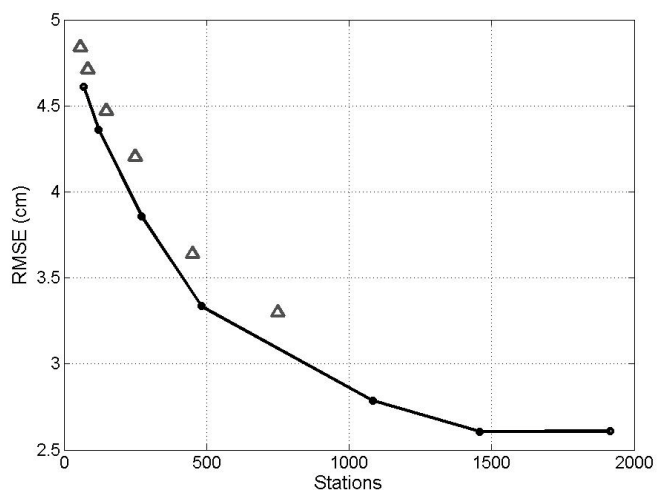
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1

2 Figure 6: (a) actual positions of the Argo array operating in the Mediterranean basin on
3 December 22, 2014 (58 floats). Colors indicate the SLA (cm) extracted at those locations from
4 the original altimetry map of that day. Panels (b), (c), (d), and (e) display the original Argo array
5 enlarged with random virtual floats in order to simulate an Argo array configuration of 84, 150,
6 250 and 450 floats, respectively.

7



1

2 Figure 7: Root mean square errors (cm) associated with the altimetry maps recovered on
3 December 22, 2014 from the different regular sub-sampled fields mentioned in the text (black
4 line). Triangles stand for the errors associated with the SLA fields retrieved for that day from
5 the different configurations of the Argo array in the Mediterranean Sea (see Figure 6). Notice
6 that an Argo array configuration with 750 floats has been also included for comparison
7 purposes.

8